Methodology for Simulation and Analysis of Preemptive Rebooking for Cancelled Airline Passengers

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METHODOLOGY FOR SIMULATION AND ANALYSIS OF PREEMPTIVE REBOOKING FOR CANCELLED AIRLINE PASSENGERS

by

Sanja Avramovic
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Fairfax, VA
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A Dissertation submitted in partial fulfillment of the requirements for the degree of Doctor of Philosophy at George Mason University

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DEDICATION

To my daughter Tara, my biggest inspiration, my husband Ivan, who is always my support, and my parents Radmila and Slobodan, who taught me the importance of imagination and discovery.
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LIST OF ABBREVIATIONS

Air Navigation Service Providers ................................................................. ANASP
Airfares Not Refund ......................................................................................... ANR
Airline On-Time Performance .......................................................................... AOTP
Airline Operations Center ................................................................................ AOC
Airline Operations Control Centers ................................................................. AOCC
Airline Origin and Destination Survey .......................................................... DB1B
Airline Service Quality Performance ............................................................. ASQP
American Airlines ......................................................................................... AA
Antonio B. Won Pat International Airport (Guam International Airport) ........ GUM
Baltimore Washington International Airport .................................................... BWI
Bureau of Transportation Statistics ................................................................ BTS
Chicago Midway International Airport ......................................................... MDW
Chicago O'Hare International Airport .............................................................. ORD
Collaborative Decision Making ........................................................................ CDM
Computer Reservation Systems ....................................................................... CRS
Corporates Travel Expense Savings .............................................................. CTES
Dallas Love Field .......................................................................................... DAL
Dallas/Fort Worth International Airport ....................................................... DFW
Denver International Airport ......................................................................... DEN
Department of Transportation ......................................................................... DOT
Detroit Metropolitan Wayne County Airport .................................................. DTW
Enhanced Traffic Management System .......................................................... ETMS
George Bush Intercontinental Airport ......................................................... IAH
Hartsfield–Jackson Atlanta International Airport ............................................. ATL
International Civil Aviation Organization ...................................................... ICAO
Irregular airline flight operations ..................................................................... IROPS
LaGuardia Airport ........................................................................................ LG
Logan International Airport .......................................................................... BOS
Los Angeles International Airport ................................................................. LAX
McCarran International Airport ..................................................................... LAS
Narita International Airport .......................................................................... NRT
National Airspace .......................................................................................... NAS
Newark Liberty International Airport ............................................................. EWR
Official Airline Guide ..................................................................................... OAG
Origin Destination ........................................................................................ O-D
Origin Hub Destination ................................................................................ O-H-D
Orlando International Airport ....................................................................... MCO
Overnight Passengers .................................................................................... SPAX
Passenger delay costs ...................................................................................... PDM
Passenger disruption costs ........................................................................... DPM
ABSTRACT

METHODOLOGY FOR SIMULATION AND ANALYSIS OF PREEMPTIVE REBOOKING FOR CANCELLED AIRLINE PASSENGERS

Sanja Avramovic, Ph.D.
George Mason University, 2015
Dissertation Director: Dr. Lance Sherry

Although, on average, only 2.1% of airline flights are cancelled each year, some of these cancellations occur in batches due to events that impact network operations such as snow storms, equipment outages, and labor issues. Batch cancellations impact a large number of passengers at the same time in the same location and have a negative effect on airline revenue (due to refund obligations), corporate profits (due to unbudgeted costs of employee travel), and overall traveler experience. Since some of these batch flight cancellations can be now be accurately predicted in advance, along with changes in the National Airspace System (NAS) that allow airlines to plan cancellations in advance (i.e. Collaborative Decision Making programs) and the ubiquity of inexpensive and reliable broadband communication services for communication between airlines and passengers, it is now possible for airlines to migrate from a concept-of-operations of rebooking
passengers after the cancellation event, to re-book passengers before the large scale cancellation event (i.e. pre-emptive rebooking).

This dissertation describes a method for Monte Carlo analysis of the feasibility and benefits of pre-emptive rebooking of airline passengers. Five case-studies of one-day cancellation events for major U.S. network carriers hub showed that: (i) pre-emptive rebooking is feasible accommodating more than 70% of the passengers seeking to re-book pre-emptively, (ii) rebooking passengers on the same day before the departure time of the cancelled flight, accommodates a majority of the passengers (i.e. previous day rebooking is not required), (iii) airlines can recoup up to an average of $297K per one-day event at each hub in airfare refund obligations, and (iv) corporations sponsoring business travel can save collectively up to an average of $49K on unplanned travel expenses for one-day event at each hub. The implications of these results for airlines, corporations and travelers are discussed.
1. INTRODUCTION

During the course of a year, there are a number of reduced flight capacity events in which, due to increased coordination amongst National Airspace (NAS) operational stakeholders, flights are cancelled by airlines *in advance* of the scheduled departure time. For example, this may occur in preparation for a major snow storm impacting a hub airport, in anticipation of a labor action, or in anticipation of staffing or equipment shortages. In 2012, there were 273 events for which airlines cancelled 5 or more flights departing from a given airport.

The traditional concept-of-operations for an airline is to *wait-and-see*. It is in the airline’s best interest to wait until two hours before the scheduled departure to announce the flight cancellation. Passengers are then rebooked on alternate flights by on-site customer service agents, or increasingly, self-service kiosks and on-line websites. The outcome of this concept-of-operations is reduced options to rebook passengers, resulting in overnight passengers, and adverse impact on passengers’ ability to arrive on time for time-critical scheduled events (e.g. business meetings, life-cycle events, sporting events, performances, etc.).

This thesis describes an alternative approach to mitigating the effects of IROPS based on exploiting the short term forecasts of cancellations and flexibility in airline passenger travel plans to rebook passengers in a preemptive manner, in advance of the anticipated IROPS. This concept-of-operations is made feasible by four changes: (1) improved forecast accuracy of airspace bottlenecks by Air Navigation Service Providers
(ANSPs), (2) dissemination of airspace information and coordination of slots by
Collaborative Decision Making (CDM), (3) ubiquitous availability of wireless broadband
for immediate communication with passengers, and (4) internet reservation systems for
rebooking.

This thesis examines the feasibility and benefits of preemptive rebooking on the
same day before the ticketed scheduled departure time, and on the previous day.
Extensive case-studies of preemptive rebooking for U.S. domestic operations, undertaken
in this thesis using a Monte Carlo assignment of passengers willing to be rebooked on an
erlier flight, showed that:

- Preemptive rebooking in advance of forecast large-scale cancellation events
  is feasible.
- About two thirds of the passengers seeking preemptive rebooking can be
  accommodated before their original scheduled flight time. The remaining
  one third of passengers seeking preemptive rebooking cannot be re-
  accommodated due to insufficient seats.
- Preemptive rebooking on the previous day was not required as it did not
  significantly change the percentage of passengers re-accommodated.
- Airlines are able to recoup up to just over a half in airfares that would
  otherwise need to be refunded.
- Corporations sponsoring business travel for their employees also benefit by
  savings of a similar magnitude, up to 53% of unbudgeted overnight costs,
  which would need to be paid if Preemptive Rebooking was not used.
- Passenger Trip Delay (PTD) and the number of overnight and passenger which cannot be rebooked decrease by a half (up to 50%), and the decrease in the number of passengers that cannot be rebooked at all is 55%.

  The results of the case-studies show the way in which increased travel flexibility for airline passengers, enabled by rapid, inexpensive wireless/broadband capabilities for communication, yields win-win results both for airline passengers and airlines.

  A degree to which flight cancellations may be subject to preemptive rescheduling is evident in the statistics from U.S. domestic operations in 2012, where there were over 3000 events in which an airline cancelled more than 5 flights (tactical cancellations as opposed to mechanical failure) on a given day, departing a given airport or arriving at a given airport.

  The preemptive rebooking concept of operations creates a win-win scenario for all stakeholders. Airlines are able to recoup revenues that would otherwise be lost to refunds. Airlines are also able to distribute the rebooking task in advance of the IROPS, taking the pressure off after the IROPS when resources are the scarcest. Airlines increase their flight operations flexibility for mitigating IROPS and maintain a greater degree of customer goodwill.

  By taking advantage of preemptive rebooking, Corporates can reduce their loss in cost for overnight stays, transportation, and similar travel expenses that would be required by their travelling employees.

  Passengers who are offered preemptive rebooking would maintain control of their destiny, by making decisions on the value of alternate travel plan options given the
cancellation of their flight. The advance warning provides them the chance to avoid last
minute changes in plans that can affect business relationships and attendance at fixed-time
events.

Airports and ANSP also benefit by avoiding passenger overnight stays at airports
as well as excessive demand for flights in capacity constrained conditions.

This thesis is organized as follows: Chapter 1 is the introduction. Chapter 2
provides a literature review. Chapter 3 describes the proposed method and the associated
models. Chapter 4 describes the method of analysis and detailed results of one case.
Chapter 5 includes a discussion of the results, future work and conclusions.

**Airline Transportation Network**

Airlines schedule flights to operate in a time-space network. In this network
airports are represented as nodes, and aircraft movements are arcs. The arcs may appear
in the same airport or between airports, and wraparound between the first and last time
node.

Well-designed itineraries (and their corresponding networks) are made so as to
maximize revenue, by meeting passengers travel demands with appropriate seat capacity,
and by minimizing costs through using the most cost-effective aircraft, maximizing
aircraft utilization, and minimizing the impact of disruptions.

Hub-and-spoke networks are centralized networks, designed like a chariot wheel,
with a hub as a center, connecting all spokes. Passengers travel from their origin to their
destinations, via the hub. The benefits of this kind of networks are that only \( n-1 \) routes are
necessary to connect all \( n-1 \) spokes, which increases efficiency and lowers operating
costs. This means that an airline can operate many origin-destination markets with fewer departures and fewer aircrafts. A centralized model decreases flexibility, increases scheduling complexity, and introduces a single point of failure (bottleneck).

An example of a hub-and-spoke network, featuring United Airlines, is shown in figure 1. The hubs are in Chicago (ORD), Newark (EWR), Boston (BOS), and San Francisco (SFO).

Another major airline transportation network type is point-to-point. In this network a central location does not exist, and \( \frac{n(n-1)}{2} \) routes are required to connect every node in the network. Often point-to-point networks operate focusing on non-stop flights with no connecting passengers, which improves aircraft and crew productivity.

Southwest Airlines is an example of an airline which uses a point-to-point network, but it uses “airport of interests” to connect passengers. Those airports are Chicago (MDW) Las Vegas (LAS), Baltimore Washington (BWI), Dallas (DAL), Phoenix (PHX), Denver (DEN), Houston (HOU), Atlanta (ATL), Los Angeles (LAX) and Orlando (MCO).
Irregular Operations

Even if cancellations are not very frequent events, their impact can be big. Daily Number of Cancelled Flights during 2012, for the UA network is shown in the figure 2. There were a total of 82 days which had more than 10 cancelled flights. A total of 7 days had more than 100 cancellations. The largest number of cancelled flights was 720 (on October 29th), while the biggest event (a 4-day cancellation event) involved 1860 cancellations total.

This dissertation shall only analyze one day events.
Figures 3-6 show the number of cancelled flights daily for the hubs ORD, DEN, SFO and EWR.
Figure 3 ORD: Number of cancelled flights
Figure 4 SFO: number of cancelled flights

Figure 5 Den: Number of cancelled flights
The largest number of cancellations per day was through EWR – 265 cancellations, then ORD – 153 cancellations, SFO – 99 cancellations, and DEN – 63 cancellations.

The statistics for Southwest Airlines during the same year look like this (Figure 7):
On Southwest, the majority of days had no more than 10 cancelled flights (129 days total). 140 days had more than 20 cancelled flights per day, out of which 14 days had more than 100 cancelled flights each. The largest number of cancelled flights on a single day was 650.

Chicago Midway (MDW) had the largest number of cancellations of all Southwest hubs (Figure 8):
There were 163 days during 2012 on which there were no more than 10 cancelled Southwest flights at the MDW hub. Just 16 days had more than 10 cancelled flights, but the number of cancelled flights had a high upper bound of 160 per day. Although these events are not frequent, the impact of the cancellations is large.

For example, on January 20th, Southwest cancelled 96 flights (inbound and outbound MDW) impacting 11,000 passengers. On the same day, 7140 passengers were rebooked, and 3560 passengers stayed overnight as a result of next-day rebooking. 300 passengers couldn’t be rebooked at all. If it is assumed that all of the passengers that
could not be rebooked on the same day would ask for the airfare refund, the estimated cost to the airline would be $1.5M for this event.

Cancellations may arise due to different circumstance, for example a major snow storm impacting a hub airport, or in anticipation of a labor action, or staffing shortfall or equipment outage. In February 2015, Southwest Airlines grounded 128 aircrafts after they discovered that the airplanes had not received required inspections. In the situation that the imminence of a big disruption is known, an airline has time to find better rebooking options for passengers. The idea of this research is to analyze the use of preemptive rebooking to add more options for the passengers’ recovery process. This means that instead of the wait-and-see approach, announcing flight cancellations at the last minute, passengers would be inexpensively contacted in advance of the cancellation to coordinate rebooking, in a way that leverages the willingness of the passengers to accept a flight earlier than the scheduled (now cancelled) flight. Advances in communication technology, in particular broadband wireless access, would allow the possibility for airlines to instantaneously, inexpensively contact all ticketed passengers on a cancelled flight to provide options for rebooking.

For the airline, preemptive rebooking increases revenue by avoiding reimbursement costs for passengers that no longer want to travel, and can improve customer relations. For the passenger, the preemptive recovery capability provides additional options to meet their attendance obligations by arriving in advance.
**Concept of Operations**

Two important changes have enabled the ability to improve passenger mobility, in the event that irregular operations which lead to large scale cancellations can be foreseen. Those two things are increasingly pro-active planning on the part of airlines, and the development of fast and efficient technologies for communicating to passengers. The former is the result of increased coordination amongst National Airspace (NAS) operational stakeholders, and the second comes in the form of inexpensive and reliable broadband communication services. In this chapter, a new concept of operations will be described: Proactive Preemptive Rebooking Concept.

**Traditional concept of operations: Reactive**

The traditional concept of operations for rebooking passengers for irregular operation is reactive. This means that the airline waits until a few hours before departure time to cancel the flight. Possible reasons for this include:

- Cancelling as late as possible leaves options open for achieving an on-time flight through aircraft *swapping*, in the case that the cancellation is the result of a maintenance issue.
- Cancelling as late as possible leaves options open for achieving an on-time flight through *slot allocation*, in the case of a ANSP coordinated delay program.
- There was previously no way to quickly and inexpensively contact passengers and re-accommodate them to suit their needs, thus there was no significant advantage to announcing cancellations earlier.

Due to the fact that there is no contact with passengers before they arrive at the airport, passengers can be re-booked on alternate flights by on-site customer service agents, self-service kiosks or on-line websites.
Often, not all passengers can be re-accommodated on the same day, so remaining passengers will be rebooked on flights the following day (overnight passengers). This increases travel costs for all passengers and adds unbudgeted travel costs to corporations whose employees are on business travel. If passengers choose to cancel their tickets, it would result in a loss of revenue for the airline, due to the need to refund airfares. The sequence of events from a passenger view point for the traditional and proposed mode of operations are described below, and summarized in the Event Sequence diagram in Figure 10.
Figure 9 - Sequence of events diagram from Post Irregular Operations rebooking: traditional (reactive) Concept of Operations
Post Irregular Operations Rebooking sequence of Events

As shown in Figure 10, the airline starts the day with an assessment of expected operations. On days where flights must be cancelled due to severe weather events or ANSP delay programs, the flights that are likely to be affected (e.g. cancelled or delayed) are identified. Due to uncertainty in the airspace system, airlines keeps their options open by listing the flight status to the ANSP and passengers as scheduled on-time.

As the scheduled departure time of a flight approaches, the airline decides to cancel it. This information is provided to the ANSP as well as to the passenger. The information to passengers is posted in the sense that is made available (i.e. through airport flight status monitors, airline websites, etc.), but not sent directly to the individual passenger.

At this time, most airlines reservation systems unilaterally automatically rebook the passenger based on next available flight. Preference is given to passenger considered high value passengers by Frequent Flyer status or other indications. The initial rebooking is made to expedite airline operations.

At this time the passengers can use customer service agents at the airport, reservation kiosks at the airport, or airline websites to accept or renegotiate the rebooking.

Unless a destination has “shuttle” operations with high frequency, it is unlikely that a passenger on an itinerary with a cancelled flight will be rebooked without a significant amount of time before the next flight. Further, passengers on an itinerary with a cancelled flight late in the day will generally be required to overnight. In some cases the passenger will not be able to arrive at the destination in time for a planned event.
Although many of large scale cancellation events can be forecast (i.e. snow storms, labor issues, and equipment outages), uncertainty in Air Traffic Control (ATC) and in competing airline response to these events obliged airline’s to reactively rebook passengers after the event. However, the advent of Collaborative Decision Making to enable equitable advance planning by the airlines (Wambsganns, 2001), along with the ubiquity of broadband wireless communication between airline and passengers, now enables airlines to proactively cancel flights in advance and rebook passengers before the cancellation event.

**New concept of operations: Proactive Preemptive Rebooking Concept**

With increased coordination amongst National Airspace (NAS) operational stakeholders, there are reduced airport and airspace capacity events (e.g. snow storms, labor shortages, infrastructure closures, equipment failures) in which flights can be cancelled by the airlines well in advance of the scheduled departure time. The development of communications, which have become ubiquitous, instantaneous, and inexpensive, has enabled a new concept of operations: proactive, preemptive rebooking.

Collaborative decision-making by ATC and airlines, has changed the way the fight operations are managed from reactive to pro-active. The proactive mode allows airlines to make decisions in advance of events. This means that capacity shortfalls at the runway are now identified at the start of the day. Sometimes the impacted flight operations can be identified in advance (from 48 to 72 hours), in the case of:

- a planned shutdown, for example runway closure, or
- predictable severe weather, for example a heavy snow storm.

As a result, some flight cancellations are known well in advance.
The other significant factor that was mentioned is that modern passengers can be contacted via social networks, text messages or emails, in order to check their willingness to re-schedule flights. In this way, it is possible to better manage large scale cancellation events. The Internet coupled with interactive airline reservation systems and broadband/wireless communication now enables airlines to contact and coordinate changes in reservations at speed and low cost considered unimaginable a decade ago.

These factors have all conspired to provide an opportunity to change the usual operations paradigm and facilitate more choice in addressing irregular operations from a passenger standpoint. The combination of airline passengers travel flexibility with rapid, inexpensive wireless/broadband capabilities to communicate with passengers, and pro-active Flow Management can yield a win-win for airlines, passengers, and corporations that rely on business travel. Airlines can increase revenue by avoiding reimbursement for passengers that no longer want to travel (Airfare Not Refund, ANF) and improve customer relationship (based on trust). Corporates can reduce unexpected costs on corporate travel (Corporate Travel Expanse Savings, CTES). Passengers, have a greater chance to meet their attendance obligations by arriving in advance (lower PTD, decrease in number of overnight passengers and passengers that cannot be rebooked).

A sequence of event diagram depicting rebooking under Post Irregular Operations circumstances, showing the proactive preemptive rebooking concept, is shown in Figure 11.
Flight Status = Cancelled

Here are your rebooking options
Thank you I’ll take an earlier flight

PAX

Airline Evaluates Day of Operations

Airline selects flights to be cancelled

EARLIER FLIGHT

Travels to Airport

Ground Transportation to Event

PROACTIVE (PRE-EMPTIVE)

Event
In this scenario, the airline starts the day with an assessment of expected operations. On days where flights must be cancelled due to severe weather events or ANSP delay programs, the flights likely to be affected (e.g. cancelled or delayed) are identified. Due to the level of collaboration and planning available from the ANSP, the airline immediately lists the flight status to the ANSP and passengers as cancelled.

At this time, the airline proactively contacts the passenger and offers preemptive rebooking options. The passenger selects the option that works best, under the circumstances, for the passenger. The passenger adjusts their personal schedule and is able to take advantage of the proactive rebooking and arrive on-time for their event at the destination.
Figure 11 Sequence of events diagram from Post Irregular Operations rebooking: Comparison between traditional and preemptive rebooking
Research Question: Preemptive Rebooking

There are five questions which this research intends to answer:

1. Does preemptive rebooking solve the problem of providing a suitable number of rebooking options? This is defined in terms of the fraction of passengers who are accommodated on preemptive flights. If passengers accept preemptive rebooking at a rate of 70%, then to be considered a satisfactory alternative, there must be at least enough preemptive seats available to rebook them on preemptive itineraries.

2. Is preemptive rebooking financially beneficial for airlines? If the assumptions is that for every preemptively rebooked passenger the airline saves $377 (DOT, 2012), the question is how much the airline will save in recouped revenue. It is assumed that in addition to recouped revenue, airlines employing preemptive rebooking will benefit through building a stronger trust relationship between the airlines and passengers.

3. Is preemptive rebooking financially beneficial for Corporates? Can Corporates reduce their loss in cost for overnight stays, transportation, and similar travel expenses that would be required by their travelling employees in case of cancellations?

4. Is preemptive rebooking financially beneficial for passengers? The costs to the passengers are modelled in terms of lost costs due to trip delays, overnight stays, and itineraries which cannot be rebooked. The passenger costs can be reduced through improving PTD, improving timeliness for private or business events, and lowering number of overnight passenger and passenger who cannot be rebooked. So if there is a net reduction in costs due to reduced delays, overnight stays, and the number of unbooked passengers, then preemptive rebooking is considered financially beneficial to passengers.

5. Is there an advantage to rebooking preemptively further than a day in advance (Same day early, Previous day)? Define the threshold for preemptive rebooking in sense the necessary time frame to rebooked passengers in advance.
Research Hypothesis

To answer the five research questions, the null hypotheses below have been formulated per question. Each must be false for the corresponding questions to be answered in the affirmative.

- \( H_0 \): Preemptive rebooking does not accommodate more than 10% of passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%).
- \( H_0 \): Preemptive rebooking does not result in additional revenue to airlines.
- \( H_0 \): Preemptive rebooking does not result in additional saving to Corporates.
- \( H_0 \): Preemptive rebooking does not result in reducing PTD and the number of overnight and passenger which cannot be rebooked.
- \( H_0 \): The number of additional passengers accommodated by adding an additional day of preemptive rebooking is greater than 25%.

Validation Plan

This thesis examines the feasibility of preemptive rebooking and the benefits to airlines and passengers. Historical data for flight statuses and passenger booking is available from the US Department of Transportation's Bureau of Transportation Statistics. This data includes airline on-time performance, flight delays, and cancellations from the Airline Service Quality Performance (ASQP) System; aircraft inventory and seat capacity from the Schedule B-43 Aircraft Inventory database; monthly aggregate airline load factors from the Form 41 Schedules T-100 database; a 10% ticketed sample showing passenger demand and route information from the Airline Origin and Destination Survey (DB1B); historical and future flight schedules from the Official Airline Guide (OAG); and on-time arrival data for non-stop domestic flights from the Airline On-Time Performance (AOTP) database.
The combined data is a base for computing average PTD for a day's flight schedule, to serve as a reference. Any model that is developed will produce its PTD estimate, which can be then compared to the historical reference value in order to verify the accuracy of the model.
2. BACKGROUND AND LITERATURE REVIEW

Airlines provide rapid, safe and affordable transportation of passengers between their origins and destinations. This service is critical to the national economy. The reliability of this service can be expressed by the difference between passengers' ticketed and actual arrival time.

A single aircraft flying from a departing airport to its next arrival destination is called a leg. Every leg flown by an aircraft is identified by a flight number, which is an identifying number consisting of one to four digits, with an optional alpha suffix code. One or more consecutive legs by an aircraft using the same flight designator (airline code, flight number, and operational suffix) is called a flight.

Every flight has a number of passengers, and each passenger has an itinerary. An itinerary is the combination of all portions of the passenger's trip, from origin to destination, which may include one or more flight legs. The portion of a flight from the point when a passenger boards to the point when he disembarks is called a flight segment. Flight segments separated by gaps may still be part of the same itinerary, and a single flight may have passengers with many different itineraries. For example, a flight traveling from ORD airport (Chicago) to SFO (San Francisco). In addition to the passengers traveling from Chicago to San Francisco, the flight may contain passengers with itineraries that take them to other cities, via San Francisco.

During the course of the day, any airline will encounter problems, including delays and cancellations of flights. A delay or cancellation may cause a passenger to miss a
subsequent flight in their itinerary. This is called missed connections. A passenger who has missed a flight due to a misconnection can have their reservations changed to put them on the next available flight in the direction of their destination, a practice which is known as rebooking.

The term flight status is used to denote the state of the flight with respect to its scheduled time. It could be on time, delayed or canceled. A flight which arrives within fifteen minutes of its scheduled time is still considered on time, but longer than that is considered a delay.

Every flight’s status could be on time, delayed or canceled. These flight statuses can apply to any flight, but within a flight cycle, subsequent flights can be affected by the status of the previous one. It means that if a flight is delayed, it makes it more likely that the next flight will also be delayed. A flight that is canceled surely means that the next one must also be canceled. If a flight is on time, the next one may also be on time, but it could still be delayed or canceled.

A flight with a particular tail number leaves an origin location (spoke) and reaches a destination (hub), and after some time, the next flight with the same tail number leaves the destination (hub) and returns to the origin of the round-trip (spoke). This refers to one flight cycle. During a day it is possible to have one or more flight cycles.

In this dissertation the reliability of service will be measured by passenger trip delay (PTD), which is the difference between the ticketed arrival time and the actual passenger arrival time. In general, less reliable service means more flight disruptions, which tends to lead to greater delays reflected in the PTD.
**Disruptions**

During a day different events, caused by airline resource shortages or airport and airspace capacity shortages may lead to irregular operations. Airline Operations Control Centers (AOCCs) are responsible for safety and managing of operations, communications between Air Navigation Providers and the other airlines, and disruption recovery.

Controllers in AOC respond to these disruptions in different ways, e.g. reassigning resources, or adjusting schedules. The schedule recovery process recovers aircrafts, crews and passengers from disruptions. Although there is similarity between flight and passenger itinerary disruptions, there are also important differences, as shown in the next table:

<table>
<thead>
<tr>
<th>Table 1Types of disruptions</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Flight Disruption</strong></td>
</tr>
<tr>
<td>Delayed</td>
</tr>
<tr>
<td>Cancelled</td>
</tr>
<tr>
<td>Diverted</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
</tbody>
</table>

Passengers itineraries can be disrupted by delayed or cancelled flights, but also by diversions, over-booking and missed connections.
**Recovery Process**

Previous research describes methods for addressing the general class of irregular operations including: robust scheduling (Clarke, 1998; Bratu, Barnhart, 2006; AhmadBeygi at al, 2010; Chiraphadhanakul, Barnhart 2013), and aircraft assignment (Eggenberg Salani, Bierlaire, 2010; Jafari, Zegordi, 2011). The effects of passenger rebooking has been analyzed by Yan et.al. (submitted 2015). Pre-emptive passenger rebooking for a possible missed connection for a forecast single delayed flight was discussed in McCarty (2012).

When the disruptions occur, scheduled operations may be adjusted in a several ways (Belobaba et al 2009). This includes delaying departures or postponing departure times (either to wait until an aircraft is ready, or to accommodate delayed passengers), canceling flights, rerouting planes, *swapping* (reassigning) aircraft between flights, and rebooking disrupted passengers. Any adjustments must be done within the framework of existing rules and regulations. Adjustments should also take into account estimated gate-to-gate ready times for flight legs.

Cancellations and delays may occur for a number of reasons, including mechanical problems, absent crew, disruptions to other flights or earlier legs of a flight, or airport capacity issues (especially due to weather), and even because of missed flight inspection. Southwest grounded 128 airplanes in February 2015, when discovered that they had not finished the required inspection by the deadline.

Flight cancellations are often used to bound the propagating effect of the delays to the rest of the system. When canceling a flight, the number of aircraft at each airport should stay roughly the same. The reason is that it is important to have enough aircraft to
service future flights. For this reason, cancellations are often done in pairs (two legs or two planes). To avoid overnight delays, the tendency is to cancel a morning leg rather than an evening leg. Cancellation decisions may also take into account aircraft crew. Since missing crew may prevent an aircraft from taking off, an aircraft is less likely to be canceled if the crew has connections on multiple different aircraft than if the crew stays together for their flights.

In the event of a disruption, the plan for recovery from the disruption may include several objectives, often based on minimizing costs such as spare aircraft, reserve crews, recovery time, or passenger time and loss of goodwill. Recovery plans are often done in phases, where recovery of planes is top priority, followed by crew. Passenger rebooking is handled only after flights and crews are accounted for. The sequential process is intended to speed up recovery decisions, and decision support tools tend to reflect the process by focusing on single phases of the process.

During the aircraft recovery phase decisions will be made involving flight leg cancellations, delays and aircraft rerouting. The goal of the aircraft recovery phase is to ensure that aircraft are present to service future flights.

In the second phase, the crew recovery phase, crews will be assigned to uncovered flight legs, either by reassignment or by calling in reserve crews. Crew can be flown as passengers in order to ensure that they are available where they need to be, or crew can be reassigned from another flight to fill in for the missing crew. A third option, using reserve crews, is a more expensive option, because both the reserve crew and the original crew need to be paid.
Finally, disrupted passengers will be rebooked. Because of the fact that a disrupted aircraft may have passengers with a number of different itineraries, a number of passengers are likely to have misconnections. According to a study by Bratu and Barnhart (2005), passengers who missed connections due to delays were very likely to be given the best (minimum delay) alternative to their original itinerary, whereas those that missed connections due to cancellation were about half as likely to be given the best alternative. This is due to the number of people needing to be rebooked in the event of the cancellation.

**Schedule recovery**

Schedule recovery should minimize cost, including the estimated cost of: reserve crews, spare aircrafts, passengers’ recovery, loss of good will and time. This process is sequential: first aircrafts, then crews and in the end passengers. Dividing the task speeds up recovery, but the big problem remains making decisions in uncertainty.

Airline schedules are optimized under an assumption of normal weather operating conditions, with slack time largely stripped from the schedule to maximize resource productivity. This leads to more difficult handling during disruptions (Lan et al., 2006).

**Slack time**

Slack time is defined as “additional time allocated beyond the expected time required for each aircraft connection, passenger connection, or flight leg” (Chiraphadhanakul, Barnhart, 2013). This means that lack of slack time leads to increased difficulty in recovery from disruption. However, maximizing resource utilization can result in plans that minimize ground time as much as possible. Chiraphadhanakul and
Barnhart proposed a flight schedule adjustment model that strategically re-allocates existing schedule slack to minimize some proxy of expected delays or disruptions, and introduced the notion of *effective slack*.

**Passenger Rebooking**

In most cases, recovery models deal primarily with aircraft, crew, and flight schedule recovery, in that order of priority. Recovery models developed by Bratu and Barnhart (2006) plan decisions to adjust the schedules of aircraft, crew, and passengers in response to delays and disruptions, in order to optimize passenger and operating costs. One of the models focuses on passenger delay costs (PDM), while the other focuses on passenger disruption costs (DPM). The models reduce delays and disruptions noticeably in a simulator, over a span of 3 days of operation under varying levels of disruption. The DPM is fast enough to be used during operations, whereas the PDM is too slow to be used in real-time during the single day window which is needed to make decisions.

**Constraint specific recovery network for solving airline recovery problems**

Eggenberg et al. model a problem as a “Recovery network”. In the model, the aircraft, crew and passengers are units. Each unit is associated with a specific recovery network (a set of nodes and arcs). Each feasible recovery method corresponds to a path through the nodes and arcs in the network. Each unit has a resource limit and the resource limit for passengers is the delay of their itinerary. The recovery network is based on a two dimensional coordinate system. The network generation algorithm is a dynamic programming algorithm.
The advantage of this approach is that the units and constraints are separated. Using a column generation algorithm is was possible to compute exact delays and recovery costs.

**Solving the Airlines Recovery Problem Considering Aircraft Rerouting and Passengers**

Le, Zhan, Wu (2008) proposed the model for aircrafts and passengers, with idea that each passenger must be delivered to his or her destination and to minimize total delay cost. The paper presents an objective function to minimize delay costs, and provides a genetic algorithm approach to finding an approximate solution.

**Preemptive Rerouting of Airline Passengers under Uncertain Delays**

Based on an assumption that delays are known in advance, McCarty and Cohn introduced preemptive rerouting of airline passengers, which occur before the length of the delay is realized, to minimize the sum of the delay costs for all passengers on the delayed flight. The idea is to rebook passengers proactively; as soon as it is known that a flight will be delayed instead waiting for missed connections. This model is a 2-stage stochastic model. In the first stage passengers are preemptively assigned to new itineraries. The second stage modifies itineraries for passengers who still miss connections. The second stage is modeled as a minimum cost model.

Suppose that a passenger has an itinerary consisting of two flights:

Flight A from EWR to DTW, departing at 12:00pm and arriving at 2:10pm.

Flight B from DTW to LAX, departing at 3:40pm and arriving at 6:25pm.

A mechanical delay occurs on Flight A, from 45 minutes up to 3 hours. Other itineraries that have available capacity:
Itinerary 1:

Flight C from EWR to CMH, departing at 12:30pm and arriving at 2:30pm
Flight D from CMH to LAX, departing at 5:30pm and arriving at 8:00pm

Itinerary 2:

Flight A from EWR to DTW, departing at 12:00pm and arriving at 2:10pm
Flight E from DTW to LAX, departing at 7:10pm and arriving at 9:55pm

If the delay on the first flight is 45 minutes, then the passenger will make the second flight and reach the destination on time. If the delay is three hours, it means that the second flight will be misconnected, and the new itinerary uses Flight E, arriving at 9:55pm.

If the passenger proactively switches to Flight C when the delay on her first flight is discovered, the passenger will arrive at the destination at 8:00pm, 1 hour and 35 minutes later than planned, rather than the 3 hours and 30 minutes that would result from misconnection.

McCarty and Cohn’s results showed that the model reduced the length of passenger delays, but the model takes into account just one delay, which is often not a realistic situation. In many cases there is a combination of delays, and sometimes just one delay can produce other, concurrent delays. The model does not scale well to large model.

**Passenger Trip Delay**

*Passenger Trip Delay* (PTD) was defined as the difference between scheduled arrival time and actual arrival time. This definition is derived from the definition of *Passenger trip time* which is the period of time between the scheduled gate departure time to the scheduled
gate arrival time. Passenger trip time is not flight time. It includes flight delays due to cancellations, missed connections and diversions (Sherry, Wang 2007).

Trip delays are a function of the number of passengers on itineraries in the time-space network of a flight. An itinerary is the sequence of flights taken by a given passenger in traveling from their Origin to Destination. An itinerary can be a direct itinerary, which consists of exactly one flight, or a connecting itinerary. One flight may have passengers with both direct and connecting itineraries on board.

PTD on direct and connected itineraries experience trip delays due several types of situations, shown in the table below.

<table>
<thead>
<tr>
<th>Passengers on Direct Itineraries (Origin-Destination)</th>
<th>Passengers on Connecting Itineraries (Origin-Hub-Destination):</th>
</tr>
</thead>
<tbody>
<tr>
<td>Delayed flight (Origin to Destination)</td>
<td>Delayed flights (Hub to Destination)</td>
</tr>
<tr>
<td>Rebooking due to cancelled flights (Origin to Destination)</td>
<td>Rebooking due to Cancelled flight (Origin to Hub)</td>
</tr>
<tr>
<td>Diverted flights (Origin to Destination)</td>
<td>Rebooking due to Cancelled flight (Hub to Destination)</td>
</tr>
<tr>
<td>Denied Boarding (Origin to Destination)</td>
<td>Rebooking due to Missed Connection (Hub)</td>
</tr>
<tr>
<td></td>
<td>Diverted flights (Hub to Destination)</td>
</tr>
<tr>
<td></td>
<td>Denied Boarding</td>
</tr>
</tbody>
</table>
When a flight is disrupted the associated itineraries will be disrupted as well.

This relationship between flight and itinerary disruptions is shown in the next table.

**Table 3 Relationship between flight and itinerary disruptions**

<table>
<thead>
<tr>
<th>Itinerary status</th>
<th>Direct Itinerary</th>
<th>Connecting Itinerary</th>
<th>PTD function</th>
</tr>
</thead>
<tbody>
<tr>
<td>On time</td>
<td>Origin – Destination flight on time</td>
<td>Origin-Hub flight on time/delayed/diverted with connection time, AND Hub-Destination flight on time</td>
<td>N/A</td>
</tr>
<tr>
<td>Delayed</td>
<td>Origin-Destination flight delayed</td>
<td>Origin-Hub flight on time/delayed/diverted within connection time AND Hub-Destination flight on time</td>
<td>Magnitude of flight delay</td>
</tr>
<tr>
<td>Passenger rebooked due to Missed Connection</td>
<td>N/A</td>
<td>Origin-Hub flight delayed/diverted and NOT within connection time, AND rebooked Hub-Destination flight on time</td>
<td>Availability of seats on later flights and frequency of flights from Hub-Destination</td>
</tr>
<tr>
<td>Passenger rebooked due</td>
<td>Origin-Destination flight cancelled AND rebooked</td>
<td>N/A</td>
<td>Availability of seats on later</td>
</tr>
<tr>
<td>to cancelled flights</td>
<td>Rebooked Origin-Destination flight on time/delayed/diverted</td>
<td>Origin-Destination flight OR Origin-Hub-Destination flight on time/delayed/diverted</td>
<td>flights and frequency of flights from Origin-Destination, Origin-Hub-Destination and Origin-Hub1-Destination</td>
</tr>
<tr>
<td>---------------------</td>
<td>----------------------------------------------------------</td>
<td>------------------------------------------------------------------</td>
<td>----------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>Diverted</td>
<td>Origin-Destination flight diverted</td>
<td>Hub-Destination flight diverted</td>
<td>Magnitude of delay due to diversion</td>
</tr>
</tbody>
</table>

Passenger itinerary data is not publicly available, for privacy and security reasons. However, a fractional sampling of itinerary data is available in public databases for analysis purposes. In practice, an airline wanting to continue experiments for its own purposes would have complete itinerary information on its own passengers.

This dissertation focuses primarily on passengers disrupted due to cancelled flights.
**Metric for Passenger On-Time Performance**

Sherry and Wang (2007) described a method for computing PTD for all hubs and airlines, but for direct flights only. The statistics were generated from the Airline On-Time Performance (AOTP) Database. Flight delays are computed based on the Computer Reservation Systems (CRS) scheduled gate arrival time.

The flight delays are computed as follows:

- **15-OTP On-Time Flights Percentage:** For each route, the number of on-time flights (with less than 15 minutes) is summed and divided by the total number of scheduled flights on that route.

- **Average Magnitude of Flight Delays:** For each route, the flight delays for delayed flights (with flight delay more than 15 minutes) are summed and divided by the total number of delayed flights on that route, yielding the Average Magnitude of Flight Delays for the route.

- **Average Worst-Case Magnitude of Flight Delays:** For each route, the top 5% of flights by flight delay are computed, and the delays are summed and divided by the total number of flights in the top 5%.

The Total PTD (TPTD) is computed through the combination of 2 algorithms:

6. **TPTD due to Delayed Flights:** compute delay time for each flight based on the AOTP data, multiplied by the average number of passengers on the flight from the T-100 data-base, in order to derive the passenger delay time for the flight. The TPTD for delayed flights is sum of the passenger delay times for all flights during the specified period.

7. **TPTD due to Cancelled Flights:** computed based on the assumption that a passenger displaced by a cancelled flight will be rebooked on a subsequent flight operated by the same carrier with the same origin/destination pair. The ability to rebook is determined by the load-factor and aircraft size of the subsequent flights.
The computed delay per passenger comes from the difference between their new arrival time and the time in their original schedule. Passengers who cannot be rebooked incur a fixed delay time.

**Modeling Passenger Travel and Delays**
Barnhart, Fearing, Vaze (2010) developed the model to analyze the causes and costs of historical U.S. passenger travel disruptions. This was a Low jet model for passenger itineraries. They developed a discrete choice model for estimating historical passenger travel, and extend a previously-developed greedy heuristic for re-accommodation and for estimating the resulting passenger delays.

In the process of passenger allocation, the first step was to join passenger and flight data from multiple sources into a large Oracle database. In the second step the data was processed in a way that would establish the inputs for the allocation (potential itineraries and seats for the passengers). The last step was to develop a discrete choice model for passenger itinerary allocation, and to train and validate the results using a small set of proprietary booking data.

The analysis is done based on the previous work of Bratu and Barnhart (2005). They analyzed a multi-day, multi-carrier rebooking process for passenger delays from 2007. In a simplified regression-based approach they estimated passenger delays directly. This linear regression model was developed to 1) identify critical characteristics of airline networks, schedules, and passenger itineraries that affect passenger delays; and 2) estimate passenger delays directly given public data, thus bypassing the process of passenger allocation and re-accommodation, shown in the following Figure:
They key findings were:

1. The ratio of average passenger delay to average flight delay is maximum for regional carriers and minimum for low-cost carriers, owing primarily to the cancellation rates and the connecting passenger percentages.

2. Passengers scheduled to transfer in one of 6 airports: Newark (EWR), Chicago O’Hare (ORD), New York La Guardia (LGA), Washington Dulles (IAD), New York Kennedy (JFK) or Philadelphia (PHL), were exposed to the longest average connecting passenger delays.

3. Domestic passenger connections are highly concentrated at the top three transfer airports: Atlanta (ATL), Chicago O’Hare (ORD), and Dallas / Fort Worth (DFW), representing approximately 43.2% of planned passenger connections.

4. Average evening passenger delay is 86.8% greater than the average morning passenger delay.
5. The average passenger delay for the three months of summer and the three months of winter was 56% higher than for the remaining six months. June was the worst month.

6. Delay to the non-stop disrupted passengers depends on the ease of rebooking and is lower for origin-destination pairs with higher daily frequency.

7. The relative benefits of flight frequency in terms of the ease of rebooking depends significantly on load factors.

8. Monday and Saturday have the lowest ratio of average passenger delay to average flight delay, and these are the only two days when the ratio is lower than the overall average value for the week.

9. Southwest Airlines has the lowest average passenger delay, nearly 55% lower than its competitors, even though its average flight delay is only 36.3% lower than other airlines.

**Passenger Itinerary Generation Algorithm**

Sherry, Samant and Calderon-Meza (2010) developed an algorithm for estimating PTD and a model for passenger itineraries. In this work, they analyzed the causes and costs of PTD for a particular year. The algorithm was used for all hubs and airlines, including direct and connecting flights.

The Passenger Itinerary Generation Algorithm was described in six steps, as follows:

Step 1: Identify all the possible itineraries. The DB1B is used to generate all the O-D and O-H-D itineraries flown during a month. These itineraries identify the sequence of airports flown by passengers on an airline (e.g., Delta Airlines: DCA-ATL, or Delta Airlines: DCA-ATL-DFW). Itineraries with more than 2 segments account for only 2.5% of the itineraries in 2007, so they are ignored. Similarly, itineraries with a very small number of passengers (less than 0.5 passengers per day) are also ignored.
Step 2: Identify the passenger itineraries. The AOTP is used to generate the individual passenger itineraries for each O-D and O-H-D itinerary in the DB1B. The output is a list of O-D and O-H-D itineraries for each day.

Step 3: Assign Seat Capacity and Total Passengers to each Flight. The T100 database is used to assign values for the number of available seats and the total number passengers on each flight.

Step 4: Estimate Total Passengers for each Day on each Itinerary. Since DB1B represents 10% of the passenger totals, then each DB1B generic itinerary has its total passengers per quarter multiplied by 10, in order to generate an estimate of the total quarterly passengers on the itinerary. The total quarterly passengers is divided by 90 days to estimate the total passengers per day on each itinerary.

Step 5: Estimate the Percentage of Passengers on each Segment of each Itinerary. Sort the itineraries that have an X-Y airport pair by airline and flight number and date. The X-Y could be in an O-D itinerary, an O-H itinerary, an O-H segment or an HD segment on an O-H-D itinerary. Divide the passenger count on individual itineraries by the sum the total passengers that share the X-Y segment. This yields the percentage of passengers on each X-Y segment of the itinerary.

Step 6: Estimate Passengers on each Flight in each Flight Itinerary. The T-100 average passengers per flight (Step 3) and the Percentage of Passengers on each segment of each Itinerary (Step 5) are used to estimate the number of passengers on each flight in each itinerary. On direct itineraries 10 passengers are added to account for a biased underestimation of passengers on DB1B direct itineraries. On connecting itineraries,
passengers are allocated evenly between the different itineraries up to the actual load factor from T100. Passenger counts are rounded up as follows: flights with less than 0.3 passengers are assigned 0 passengers, while flights with 0.3 to 0.9 passengers are rounded up.

**Passenger Trip Delay Model**

The algorithm for estimating Passenger Trip Delays uses the Passenger Itineraries described above and the Airline OnTime Performance (AOTP) data-base. The algorithm is shown in the following figure (Sherry, Samant, Calderon-Meza, 2010):
Figure 3: PIDA Algorithm (Sherry, Samant, Calderon-Meza, 2010)

The shaded (left) part shows the processing of connecting itineraries, and right part of the figure shows the processing of direct itineraries.

The Origin-Hub flight is cancelled:

Passengers are rebooked on an itinerary by the same airline from the origin to the destination. The rebooked passenger itinerary may be direct, connect through the original
hub, or connect through an alternate hub. The Passenger Trip Delay for these passengers is based on the frequency of service and availability of seats. Passengers that cannot be rebooked on the day of the flight are assigned a Passenger Trip Delay of 900 minutes (15 hours), which is an estimate of the delay accrued due to rebooking a flight the following day.

**The Origin-Hub flight is not cancelled:**

Check to see if the flight was diverted. An estimate of the delay due to diversion is computed based on roundtrip flying time and turnaround time to the closest airports with appropriate length runways. If the diverted flight returns to the Hub airport and misses the connecting flight, the passengers are rebooked from the Hub to the Destination on the same airline. The rebooked passenger itinerary is direct to the destination. The Passenger Trip Delay for these passengers is based on the frequency of service and availability of seats. Passengers that cannot be rebooked on the day of the flight are assigned a Passenger Trip Delay of 900 minutes (15 hours).

**Arrival time on the Origin-Hub flight:**

If the O-H flight is delayed and arrives less than 30 minutes from the departure of the H-D flight, the passengers are considered to have missed their connection. Passengers who miss connections are rebooked from the Hub to the Destination.

**Hub-Destination flight is cancelled.**

The passengers are rebooked from the Hub to the Destination. The Passenger Trip Delay for these passengers is based on the frequency of service and availability of seats.
Passengers that cannot be rebooked on the day of the flight are assigned a Passenger Trip Delay of 900 minutes (15 hours).

**Diversion on the H-D segment.**

If the flight is diverted, the estimate of the delay due to the diversion is computed based on roundtrip flying time and turnaround time to the closest airports with appropriate runways.

Finally, the algorithm checks the H-D flight for delays. If the flight is more than 15 minutes late, the non-zero PTD is computed.

**Direct Itinerary.**

Check for cancelled flights, diverted flights and delayed flights, rebooking and/or assigning passenger delays as described above (Sherry, Samant, Calderon-Meza, 2010).

**Modeling Passenger Trip Reliability**

Sherry (2011) introduced a probabilistic model for the operation of hub-and-spoke networks, describing passenger trip reliability metrics. The passenger trip reliability is a complex phenomenon, and it is determined not just by flight on-time performance, but by factors that have nothing to do with flight performance: load factors, frequency of service, and itinerary design.

This model shows that flight delays are one of four factors that affect passenger trip reliability. An important part are structure of the airline space-time network, fleet mix, and airline revenue and competitive strategies.
Factors which have a significant impact on passenger trip reliability metrics include:

1. Changes in aircraft size, and revenue management improvements that increase load factor (the largest impact on passenger trip reliability metrics).
2. Banking structure and frequency of service.

A method for quantifying travel productivity for corporate travel managers

Corporate Travel Managers do not plan for travel disruptions, and indirect charges due to disruptions are not budgeted and not provided in available data sets. Disruptions will thus impact the magnitude of lost billable revenue, which is not known. Sherry (2014) described a method for providing Corporate Travel Departments travel disruption statistics and their impact on revenue and profits.

The paper describes the implications for Corporate Travel Management (CTM), including the possibility to improve productivity strategies, corporate travel and indirect budgets, contracts with travel providers, and travel insurance.

Handling flight delays and cancellations

Airlines try to be as reliable as possible, but disruptions are an unavoidable phenomena. Sometimes weather, air traffic control problems and mechanical problems can result in cancellations and large delays. In many of these situations, passengers don’t need to be rebooked, but in some cases, it is necessary. In the next section, the way that Southwest Airlines and United Airlines handle passenger disruption accommodation shall be described.
Southwest Airlines
Southwest Airlines (Southwest, 2015) will not begin the boarding process if they know that a flight will be delayed at the gate for two or more hours. If the duration of the delay is known the boarding will start 30 minutes in advance of a revised estimated departure time.

If an itinerary is disrupted when the departure is from the passengers’ city of origin and Southwest is unable to transport passengers to their destinations as scheduled, passengers will be accommodated on the next flight(s) to their destinations with no charge (if seats are available).

If the passengers miss the last possible flight or connection of the day to their destinations, and if it is the case of circumstances within the airline’s control (for example due to swapping aircraft or mechanical problems), Customer Service will arrange for overnight lodging at no additional cost. In the case when it is not within the airlines control, Southwest offers to assist passengers for overnight lodging, by providing a discounted rate at a hotel or motel at or near the airport. Southwest refunds tickets on its own flights, but it does not pay for tickets from another airline, nor does it match the difference between their fares and higher fares on other airlines.

The Southwest disruption policy is called Southwest Operational Disruption Accommodation (SODA). Under SODA, if a flight does not operate as scheduled, the company would refund the unused portion of the fare, without additional charge. Another option that is provided is to arrange transportation for the passenger on another airline’s flight, assuming availability of seating.
In the case of events that may affect departures (inclement conditions, etc.) Southwest offers the opportunity to change travel dates and/or flight times at no additional charge. Even the original flight(s) may operate as scheduled, changing travel dates and/or flight times is voluntary and not required.

In those cases The Southwest Network Operations Control Team anticipates and announces, via a Travel Advisory on its site southwest.com, that the flight schedule will be impacted. Affected passengers can be re-accommodated at no charge provided that:

1. Travel is completed within 14 days of the originally scheduled itinerary and no changes are made to the originally reserved city-pairs: for example, if San Francisco International Airport is affected, passengers can change it to the San Jose or Oakland airports.
2. If the passengers chooses to travel beyond 14 days, they are responsible for any fare difference for the new reservation.
3. If changes are made outside the parameters, the changes are subject to the original fare restrictions and may result in a higher fare. Customers who have booked a Business Select Fare may change to another flight with Business Select or may switch to an Anytime Fare when Business Select is unavailable.

SODA is a rule under which passengers have to be rebooked in the same buckets: if passengers are rebooking to a flight within 14 days of the event, Southwest will overbook the bucket, even if those seats aren't available and confirm the passengers. If the airplane is sold out, they may standby at no additional charge.
United Airlines

Notification of delays and cancellations (the status of flights) is available from several sources (United, 2015):

1. On flight information screens located within airport facilities.
2. United representatives provide information in airports and onboard the aircraft in a reasonably timely manner as to known delays, cancellations and diversions.
3. Passengers may subscribe to flight status notification subscription service, which allow themselves to be notified of disruptions in advance via mobile device. Passengers provide reliable contact information, including possibly mobile device information, during reservations.

In situations where rebooking is required, United has automated systems which will attempt to confirm passengers on the next United flight that has available seats.

Using the original confirmation number, passengers can find out if and how they have been rebooked in several ways, including visiting airport kiosks, using the flight status notification subscription service, contacting United Customer Contact Center, speaking with a United airport representative, or managing reservations from the on-line site. If the next available flight is not within the next several hours (it could potentially be several days in the future), and the trip is severely delayed, passengers’ options include:

1. Standby for flights. The passenger is rebooked on a later flight, but an earlier flight with no available seats may exist. Passengers may request to stand by on the earlier flight for no charge. If seats become available, the standby passengers will be inform approximately 15 minutes before departure. If seats are not available, passengers will be placed on a standby list for the next flight, or they may request additional assistance.
2. Using another airport. Flights into or out of nearby airports may be available. Typically, ground transportation and re-route checked bags are services which are not provided.

3. Using another airline. If United delayed or canceled a flight due to reasons which are within airlines control and the delay is more than two hours, United will try to rebook passengers on another airline with which they have a rebooking agreement. When the delay or cancellation is due to reasons which are not within airlines control, United typically does not rebook customers on other airlines. At the United hub airports, Chicago (ORD), Denver (DEN), Guam (GUM), Houston (IAH), Los Angeles (LAX), New York/Newark (EWR), San Francisco (SFO), Tokyo (NRT) and Washington, D.C. (IAD), it is less likely you will be rebooked on another airline because other airlines offer fewer flights compared to United.

4. Rescheduling trip. If a flight is delayed two or more hours, and passengers want to postpone or cancel, United may waive any change penalties that may apply to your ticket.

In the case of disruptions within United’s control, United may offer food and beverages. For delays that exceed 4 hours and occur between 10 p.m. and 6 a.m., customers in a connecting city may be offered overnight lodging, depending on the amount of time involved and the location of the hotel. Hotel accommodations may not be provided if transportation is offered to or from a nearby airport.

In the case of disruptions that were not within United’s control, and depending on the duration of the delay, United may offer food and beverages. If passengers have to stay overnight, United offers distressed passenger rate vouchers for a discounted rate at nearby hotels. However, accommodations will not be refunded.
Data Source

The data used for computing passenger trip delay, including flight schedules, aircraft inventory, on-time performance of airlines and passenger demand, is derived from publicly available data sources provided by the Bureau of Transportation Statistics (BTS).

The lists of datasets described in this section are shown in the following table:

<table>
<thead>
<tr>
<th>Table 4 List of data sets</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Airline Service Quality Performance (ASQP)</strong></td>
</tr>
<tr>
<td>Schedule B-43 Aircraft Inventory (B-43)</td>
</tr>
<tr>
<td>Enhanced Traffic Management System (ETMS)</td>
</tr>
<tr>
<td>Form 41 Schedules T-100</td>
</tr>
<tr>
<td>Airline Origin &amp; Destination Survey (DB1B)</td>
</tr>
<tr>
<td>Official Airline Guide (OAG)</td>
</tr>
<tr>
<td>Airline On-time Performance (AOTP)</td>
</tr>
</tbody>
</table>

1) Airline Service Quality Performance (ASQP) (BTS, 2012a)
The Airline Service Quality Performance System (ASQP) provides information about airline on-time performance, flight delays, and cancellations. It is based on data filed by airlines each month with the Department of Transportation’s Bureau of Transportation Statistics (Office of Airline Information), as described in 14 CFR Part 234 of DOT's regulations. The ASQP system includes data for U.S. air carriers (15 as of April 2011) that have at least one percent of total domestic scheduled-service passenger revenues, and one carrier that currently reports data voluntarily. ASQP data are updated on a monthly basis, approximately 25 days after the end of the month.

2) Schedule B-43 Aircraft Inventory (2012f)

The Schedule B-43 Aircraft Inventory, published by BTS (Bureau of Transportation Statistics) provides an annual list of aircraft in inventory for most airlines including seat capacity, serial number, tail number, aircraft manufacturer, model, total capacity of the aircraft, acquisition date etc.

3) Enhanced Traffic Management System (ETMS) (2012e)

ETMS is the system used by the FAA’s Traffic Management Personnel to predict, on national and local scales, traffic surges, gaps, and volume based on current and anticipated airborne aircraft. ETMS database provides schedule information for all flights with its International Civil Aviation Organization (ICAO) code tracked by the air traffic control. This dataset is not available in the public domain.
4) Form 41 Schedules T-100 (BTS, 2012b):

The Form 41 Schedules T-100 database includes the following four datasets.

a) T-100 Data Bank 28DM Domestic Market Data: This file contains data reported by U.S. carriers operating between airports located within the boundaries of the United States and its territories, and contains information on passengers; freight and/or mail enplaned at the origin airport and deplaned at the destination airport.

b) T-100 Data Bank 28DS Domestic Segment Data: This file contains data reported by U.S. carriers operating nonstop between airports located within the boundaries of the United States and its territories and contains information by aircraft type and service class for departures performed, available capacity and seats, passengers transported, freight and mail transported, scheduled departures, and aircraft hours ramp-to-ramp and airborne.

c) T-100 Data Bank 28IM International Market Data: International Market Data are subject to access restrictions. The data fields contain information on passengers; freight and/or mail enplaned at the origin airport and deplaned at the destination airport.

d) T-100 Data Bank 28IS International Segment Data: International Market Data are subject to access restrictions. These data fields contain information by aircraft type and service class for departures performed, available capacity and seats, passengers transported, freight and mail transported, scheduled departures, and aircraft hours ramp-to-ramp and airborne.
5) Airline Origin and Destination Survey (DB1B) (BTS, 2012c)

The Airline Origin and Destination Survey (DB1B) is a 10% sample of airline tickets from reporting carriers collected by the Office of Airline Information of the Bureau of Transportation Statistics. Data includes origin, destination and other itinerary details of passengers transported. This database is used to determine air traffic patterns, air carrier market shares and passenger flows. It consists of three data sets.

a) *DB1B Coupon:* provides coupon-specific information for each domestic itinerary of the Origin and Destination Survey, such as the operating carrier, origin and destination airports, and number of passengers, fare class, coupon type, trip break indicator, and distance.

b) *DB1B Market:* provides directional market characteristics of each domestic itinerary of the Origin and Destination Survey, such as the reporting carrier, origin and destination airport, prorated market fare, number of market coupons, market miles flown, and carrier change indicators.

c) *DB1B Ticket:* contains summary characteristics of each domestic itinerary on the Origin and Destination Survey, including the reporting carrier, itinerary fare, number of passengers, originating airport, roundtrip indicator, and miles flown.

6) Official Airline Guide (OAG) (2012g)
The Official Airline Guide (OAG) database stores future and historical flight
details and schedules for more than 1,000 airlines and over 4,000 airports.

7) Airline On-Time Performance (AOTP) (BTS, 2012d)

The Airline On-Time Performance database contains on-time arrival data for non-
stop domestic flights by major air carriers, and provides such additional items as
departure and arrival delays, origin and destination airports, flight numbers, scheduled
and actual departure and arrival times, cancelled or diverted flights, taxi-out and taxi-in
times, air time and non-stop distance.
3. METHOD AND MODEL

Method of Analysis
To analyze the impact of preemptive rebooking strategies, the PTD Calculator (Sherry et al., 2011) is embedded in a Monte Carlo Simulation (Figure 13). The PTD Calculator, Monte Carlo Simulation, and the Design of Experiment are described in this section.

Figure 13 PTD Calculator, embedded in a Monte Carlo Simulation

Passenger Itinerary Delays Algorithm
The “Passenger Itinerary Delays Algorithm (PIDA)” is used to generate statistics for trip delays for each ticketed passenger itinerary (Sherry et al., 2011). The PIDA takes as an input each individual passenger itinerary. This includes both direct itineraries and connecting itineraries. The scheduled itineraries are compared against the actual performance of the flights associated the itinerary. If a flight is on-time, no passenger trip delay is accrued. If the flight is delayed more than 15 minutes, PIDA calculates the delay. In the case of cancellations or missed connections, passengers are rebooked on
flights operated by the same airline or their subsidiaries, using departure times after the scheduled departure of the cancelled or miss connected passenger.

Passengers are rebooked according to the following heuristics. First, passengers are rebooked on flights operated by the same airline or their subsidiaries with departure times after the scheduled departure of the cancelled or miss connected passenger. If the passenger cannot be rebooked on those flights (due to unavailability of seats), the passenger is then rebooked on flights operated by other airlines that depart after the scheduled departure time of the cancelled or miss connected passenger.

Passengers who are not accommodated on the same day are considered “overnight” passengers. These passengers are rebooked on the following day, first on the same airline and its subsidiaries, and then on other airlines. Passengers that cannot be accommodated in this way are considered “passengers not rebooked.”

The ability to be rebooked is dependent on the availability flights from the passenger’s location (i.e. origin airport or hub airport) to the destination, as well as the availability of seats on those flights. In this way, the Seat Size of the aircraft along with the Load Factor (the % seats occupied) determines the ability for passengers to get rebooked.

**Model of preemptive rebooking**

In the model of preemptive rebooking, passengers are divided into two big groups:

1) *Passengers on cancelled flights*, and

2) *Passengers on not cancelled flights*
Passengers on the cancelled flights are further grouped as:

1) *Passengers that CAN be rebooked*, and

2) *Passengers that CANNOT be rebooked*

If a passengers cannot be rebooked, it is either because there are no available alternative itineraries, or because the itineraries that exist lack available seats.

Passengers that can be rebooked, in traditional rebooking model, can be rebooked on the *same day later* (after the scheduled departure time), or if there is not available itinerary or seats, on the first available flight on the next day. Those that are rebooked on the following day are considered *overnight passengers*.

In the preemptive rebooking model there are four options for Passengers that CAN be rebooked. They can be rebooked:

1. *Preemptively Same Day*. The passenger is rebooked on a flight earlier in the day, before the scheduled departure time. If there are no such available seats, then the passenger is rebooked later, following the traditional rebooking model.

2. *Preemptively Previous + Same Day*. Like the preemptively same day rebooking option, the passenger is rebooked on a flight earlier that the scheduled departure time. If no alternate flights are available on the day of the original flight, then then alternate flights are sought on the previous day, before resorting to the traditional rebooking model.

3. *Same Day Later*. Describes traditional rebooking, taking place after the scheduled departure time of the original flight.

4. *Next Day*. If there is no available itinerary or seats, the passenger is rebooked on the first available flight on the next day. Such a passenger is considered an *Overnight Passenger*.
The difference between traditional and preemptive rebooking passenger groups is shown in the figure 14.

The methodology for preemptive rebooking is shown in Figure 13.

This section will explain how the rebooking algorithm takes the available itinerary information and uses it to make rebooking decisions.

The rebooking algorithm gathers all disrupted itineraries which require rebooking, and for each one, prepares a list of suitable alternative itineraries. Each itinerary includes the number of passengers using the itinerary, and each alternative has a certain number of available seats.

The algorithm selects disrupted itineraries one by one, and takes all of the passengers that need to be rebooked. The passengers in need of rebooking can be split into groups, depending on whether they are willing to accept preemptive rebooking or
not. For as long as passengers remain in need of rebooking, and for as long as alternate flights remain, remaining passengers are assigned to available seats.

First, passengers who have accepted preemptive rebooking are checked on preemptive alternative flights on the same day. Failing in that, rebooking options are checked on preemptive flights on the previous day. After that, any remaining passengers who have accepted preemptive rebooking are merged with the group of other remaining passengers, and rebooking options are checked for same-day flights later in the day. Finally, next day flights are checked. Having completed all of that, any remaining passengers are considered not rebooked, and the algorithm repeats using the next disrupted itinerary.

For example, if a flight is cancelled at 11:36 am, traditional rebooking would rebook passengers on the first available flight after original (cancelled) scheduled departure time (figure 15). With preemptive rebooking they will be rebooked on the first available flight before original (cancelled) scheduled departure time of 11:36, for instance at 10am. If there are no itineraries or seats available between 11:36 am and 6 am, passengers will be rebooked on the previous day (starting from 11pm to 6 am). If there are no itineraries or available seats on the previous day, they will be rebooked first available flight after 11:36 am, the original scheduled departure time.
Preemptive rebooking algorithm
The methodology for preemptive rebooking is shown in Figure 16.
REBOOKING METHODOLOGY

Import all flights with disruptions (with pax) and all available alternative flights (with seats)

Is there still disrupted flights which are not been processed

Take the next unprocessed disrupted flight

Take the number of pax on the flight, and based on preemptive percent, determine how many of the pax are previous day, same day preemptive, or regular (same day later, next day)

Number of previous day + same day preemptive pax > 0

Assume all remaining preemptive pax are previous day pax, and place the overflow in the regular remaining pax group

Add the number of remaining preemptive pax (same and previous day) to the number of regular remaining pax

Number of pax > 0

Are there still unprocessed same day early alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Are there still unprocessed same day late alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Are there still unprocessed next day alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Number of previous day pax = 0

Extract remaining previous day pax

Are there still unprocessed previous day alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Number of pax = 0

Are there still unprocessed same day early alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Number of pax > 0

Are there still unprocessed same day late alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Are there still unprocessed next day alternatives for the current disrupted flight

Take the next unprocessed alternative flight

Check if there are unfilled seats on the alternative flight

Put as many pax as possible on the alternative flight, and deduct from remaining preemptive

Figure 16: Preemptive Rebooking Methodology
This section will show how the rebooking algorithm takes the available itinerary information and uses it to make rebooking decisions. Passengers who cannot be rebooked on the same day, will be rebooked on the next day. Remaining passengers who cannot be rebooked on the next day will be considered as passengers not rebooked. Rebooking depends on:

- The availability flights from the passengers’ location (i.e. origin airport or hub airport) to the destination.
- The availability of seats on those flights.

Due to the fact that seat availability is a factor, the Seat Size of the aircraft and the Load Factor (the % of seats occupied) are determining factors in the ability for passengers to be rebooked.

The algorithm selects disrupted itineraries one by one, and takes all of the passengers that need to be rebooked. The passengers needing to be rebooked can be split into groups depending on whether they are willing to accept preemptive rebooking or not. For as long as passengers remain in need of rebooking, and for as long as alternate flights remain, remaining passengers are placed in empty seats.

First the passengers who accept preemptive rebooking are checked on preemptive alternative flights on the same day. Failing in that, rebooking options are checked on preemptive flights on the previous day. After that, any remaining passengers who have accepted preemptive rebooking are merged with the other remaining passengers, and rebooking options are checked for same-day flights later in the day. Finally, next day flights are checked. Having completed all of that, any remaining passengers are considered unbooked, and the algorithm repeats with the next disrupted itinerary.
Code Optimization and Method Contribution

The code used by the Monte Carlo simulator and PTD calculator (Figure 16 above pp 57) was derived from existing code for the PTD calculator without preemptive rebooking.

Previously existing code consists of five steps written in Microsoft SQL which needed to be run consecutively. The code has limited flexibility in terms of parameters such as the dates, number of days, and companies used in the simulation. Code written for this research simplified the existing code into a single step, supported by several stored procedures with configurable parameters. Several capabilities were added to the code, as described below.

Code Contribution

Many of the sections of code which involved creating, populating, and updating tables were moved into stored procedures. This made it easier to modify parameters and call repeated sections of code without redundancy. The remaining code was greatly simplified by observing that repeated sections of code could be combined to reduce redundancy. This especially impacted the DB1B to T100 conversion, as well as the rebooking algorithm itself. The difference is shown in the figure 17.

The code contribution is summarized in the next table:
<table>
<thead>
<tr>
<th>Table 5 Preemptive Rebooking: code contribution</th>
</tr>
</thead>
<tbody>
<tr>
<td>Preemptive rebooking</td>
</tr>
<tr>
<td><strong>Load factor bias</strong></td>
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<tr>
<td><strong>Passengers groups:</strong></td>
</tr>
<tr>
<td>Passengers who can be rebooked</td>
</tr>
<tr>
<td>Previous Day,</td>
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<tr>
<td>Same Day early,</td>
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<tr>
<td>Same Day After,</td>
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<tr>
<td>Next Day (Overnight: Stranded passengers)</td>
</tr>
<tr>
<td>Passengers who cannot be rebooked:</td>
</tr>
<tr>
<td>No Itineraries</td>
</tr>
<tr>
<td><strong>Trip Time (Scheduled, Alternative, Actual)</strong></td>
</tr>
<tr>
<td>Automated Statistics:</td>
</tr>
<tr>
<td>Airfares Refund/Not Refund, /</td>
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<tr>
<td>Corporate Travel Expense Saving,</td>
</tr>
<tr>
<td>Corporate Travel Expense/ Corporate</td>
</tr>
<tr>
<td>Travel Expense Saving,</td>
</tr>
<tr>
<td>Passenger statistics</td>
</tr>
<tr>
<td>Run statistics</td>
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<tr>
<td><strong>Efficiency of preemptive scheduling</strong></td>
</tr>
</tbody>
</table>
**Debugging**

A significant error in the original code was located and corrected. The cursor which processed all alternative options was a static cursor, meaning that all updates to the underlying tables would not be reflected in the data retrieved by the cursor. As a result, changes made by the rebooking algorithm would be ignored as the rebooking algorithm continued to operate. This meant that, for example, a flight that was filled by the rebooking algorithm could be refilled again later. The error was fixed by maintaining the needed data in local variables whenever possible (for example, the number of remaining passengers on a flight is now stored in a local variable after a new flight is read), and dynamic cursors are used when the underlying data can change.

The new code uses a Monte Carlo simulation which did not exist in the original code. In order to implement the simulator, the relevant parts of the code (e.g. step three of the original code) were placed in a loop that iterates the desired number of times.

The ability to process preemptive rebooking was added to the code. This involved adding a loop which would re-run the simulation for each choice of preemptive percent. The original code would consider alternate flights in order. The new code does the same, but early flights are added to the group of flights to be considered. That means that alternatives are considered in the order beginning with flights departing prior to, but on the same day as, the original scheduled departure of the flight, followed by flights departing on the previous day, followed by flights departing after, but on the same day, followed by flights departing on the next day. Additionally, in each group, alternate flights on different airlines were considered after all alternates on the same airline were
exhausted. A number representing priority category was introduced into the code, and used to establish the sorting order.

To make it easier to keep track of the number of passengers which could not be rebooked, two virtual priority categories were added, accounting for passengers which could not be rebooked, either due to the lack of available alternative flight options (no itineraries), or the lack of available seats. The new code is more rigorous about counting flights without alternative itineraries, due to the fact that those flights are now included in the alternative table under a priority level which is unused by the rebooking algorithm.

Due to the fact that some passengers rebook preemptively and some do not, a binomial distribution calculator was added to the code, and used to randomly determine the number of passengers which would accept preemptive rebooking. Whenever a new flight is processed, the distribution calculator is be used together with the preemptive acceptance rate, to determine the number of passengers for previous day preemptive rebooking and same day preemptive rebooking. The preemptive alternatives would only be processed for the relevant preemptive passengers, whereas the non-preemptive alternative options would be processed for all remaining passengers (see the model in the previous section).

The ability to adjust a load factor bias was also added to the code. This feature was implemented by adding a parameter which would adjust the number of available seats at the time that the list of alternative flight options was being created.

Another added capability was the ability to compute trip time, including actual, scheduled, and alternative trip time. The trip time capability is supported by adding per
passenger multipliers to the alternative options table. The trip time in each category can be determined by multiplying the number of passengers on each alternative option by the multiplier for that option.

The new code works together with a new statistics script, which automates the computation of results such as revenue recouped, indirect costs, and the efficiency of preemptive scheduling. To enable a detailed statistics computations, the code now stores a table of results which includes information broken down by Monte Carlo run, alternative priority, date, and type of disruption. The statistics script uses the information from different runs to compute means, medians, and standard deviations of the desired values.

**Code sequence diagram**
Figure 17 Code Sequence Diagram

Code statistics:

- 26 stored procedures (Table 6):

<table>
<thead>
<tr>
<th>Stored procedures</th>
<th>Lines</th>
<th>Stored procedures</th>
<th>Lines</th>
<th>Stored procedures</th>
<th>Lines</th>
</tr>
</thead>
</table>

Table 6 Stored Procedures
### Stranded/Remaining Passengers Calculation

Stranded passengers are the passengers who are rebooked on a subsequent day, meaning that they must stay overnight to meet their flight schedule. Remaining passengers are passengers who do not have an alternative option for their itinerary and must remain unbooked.

The number of stranded passengers is the number of passengers who have been rebooked on a next-day alternative itinerary. In the code, this number can be determined by the number of passengers in the alternative priority categories 7 and 8 in the results.
table. The number of remaining passengers is the number of passengers who have not been rebooked. This is stored in the results table as the number of passengers in alternative priority categories -1 (no itineraries) and 0 (no seats), which are determined from the number of remaining passengers on flights after the rebooking algorithm has completed.

Formulas and Definitions

\( i \): Itinerary, assuming that there are \( n \) itineraries, numbered 1 through \( n \).

\( Sch \): Scheduled, referring to the departure and arrival times scheduled for the flight.

\( Act \): Actual, referring to the actual departure and arrival times of a flight.

\( Reb \): Rebooked, referring to rebooked flights. A rebooked flight will also have scheduled and actual departure and arrival times, so for example, the scheduled departure time of a rebooked flight would be written as \( Sch\text{Departure}_{Reb} \).

\( PTD \): Passenger Trip Delay, a measure of the delay between each passenger’s scheduled arrival time and their actual arrival time (assuming that a window of up to 15 minutes is not considered a delay). If a passenger arrives before the ticketed time (either due to preemptive rebooking, or simply a fast flight), the passenger’s contribution to the PTD is zero.

\[
PTD = \sum_{i=1}^{n} \max(0, Act\text{Arrival}(i) - Sch\text{Arrival}(i) - 15\text{min})
\]

Reduction in airline lost revenue, AIRFARE NOT REFUND (ANF)

Airfare not refund: \( ANR \) is the difference in airfare refunded by the airline without preemptive rebooking and with preemptive rebooking. The assumption is that for
passengers not rebooked, airlines have to refund an average airfare of $377 per passenger.

Airfare refund is an amount which airline pays to passengers not rebooked. The baseline is the amount of refund expected without preemptive rebooking. The difference between baseline and the amount with preemptive rebooking is the airline’s savings (ANR).

\[ AR_{\text{base}} = (PAX_{\text{RebNDBase}} + PAX_{\text{RemBase}}) \cdot $377 \]

\[ AR = (PAX_{\text{RebND}} + PAX_{\text{Rem}}) \cdot $377 \]

\[ ANR = AR_{\text{base}} - AR \]

\[ \%ANR = 100\% \cdot \frac{ANR}{AR_{\text{base}}} \]

Where:

- \( PAX_{\text{RebNDBase}} \): Number of passengers rebooked Next Day, with no preemptive rebooking.
- \( PAX_{\text{RebND}} \): Number of passengers rebooked Next Day, using the chosen level of preemptive rebooking.
- \( PAX_{\text{RemBase}} \): Number of unbooked passengers remaining, with no preemptive rebooking.
- \( PAX_{\text{Rem}} \): Number of unbooked passengers remaining, using the chosen level of preemptive rebooking.
- \( AR_{\text{base}} \): Airfare refund for the baseline case without preemptive rebooking
CTES is the difference in additional travel expenses accrued by corporate travelers without preemptive rebooking and with each of the preemptive rebooking percentages that are required to overnight due to rebooking the next day. The assumption is that corporates have to pay additional costs for overnight business travelers, which is $250 on average, $160 for hotel accommodation and $90 for food and transportation expenses (Sherry, 2014). It is assumed that 50% of the passengers are not at their home town airport and would require overnight hotel accommodation (Li, Baik, Trani, 2010). Further 50% of these passengers are estimated to travel on corporate expense accounts (DoT, 2015).

The baseline is the cost which corporates have to pay in the case without preemptive rebooking. The difference between the baseline and amount with preemptive rebooking is the corporates’ savings.

\[
CTE_{\text{Base}} = \$250 \times (PAX_{\text{RebND Base}} + PAX_{\text{Rem Base}} + PAX_{\text{RebPDBase}}) \times 0.5 \times 0.5
\]

\[
CTE = \$250 \times (PAX_{\text{RebND}} + PAX_{\text{Rem}} + PAX_{\text{RebPD}}) \times 0.5 \times 0.5
\]

\[
CTES = CTE_{\text{Base}} - CTE
\]

\[
\%CTES = 100\% \times CTES/CTE_{\text{Base}}
\]

\[PAX_{\text{RebPDBase}}\]: Number of passengers rebooked Previous Day, with no preemptive rebooking.

\[PAX_{\text{RebPD}}\]: Number of passengers rebooked Previous Day, using the chosen level of preemptive rebooking.

**PASSENGERS**

Percentage of preemptively rebooked passengers is calculating like this:
\[ \% \text{ preemptively rebooked pax} = 100\% \cdot \frac{\text{Pax}_{\text{reb}}}{\text{Pax}_{\text{can}}} \]

\( \text{Pax}_{\text{reb}} \): Number of pax rebooked preemptively.

\( \text{Pax}_{\text{can}} \): Number of pax on the cancelled flights.

**Percent of passengers accommodated**

Let us suppose that 10\% of passengers accept preemptive rebooking. Then how many of those can be accommodated on an early flight? Number of passengers accommodated is defined as number of passengers willing to be rebooked earlier (preemptively rebooked), who are rebooked on an earlier flight. Not all passengers who are willing to be preemptively rebooked can be rebooked on earlier flights, because itineraries or seats might not be available. The algorithm first checks the available itineraries and available seats. Seats are number of passengers who can be rebooked earlier. Then the algorithm fills the seats until there are empty seats.

*Overnight passengers*, also called *stranded passengers*, are passengers who are rebooked on the day after a flight’s scheduled departure time.

*Passengers that cannot be rebooked* are remaining passengers who cannot be rebooked, as there are no options for their rebooking.

*TTT*: Total Trip Time, a measure of the combined time that each passenger is scheduled to travel, from initial departure to final arrival.

\[ TTT_{\text{Sch}} = \sum_{i=1}^{n} [\text{SchArrival}(i) - \text{SchDeparture}(i)] \]
\[ TTT_{Reb} = \sum_{i=1}^{n} [ActArrival_{Reb}(i) - SchDeparture_{Reb}(i)] \]

\[ \Delta TTT : \text{Change in Total Trip Time, from the original schedule to the final actual flight time including rebooking.} \]

\[ \Delta TTT = TTT_{Reb} - TTT_{Sch} \]

As an example, a passenger may have a 2 hour flight scheduled at noon, so their scheduled trip time would be 2 hours. Suppose that their flight is cancelled, and they are preemptively rebooked on a 2 hour flight that leaves at 8 in the morning, but the flight is delayed for an hour. Their rebooked trip time would now be 3 hours, due to the delay, and they would arrive at 11am, which is still earlier than the original scheduled arrival. That means that the trip time has gone up while the PTD is zero.

**Monte Carlo Simulation**

To achieve the objectives of the analysis, various parameters that are inputs to the PTDC are modified over multiple runs in the Design of Experiment (see below). Parameters that are modified include:

- Percentage of passengers that choose to accept a preemptive rebooking option. The actual passengers that preemptively rebook is taken by randomly selecting passengers at a uniform rate from all passengers on all itineraries, using the preemptive percentage as the probability for each passenger.
- The time in advance that the preemptive rebooking option is made available to passengers. Passengers may choose Same day before or Previous day + Same day preemptive rebooking, and the algorithm will accepts these two options.
• Load Factor Bias. Levels of Load Factor which differ from the baseline Load Factor may be used.

The Monte Carlo Simulation is executed a fixed number \( n \) times for each replication. The epoch size \( n \) is set as a configurable parameter in the code. The results are stored and then used to generate statistics by a processing algorithm.

In order to determine the minimum but a sufficient number of runs \( n \), a confidence interval approach was used. For a confidence interval of 95%, it means that if the suggested number of runs \( n \) is used, there is less than 5% chance that a bad estimate will occur through pure chance. A confidence interval is calculated by finding the area contained in the tail ends of a normal distribution.

The value of \( n \) for a given confidence interval is given as
\[
\begin{align*}
n &= \left(\frac{z \cdot \sigma}{E}\right)^2,
\end{align*}
\]
where \( z \) is the critical value for the chosen confidence interval (for 95% confidence, the parameter \( z = 1.96 \)), \( \sigma \) is the standard deviation of the data, and \( E \) is the interval of tolerance for the data. As an example, if the data values are 5 digits, and two significant figures are desired, then a tolerance of \( E = 0.5 \times 10^{(5-2)} = 500 \) is needed.

For rebooked passengers (the calculations is done for number of passengers rebooked preemptively –same day, preemptively same + previous day, same day later and next day), the data values and the range of \( \sigma \) is shown in table 7.

<table>
<thead>
<tr>
<th>Same Day Preemptively</th>
<th>Same + Previous Day Preemptively</th>
</tr>
</thead>
<tbody>
<tr>
<td>10% Pax Rebooked</td>
<td></td>
</tr>
<tr>
<td>( \mu )</td>
<td>( \sigma )</td>
</tr>
<tr>
<td></td>
<td>( \mu )</td>
</tr>
<tr>
<td></td>
<td>Same day Early</td>
</tr>
<tr>
<td>------------------------</td>
<td>----------------</td>
</tr>
<tr>
<td><strong>Same day Early</strong></td>
<td>380.90</td>
</tr>
<tr>
<td><strong>Same Day Late</strong></td>
<td>2674.50</td>
</tr>
<tr>
<td><strong>Next Day</strong></td>
<td>2139.60</td>
</tr>
<tr>
<td><strong>30% pax Rebooked</strong></td>
<td>μ</td>
</tr>
<tr>
<td>Same day Early</td>
<td>1144.08</td>
</tr>
<tr>
<td>Same Day Late</td>
<td>2197.64</td>
</tr>
<tr>
<td>Next Day</td>
<td>1853.28</td>
</tr>
<tr>
<td><strong>50% pax Rebooked</strong></td>
<td>μ</td>
</tr>
<tr>
<td>Same day Early</td>
<td>1910.80</td>
</tr>
<tr>
<td>Same Day Late</td>
<td>1695.80</td>
</tr>
<tr>
<td>Next Day</td>
<td>1588.30</td>
</tr>
<tr>
<td><strong>70% pax Rebooked</strong></td>
<td>μ</td>
</tr>
<tr>
<td>Same day Early</td>
<td>2678.60</td>
</tr>
<tr>
<td>Same Day Late</td>
<td>1181.20</td>
</tr>
<tr>
<td>Next Day</td>
<td>1335.10</td>
</tr>
</tbody>
</table>

In the case $\sigma=116$, for 2 significant digit accuracy, $E = 50$, resulting in a need for at least $n>20.7$ runs, which confirms that 25 runs are sufficient to achieve the desired confidence. Every additional significant digit of accuracy of the results will divide the value of $E$ by 10, meaning that it would require a 100 times increase in the number of runs.

**Design of Experiment**
The Design of Experiment is summarized in Table 1 and 2. Two preemptive rebooking options are considered: Same day earlier (ticketed scheduled departure time), and Previous day + Same day earlier. The percentage of passengers rebooking is incremented beginning from none (i.e. 0%), 10%, 30%, 50%, 70%. The 0% is the baseline where no preemptive rebooking is used. The template for collecting results as a function of this independent parameter is shown in the following table.

<table>
<thead>
<tr>
<th>RESULTS</th>
</tr>
</thead>
<tbody>
<tr>
<td>COONTROL PARAMETERS</td>
</tr>
<tr>
<td>Same Day Before</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Previous day plus Same Day Before</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
</tbody>
</table>

Table 8 Design of Experiment
4. RESULTS

Multiple case-studies were run for United Airlines and Southwest Airlines scheduled flights in 2012 (table 9). The results are discussed in the next section.

<table>
<thead>
<tr>
<th>Table 9 Case studies</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>United Airlines (UA)</strong></td>
</tr>
<tr>
<td>Case study 1 – UA, ORD one day (Jan 12 2012)</td>
</tr>
<tr>
<td>Case study 2 – UA, ORD 2012</td>
</tr>
<tr>
<td>Case study 3 – UA, Hubs 2012</td>
</tr>
<tr>
<td><strong>Southwest Airlines (SW)</strong></td>
</tr>
<tr>
<td>Case study 4 – SW, MDW, 2012</td>
</tr>
<tr>
<td>Case study 5 – SW, Hubs, 2012</td>
</tr>
</tbody>
</table>

For all case studies, the $H_0$ hypotheses are rejected (Table 10). The p-value represents the probability that a null hypothesis cannot be rejected given the test results.

As shown in the table, the probability represented by the p-value is small everywhere, with a value less than 0.1%, meaning that there is high statistical significance to the claim that the null hypothesis were all rejected.
Table 10 p-values for case studies:

<table>
<thead>
<tr>
<th>Case Studies</th>
<th>H₀ 1: PR does not accommodate more than 10% pax on cnx flights</th>
<th>H₀ 2: Preemptive rebooking does not result in additional revenue to airlines</th>
<th>H₀ 3: Preemptive rebooking does not result in additional saving to Corporation s</th>
<th>H₀ 4: PR does not result in reducing PTD and number of overnight and passenger which cannot be rebooked</th>
<th>H₀ 5: Number of additional pax accommodated by adding an additional day of preemptive rebooking is greater than 10%</th>
</tr>
</thead>
<tbody>
<tr>
<td>UA, Jan 12</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
</tr>
<tr>
<td>UA, ORD</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
</tr>
<tr>
<td>UA</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
</tr>
<tr>
<td>SW, MDW</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
</tr>
<tr>
<td>SW</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
<td>p&lt;0.001</td>
</tr>
</tbody>
</table>

**United Airlines**

In the first three case studies the observed airline was United Airlines. During 2012 the airline had a total of 531,245 flights, out of which, 7599 were cancelled (1.43%). The cancellations per day are shown in the next histogram (Figure 18).
The biggest number of days had less than cancellations (188 days), and more than 100 cancellations per day had 7 days. The biggest number of cancelled flight in one day was 720. Mostly all cancellation events were one-day events. In this thesis just those one day events, when number of cancelled flights is bigger then 20, will be analyzing. First case study is a one day event, January 12\textsuperscript{th} 2012.

**Case Study 1**
A case-study was conducted for the scheduled domestic flights for a United Airlines operating from ORD hub for an event on January 12, 2012. Snow accumulation of 4.7” started around Noon and lasted through the evening ((NOAA, 2015). There were a total of 56 cancelled flights to or from the hub airport (Figure 19) impacting an estimated 5250 passengers (Table 11).
Figure 19 UA, ORD, January 12th 2012, Cancellations during the day

### Table 11 Flight Statistic for the event January 12th, 2012

<table>
<thead>
<tr>
<th>%Flights Cancelled</th>
<th>Cancelled Flights</th>
<th>Passengers</th>
<th>Passengers on the Cancelled Flights</th>
</tr>
</thead>
<tbody>
<tr>
<td>16.4%</td>
<td>56</td>
<td>29,052</td>
<td>5,250</td>
</tr>
</tbody>
</table>

The average load factor was 80% with a minimum of 32% and maximum of 96% (Table 12).

### Table 12 Load Factor statistics

<table>
<thead>
<tr>
<th>LF percent</th>
<th>Number of cancelled flights (with Origin/Destination in the hub)</th>
<th>Number of flights</th>
<th>Number of flights with Origin/Destination in the hub</th>
</tr>
</thead>
</table>
For the existing re-booking paradigm, assuming all passengers are to be rebooked, 55% of the cancelled passengers are accommodated with seats on the same day after the departure time of the cancelled flight. Forty-three percent must be rebooked on flights the following day. Approximately 1% of the cancelled passengers cannot be rebooked due to the absence of available seats to their desired destination (Figure 20). The upper bound for refunded airfares is $833K. The upper bound for unplanned (i.e. unbudgeted) Corporate Travel Expenses is $292.8K.
Case study 1: \( H_0: \) Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%)

Based on results it is shown that Preemptive Rebooking accommodated 73% of passengers, hence Null hypothesis, \( H_0: \) Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%) is rejected.

Monte Carlo simulation of randomly selected passengers seeking preemptive rebooking for cancelled flights were accommodated 72% of the time (\( \sigma=6\% \)). The relationship between the percent of Preemptively Rebooked Passengers and the percent of Passengers seeking Preemptive Rebooking is as follows (\( R^2=0.996 \)):

\[
\% \text{ Pax Preemptively Rebooked} = 0.72 \times \% \text{ Pax Seeking Rebooking}
\]

For example, for this case study, when 10% of the passengers pursued preemptive Same Day Early Rebooking, 7.2% were accommodated. When 70% of the passengers pursued preemptive Same Day Early Rebooking 51% were accommodated (Figure 21). The yellow line shows full accommodation projection, the projected value if all passengers willing to be preemptively rebooked are accommodated.
For the passengers that sought preemptive rebooking but could not be accommodated, it was due to the absence of seats on flights, not the absence of itineraries.

Case study 1: H₀: The number of additional pax accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated

Based on results it is shown that a majority of the passengers (>93%) can be accommodated earlier on the same day. Hypothesis H₀: The number of additional pax accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated is rejected.

Randomly selected passengers seeking preemptive rebooking for cancelled flights were accommodated 79% of the time (σ=6%). The relationship between the percent of
Preemptively Rebooked Passengers and the percent of Passengers seeking Preemptive Rebooking is as follows (R2=0.996):

\[ \% \text{ Pax Preemptively Rebooked} = 0.79 \times \% \text{ Pax Seeking Rebooking} \]

For example, for this case study, when 10\% of the passengers pursued preemptive Same Day Early Rebooking, 7.9\% were accommodated. When 70\% of the passengers pursued preemptive Same Day Early Rebooking 55.4\% were accommodated.

For the passengers that sought preemptive rebooking but could not be accommodated, it was due to the absence of seats on flights, not the absence of itineraries.

These results were a 4\% increase over the Preemptive Rebooking for the Same Day only. The benefits of Previous Day rebooking are marginal (Figure 22). The yellow line shows full accommodation projection, the projected value if all passengers willing to be preemptively rebooked are accommodated.
Case Study 1: H0: Preemptive rebooking does not result in additional revenue to airlines: Rejected

Based on the Monte Carlo results it is shown that Preemptive rebooking results in additional revenue to airlines up to $359K for same day and $444K for same + previous day. Therefore, the hypothesis H0: Preemptive rebooking does not result in additional revenue to airlines is rejected.

The upper bound (i.e. case when all passengers who stay overnight and the passengers that cannot be rebooked ask for refunds) of ANR when 7.3% of the passengers are pre-emptively rebooked is $56K, and the upper bound of ANR when 51% of the passengers are pre-emptively rebooked is $359K for the same day, it is $67K for 10% and $444K for 70% for same+previous day (figure 22, shows percent of ANR with decrease of percentage of passengers willing to accept preemptive rebooking).
The total savings in ANR (upper bound) for same day preemptive rebooking and same + previous day preemptive rebooking is shown in the following table:

<table>
<thead>
<tr>
<th>% of Pax accepted preemptive rebooking</th>
<th>Same Day Early [§]</th>
<th>Same+Previous Day [§]</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>10</td>
<td>56K</td>
<td>67K</td>
</tr>
<tr>
<td>30</td>
<td>164K</td>
<td>201K</td>
</tr>
<tr>
<td>50</td>
<td>264K</td>
<td>326K</td>
</tr>
<tr>
<td>70</td>
<td>360K</td>
<td>444K</td>
</tr>
</tbody>
</table>

Figure 23 Airfares Not Refund, same and same + previous day
Case Study 1 H0: Preemptive rebooking does not result in additional saving to Corporates: Rejected

Based on the simulation results it is shown that Preemptive rebooking results in additional revenue to corporates up to K for same day and K for same + previous day. Hence, the hypothesis H0: Preemptive rebooking does not result in additional revenue to corporates is rejected.

Pre-emptive rebooking also has an effect of reducing the number of overnight passengers that could accrue unbudgeted Corporate Travel Expenses (Figure 24). The upper bound for Corporate Travel Expense Savings (CNES) is represented by the following relationship ($R^2 = 0.9987$):

\[ CNES (\$) = 1.2K \times \% \text{ Pax Preemptively Rebooked} \]

![Figure 24 CTES](image-url)
For example, the upper bound of ANR when 7.2% of the passengers are pre-emptively rebooked is $9K. The upper bound of ANR when 51% of the passengers are pre-emptively rebooked is $44K (Table 14).

<table>
<thead>
<tr>
<th>% of pax willing to be preemtively rebooked</th>
<th>Same day</th>
<th>Previous + Same day</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>10</td>
<td>9K</td>
<td>9K</td>
</tr>
<tr>
<td>30</td>
<td>27K</td>
<td>27K</td>
</tr>
<tr>
<td>50</td>
<td>44K</td>
<td>43K</td>
</tr>
<tr>
<td>70</td>
<td>59K</td>
<td>59K</td>
</tr>
</tbody>
</table>

**Case Study 1: H₀: Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked: Rejected**

Based on the simulation results it is shown that Preemptive rebooking results in reducing PTD, the number of overnight stays, and passenger which cannot be rebooked. Hence, the hypothesis H₀: Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked is rejected.

Preemptive rebooking accepted by random passengers from the cancelled flights has the effect of freeing-up seats on flights after the cancelled flight. This allows passengers that would otherwise be rebooked the next day to be rebooked on the same day. The percentage in reduction in the number of passengers rebooked on the next day as a function of the % Passengers Rebooked Preemptively is represented by the equation below ($R² = 0.9987$).
\[
\text{\% Reduction in Pax Preemptively Rebooked on the Next Day} = \\
0.35 \cdot \% \text{ Pax Preemptively Rebooked} + 0.12
\]

For example, for this case study, the preemptive rebooking of 7.2\% of the passengers reduced the percentage of passengers rebooked overnight by 14\%. The preemptive rebooking of 51\% of the passengers, reduced the percentage of passengers rebooked overnight by 29.9\% (figure 25).

![Figure 25 Decrease in Overnight Pax](image)

Average PTD decreased from 499 (base line) to 274min (70\%), and total PTD from 1818 days (base line) to 1000 days (70\%), and number of passengers who cannot be rebooked is decreased by 42.5\%. The decrease in PTD is shown in the figure 26.
Figure 26 Decrease in Total PTD

Total trip time was not changed significantly (Figure 27).
Case Study 1: UA, ORD, Jan 12, Summary of results

Case study 1 for United Airlines, hub ORD, January 12 shows that Pre-emptive rebooking in advance of forecast large-scale cancellation events is feasible. At least 72% of the passengers seeking pre-emptive rebooking can be accommodated before their original scheduled flight time. The remaining 28% of the passengers cannot be re-accommodated due to insufficient seats. Pre-emptive rebooking on the previous day was not required. Rebooking on the previous day did not significantly change the percentage of passengers re-accommodated. Pre-emptive rebooking also creates a win-win for all the stakeholders. Airlines are able to recoup up to $7.7K \cdot \%\text{Passengers Rebooked} Preemptively that otherwise might be refunded. Corporations sponsoring business travel for their employees also benefit by saving up to $1.2K \cdot \%\text{Passengers Rebooked} Preemptively through unbudgeted overnight costs.
Case Study 2
A case-study was conducted for the scheduled domestic flights for a United Airlines operating from ORD hub for all one day cancellation events in 2012. There were a total of 20 days with more than 20 cancelled flights in that year, out of which 13 days were 1-day cancellation events (Table 15).

<table>
<thead>
<tr>
<th>Month</th>
<th>Day</th>
<th>Number of Cancelled Flights</th>
<th>Arrival Flights Cancelled</th>
<th>Departure Flights Cancelled</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>12</td>
<td>56</td>
<td>31</td>
<td>26</td>
</tr>
<tr>
<td>1</td>
<td>20</td>
<td>77</td>
<td>33</td>
<td>44</td>
</tr>
<tr>
<td>1</td>
<td>23</td>
<td>38</td>
<td>23</td>
<td>15</td>
</tr>
<tr>
<td>2</td>
<td>23</td>
<td>28</td>
<td>11</td>
<td>17</td>
</tr>
<tr>
<td>5</td>
<td>6</td>
<td>26</td>
<td>8</td>
<td>18</td>
</tr>
<tr>
<td>5</td>
<td>29</td>
<td>30</td>
<td>15</td>
<td>15</td>
</tr>
<tr>
<td>6</td>
<td>22</td>
<td>28</td>
<td>12</td>
<td>16</td>
</tr>
<tr>
<td>6</td>
<td>29</td>
<td>22</td>
<td>10</td>
<td>12</td>
</tr>
<tr>
<td>7</td>
<td>13</td>
<td>34</td>
<td>16</td>
<td>18</td>
</tr>
<tr>
<td>7</td>
<td>15</td>
<td>25</td>
<td>12</td>
<td>13</td>
</tr>
<tr>
<td>7</td>
<td>18</td>
<td>37</td>
<td>14</td>
<td>23</td>
</tr>
</tbody>
</table>
In 2012 at the mid-west hub the airline experienced 4 days (2.5%) with between 20 and 30 flight cancellations. There were 5 days (1.4% of the year) with 30 and 40 cancelled flight, 4 days (1.1% of the year) with between 40 and 60 cancelled flights, and 2 days (0.54% of the year) with more than 100 cancelled flights.

For the day of the one day cancellation events, the average load factor on all flights was 83% with a minimum of 25% and maximum of 100%. The average load factor on the cancelled flights was 80% with a minimum of 32% and maximum of 97%.

Figure 28 shows number of days with one day events, with more than 20 cancellations per day.
The load factor statistics are shown in Table 16.

Table 16: Load Factor Statistics, 2012

<table>
<thead>
<tr>
<th>LF percent</th>
<th>% of cancelled flights (with Origin/Destination in the hub)</th>
<th>% of flights</th>
<th>% of flights with Origin/Destination in the hub</th>
</tr>
</thead>
<tbody>
<tr>
<td>0-50%</td>
<td>0.6</td>
<td>0.61</td>
<td>1.56</td>
</tr>
<tr>
<td>50%-60%</td>
<td>1.16</td>
<td>1.83</td>
<td>2.2</td>
</tr>
</tbody>
</table>
Baseline Rebooking (After Cancelled Flight)

For the existing re-booking paradigm, no passengers will be booked on preemptive flights, so the percent of passengers rebooked early will automatically be zero. The result is a slope of zero in the equation calculating the percent of passengers rebooked early. For similar reasons, the ANR and CTES slopes will also be zero.

<table>
<thead>
<tr>
<th>Percent Range</th>
<th>Slope (Baseline)</th>
<th>Slope (Preemptive)</th>
<th>Slope (Cancelled)</th>
</tr>
</thead>
<tbody>
<tr>
<td>60%-70%</td>
<td>6.5</td>
<td>4.9</td>
<td>8.78</td>
</tr>
<tr>
<td>70%-80%</td>
<td>24.3</td>
<td>21.21</td>
<td>28.89</td>
</tr>
<tr>
<td>80%-90%</td>
<td>44.98</td>
<td>49.01</td>
<td>42.65</td>
</tr>
<tr>
<td>&gt;90%</td>
<td>22.46</td>
<td>22.44</td>
<td>15.93</td>
</tr>
</tbody>
</table>

Case study 2: \( H_0: \) Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%)

Based on the results that a projected 70% for upper bound of the passengers willing to be preemptively rebooked, will be accommodated it is shown that a hypothesis setting that Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%) is rejected.

The Monte Carlo simulation shows that randomly selected passengers seeking pre-emptive rebooking for cancelled flights are accommodated at a rate of 70% of the number of passengers accepting early rebooking (the slope is 70%). If no passengers accept early rebooking, no passengers are accommodated preemptively, while if 70% of
the passengers accept early rebooking, then 70% • 70% = 49% of the total cancelled passenger are accommodated. If all passengers were to accept preemptive rebooking, a projected 70% would be accommodated.

The equation below represents the %Pax Accomodated, where the %Pax Accomodated represents the y-value, the %Pax Pre-emptively Rebooked represents the x-value, and 0.70 is the slope (all equations are in table 17). Subsequent equations follow a similar convention.

\[
\%\text{Pax Accomodated} = 0.70 \cdot \% \text{Pax Pre-emptively Rebooked}
\]

<table>
<thead>
<tr>
<th>Preemptively rebooked passengers</th>
<th>Same day Early</th>
<th>Same +Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>m</td>
<td>R²</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.73</td>
<td>1</td>
</tr>
<tr>
<td>20-Jan</td>
<td>0.69</td>
<td>1</td>
</tr>
<tr>
<td>23-Jan</td>
<td>0.84</td>
<td>1</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.67</td>
<td>1</td>
</tr>
<tr>
<td>6-May</td>
<td>0.65</td>
<td>1</td>
</tr>
<tr>
<td>29-May</td>
<td>0.76</td>
<td>1</td>
</tr>
<tr>
<td>22-Jun</td>
<td>0.76</td>
<td>1</td>
</tr>
<tr>
<td>29-Jun</td>
<td>0.74</td>
<td>1</td>
</tr>
<tr>
<td>13-Jul</td>
<td>0.67</td>
<td>1</td>
</tr>
<tr>
<td>15-Jul</td>
<td>0.67</td>
<td>1</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.65</td>
<td>0.99</td>
</tr>
<tr>
<td>7-Nov</td>
<td>0.63</td>
<td>0.99</td>
</tr>
<tr>
<td>µ</td>
<td>0.70</td>
<td></td>
</tr>
<tr>
<td>σ</td>
<td>0.066</td>
<td></td>
</tr>
<tr>
<td>µ/σ</td>
<td>11.37</td>
<td>16</td>
</tr>
</tbody>
</table>
Case study 2: $H_0$: The number of additional passengers accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated: Rejected

Based on the Monte Carlo simulation results showing that a projected 70% for upper bound of passengers are accommodated on the same day which is 7% less than on the same + previous day, it is shown that the hypothesis $H_0$: The number of additional passengers accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated is rejected.

The next figure shows the percentage of passengers accommodated on the same day for the hub. All days with cancellations bigger than 20 shows a similar trend (more than 64% of passengers can be accommodated on the same day before the scheduled departure time).

![Figure 29 Percentage of passengers accommodated same day early](image-url)
Case Study 2: H0: Preemptive rebooking does not result in additional revenue to airlines: Rejected

Based on the Monte Carlo simulation results that a decrease in airfares refunds is for 64% of passengers, it is shown that the hypothesis H0: Preemptive rebooking does not result in additional revenue to airlines is rejected.

Preemptive rebooking also has the effect of reducing the number of overnight passengers that could be eligible for airline refunded tickets. The relationship between Airfares Not Refunded (ANR) and preemptive rebooking is represented by the following equation:

\[
\%\text{Decrease in Airfares Refunded} = 0.64 \cdot \% \text{ Pax Pre-emptively Rebooked}
\]

For the same day preemptively rebooked the slope for ANR is 64%, giving an ANR of 45% when 70% of the passengers accept early rebooking (table 18).

<table>
<thead>
<tr>
<th>Date</th>
<th>Slope (m)</th>
<th>R²</th>
<th>Date</th>
<th>Slope (m)</th>
<th>R²</th>
</tr>
</thead>
<tbody>
<tr>
<td>12-Jan</td>
<td>0.58</td>
<td>0.9991</td>
<td>12-Jan</td>
<td>0.72</td>
<td>0.9993</td>
</tr>
<tr>
<td>20-Jan</td>
<td>0.46</td>
<td>0.9959</td>
<td>20-Jan</td>
<td>0.66</td>
<td>0.9994</td>
</tr>
<tr>
<td>23-Jan</td>
<td>0.65</td>
<td>0.9938</td>
<td>23-Jan</td>
<td>0.65</td>
<td>0.9942</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.52</td>
<td>0.9997</td>
<td>23-Feb</td>
<td>0.60</td>
<td>0.9998</td>
</tr>
<tr>
<td>6-May</td>
<td>0.62</td>
<td>0.9894</td>
<td>6-May</td>
<td>0.82</td>
<td>0.9975</td>
</tr>
<tr>
<td>29-May</td>
<td>0.82</td>
<td>0.9982</td>
<td>29-May</td>
<td>0.88</td>
<td>0.9985</td>
</tr>
<tr>
<td>22-Jun</td>
<td>0.92</td>
<td>0.9953</td>
<td>22-Jun</td>
<td>0.94</td>
<td>0.9963</td>
</tr>
<tr>
<td>29-Jun</td>
<td>0.39</td>
<td>0.996</td>
<td>29-Jun</td>
<td>0.64</td>
<td>0.9971</td>
</tr>
<tr>
<td>13-Jul</td>
<td>0.63</td>
<td>0.9997</td>
<td>13-Jul</td>
<td>0.73</td>
<td>0.9999</td>
</tr>
<tr>
<td>15-Jul</td>
<td>0.73</td>
<td>0.9977</td>
<td>15-Jul</td>
<td>0.76</td>
<td>0.997</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.60</td>
<td>0.9989</td>
<td>18-Jul</td>
<td>0.70</td>
<td>0.9994</td>
</tr>
</tbody>
</table>
Case Study 2 H0: Preemptive rebooking does not result in additional saving to Corporates: Rejected

Based on the results that an increase in Corporate Travel Expense savings for 64% of passengers, it is shown that the hypothesis $H_0$: Preemptive rebooking does not result in additional savings to corporates is rejected. Preemptive rebooking also has the effect of reducing the number of overnight passengers that could accrue unbudgeted Corporate Travel Expenses. The relationship between Corporate Travel Expense Savings (CNES) and preemptive rebooking is represented by the following equation:

$\%\text{Increase in CNES} = 0.64 \cdot \% \text{Pax Pre-emptively Rebooked}$

For the same day preemptively rebooked the slope CTES is 64% as well (table 19).

<table>
<thead>
<tr>
<th>Table 19 CTES (m and R²) same and same + previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>CTES</strong></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Same day Early</td>
</tr>
<tr>
<td>m</td>
</tr>
<tr>
<td>---</td>
</tr>
<tr>
<td>12-Jan</td>
</tr>
<tr>
<td>20-Jan</td>
</tr>
<tr>
<td>23-Jan</td>
</tr>
<tr>
<td>23-Feb</td>
</tr>
<tr>
<td>6-May</td>
</tr>
<tr>
<td>29-May</td>
</tr>
<tr>
<td>Date</td>
</tr>
<tr>
<td>---------</td>
</tr>
<tr>
<td>22-Jun</td>
</tr>
<tr>
<td>29-Jun</td>
</tr>
<tr>
<td>13-Jul</td>
</tr>
<tr>
<td>15-Jul</td>
</tr>
<tr>
<td>18-Jul</td>
</tr>
<tr>
<td>7-Nov</td>
</tr>
</tbody>
</table>

Case Study 2: \( H_0 \): Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked: Rejected

Based on the results that number of overnight passengers is decreased up to 65% (same day) to 75% (same + previous day) and PTD is decreased 53% (same day) to 64% (same + previous day) it is shown that hypothesis \( H_0 \) : Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked is rejected.

As noted above, preemptive rebooking has the effect of reducing the number of overnight passengers and decreasing PTD. The relationship between stranded passengers (next day passengers and passengers that cannot be rebooked) and preemptive rebooking is represented by the following equation:

\[
\% \text{Decrease in SPAX} = 0.65 \cdot \% \text{Pax Pre-emptively Rebooked}
\]

And the relationship between PTD and preemptive rebooking is represented by following equation:

\[
\% \text{Decrese in PTD} = 0.53 \cdot \% \text{Pax Pre-emptively Rebooked}
\]

The tables 20 and 21 show all results for stranded passengers and PTD.
Table 20 Slope and R\(^2\) for Overnight pax and pax that cannot be rebooked

<table>
<thead>
<tr>
<th></th>
<th>Next day pax &amp; Pax that cannot be rebooked</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Same day Early</td>
<td>Same +Previous day</td>
</tr>
<tr>
<td></td>
<td>m</td>
<td>R(^2)</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.59</td>
<td>1.00</td>
</tr>
<tr>
<td>20-Jan</td>
<td>0.48</td>
<td>0.99</td>
</tr>
<tr>
<td>23-Jan</td>
<td>0.69</td>
<td>0.99</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.52</td>
<td>1.00</td>
</tr>
<tr>
<td>6-May</td>
<td>0.65</td>
<td>0.99</td>
</tr>
<tr>
<td>29-May</td>
<td>0.83</td>
<td>1.00</td>
</tr>
<tr>
<td>22-Jun</td>
<td>0.97</td>
<td>0.99</td>
</tr>
<tr>
<td>29-Jun</td>
<td>0.40</td>
<td>0.99</td>
</tr>
<tr>
<td>13-Jul</td>
<td>0.64</td>
<td>1.00</td>
</tr>
<tr>
<td>15-Jul</td>
<td>0.65</td>
<td>1.00</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.61</td>
<td>1.00</td>
</tr>
<tr>
<td>7-Nov</td>
<td>0.62</td>
<td>1.00</td>
</tr>
<tr>
<td>µ</td>
<td>0.65</td>
<td></td>
</tr>
<tr>
<td>σ</td>
<td>0.15</td>
<td></td>
</tr>
<tr>
<td>µ/σ</td>
<td>4.43</td>
<td></td>
</tr>
</tbody>
</table>

Table 21 Slope and R\(^2\): PTD

<table>
<thead>
<tr>
<th></th>
<th>Same day Early</th>
<th>Same +Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>m</td>
<td>R(^2)</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.48</td>
<td>1.00</td>
</tr>
<tr>
<td>20-Jan</td>
<td>0.47</td>
<td>1.00</td>
</tr>
<tr>
<td>23-Jan</td>
<td>0.54</td>
<td>0.97</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.39</td>
<td>1.00</td>
</tr>
<tr>
<td>6-May</td>
<td>0.55</td>
<td>0.98</td>
</tr>
<tr>
<td>29-May</td>
<td>0.74</td>
<td>1.00</td>
</tr>
<tr>
<td>22-Jun</td>
<td>0.65</td>
<td>0.99</td>
</tr>
<tr>
<td>29-Jun</td>
<td>0.44</td>
<td>1.00</td>
</tr>
<tr>
<td>13-Jul</td>
<td>0.47</td>
<td>1.00</td>
</tr>
<tr>
<td>15-Jul</td>
<td>0.46</td>
<td>0.99</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.52</td>
<td>1.00</td>
</tr>
</tbody>
</table>
Case Study 2: UA, ORD, Summary of results
This case study rejects all $H_0$ hypotheses. The analysis is shown in Table 22.

<table>
<thead>
<tr>
<th>Analysis</th>
<th>Same Day Preemptive</th>
<th>Same Day+Previous Day Preemptively</th>
</tr>
</thead>
<tbody>
<tr>
<td>% Pax Rebooked</td>
<td>$m_\mu$ 0.70</td>
<td>$m_\sigma$ 0.07</td>
</tr>
<tr>
<td>%AF Not Refund</td>
<td>0.63 0.14</td>
<td>4.60</td>
</tr>
<tr>
<td>%CTES</td>
<td>0.64 0.14</td>
<td>4.57</td>
</tr>
</tbody>
</table>

Pre-emptive rebooking in advance of forecast large-scale cancellation events is feasible. At least 72% of the passengers seeking pre-emptive rebooking can be accommodated before their original scheduled flight time. The remaining 28% of the passengers seeking pre-emptive rebooking cannot be re-accommodated due to insufficient seats.

Pre-emptive rebooking on the previous day was not required as it did not significantly change the percentage of passengers re-accommodated.
The signal to noise ratio in Table shows that signal is 11 (same day) and 16 (same + previous day) times stronger than noise (for passengers accommodated case), and 4.6 (same day) and 7.4 (same + previous day) for ANR (table 10). The worst is for CTES, 4.6 for the same day and 3.7 for the previous + same day (table 11). In the case of CTES the best is if the passengers are rebooked on the same day because on other days (previous or next) they face extra costs due to lodging, etc.

Pre-emptive rebooking also creates a win-win for all the stakeholders. Airlines are able to recoup up to 0.74% • %Passengers Rebook Preemptively that otherwise might be refunded. Corporations sponsoring business travel for their employees also benefit by saving up to 0.64 • %Passengers Rebook Preemptively through unbudgeted overnight costs.

**Case Study 3**

Case study 3 is done for United Airlines, year 2012 and hubs: ERW and BOS, figure 30 shows the number of cancellations per day for UA.

The biggest cancellation events are shown in the table 23. A high percentage of cancellations was on November 7th (61%). Boston had the smallest number of flights per day. These hubs show similar results like ORD in the two previous studies.
Figure 30 Number of cancelled flights per day

Table 23 Days with the biggest number of cancelled flights

<table>
<thead>
<tr>
<th>&gt;60</th>
<th>month</th>
<th>day</th>
<th>cancellation</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1</td>
<td>20</td>
<td>101</td>
</tr>
<tr>
<td></td>
<td>6</td>
<td>22</td>
<td>68</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>18</td>
<td>78</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>26</td>
<td>78</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>7</td>
<td>208</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>8</td>
<td>85</td>
</tr>
</tbody>
</table>
Baseline Rebooking (After Cancelled Flight)
Similar to the Case Study 2, for the existing re-booking paradigm, no passengers will be booked on preemptive flights, so the percent of passengers rebooked early will automatically be zero. The result is a slope of zero in the equation calculating the percent of passengers rebooked early. For similar reasons, the ANR and CTES slopes will also be zero.

Table 24 Preemptively rebooked passengers

<table>
<thead>
<tr>
<th></th>
<th>Preemptively rebooked passengers</th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Same day Early</td>
<td>Same +Previous day</td>
<td>Previous day</td>
<td>Same Day</td>
<td>Prev+Same</td>
</tr>
<tr>
<td>Den Feb 3rd</td>
<td>0.44</td>
<td>1.00</td>
<td>0.27</td>
<td>1.00</td>
<td>0.44</td>
</tr>
<tr>
<td>Den Jun 7th</td>
<td>0.68</td>
<td>1.00</td>
<td>0.27</td>
<td>1.00</td>
<td>0.68</td>
</tr>
<tr>
<td>EWR July 18th</td>
<td>0.82</td>
<td>1.00</td>
<td>0.27</td>
<td>1.00</td>
<td>0.82</td>
</tr>
<tr>
<td>EWR Aug 14th</td>
<td>0.81</td>
<td>1.00</td>
<td>0.27</td>
<td>1.00</td>
<td>0.81</td>
</tr>
<tr>
<td>µ</td>
<td>0.69</td>
<td></td>
<td>0.27</td>
<td></td>
<td>0.69</td>
</tr>
<tr>
<td>σ</td>
<td>0.18</td>
<td></td>
<td>0.00</td>
<td></td>
<td>0.18</td>
</tr>
<tr>
<td>µ/σ</td>
<td>3.82</td>
<td></td>
<td>3.83</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Case study 3: *H₀: Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%)*
Based on the results that a projected 69% as the upper bound, of the passengers willing to be preemptively rebooked (Table 24), will be accommodated it is shown that the hypothesis that Preemptive rebooking does not accommodate more than 10%
passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%) is rejected.

**Case study 3: H₀: The number of additional pax accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated**

Based on the results that a projected 69%, as the upper bound for passengers accommodated on the same day, which is the same as in the same + previous day (table 24), it is shown that the hypothesis H₀: The number of additional passengers accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated is rejected.

**Case Study 3: H₀: Preemptive rebooking does not result in additional revenue to airlines: Rejected**

Based on the results that airlines can save in ANF up to $1M for the same day and $1.2M for previous day preemptively rebooked passengers for not refunded tickets (upper bound), it is shown that the hypothesis H₀ Preemptive rebooking does not result in additional revenue to airlines is rejected.

The full table of estimated savings is shown in the table 25.

---

**Table 25 Savings in not refunded airfares**

<table>
<thead>
<tr>
<th>ANR</th>
<th>% of pax accepted preemptively rebooking</th>
<th>Same</th>
<th>Prev</th>
</tr>
</thead>
<tbody>
<tr>
<td>Den Feb 3rd</td>
<td>0</td>
<td>$0</td>
<td>$0</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>$21K</td>
<td>$43K</td>
</tr>
<tr>
<td></td>
<td>30</td>
<td>$60K</td>
<td>$125K</td>
</tr>
</tbody>
</table>
The relationship between Airfares Not Refunded (ANR) and preemptive rebooking is represented by the following equation:

\[
% Decrease in Airfares Refunded = 0.78 \cdot % Pax Pre-emptively Rebooked
\]

(Table 26).

<table>
<thead>
<tr>
<th></th>
<th>Den Jun 7th</th>
<th>EWR July 18th</th>
<th>EWR Aug 14th</th>
</tr>
</thead>
<tbody>
<tr>
<td>50</td>
<td>$98K</td>
<td>$64K</td>
<td>$47K</td>
</tr>
<tr>
<td>70</td>
<td>$197K</td>
<td>$192K</td>
<td>$134K</td>
</tr>
<tr>
<td>10</td>
<td>$38K</td>
<td>$310K</td>
<td>$203K</td>
</tr>
<tr>
<td>30</td>
<td>$136K</td>
<td>$196K</td>
<td>$136K</td>
</tr>
<tr>
<td>50</td>
<td>$167K</td>
<td>$192K</td>
<td>$203K</td>
</tr>
<tr>
<td>70</td>
<td>$254K</td>
<td>$399K</td>
<td>$253K</td>
</tr>
</tbody>
</table>
### Case Study 3 H0: Preemptive rebooking does not result in additional saving to Corporates: Rejected

Based on the results that an increase in Corporate Travel Expense savings is for 64% it is shown that hypothesis $H_0$ : Preemptive rebooking does not result in additional savings to corporates is Rejected.

Preemptive rebooking also has the effect of reducing the number of overnight passengers that could accrue unbudgeted Corporate Travel Expenses. The relationship between Corporate Travel Expense Savings (CNES) and preemptive rebooking is represented by the following equation:

$$%\text{Increase in CNES} = 0.78 \cdot % \text{ Pax Pre-emptively Rebooked}$$
Table 27 UA, Hubs, CTES

<table>
<thead>
<tr>
<th></th>
<th>Same day Early</th>
<th>Same +Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>M</td>
<td>R</td>
</tr>
<tr>
<td>Den Feb 3rd</td>
<td>0.31</td>
<td>1.00</td>
</tr>
<tr>
<td>Den Jun 7th</td>
<td>0.68</td>
<td>0.99</td>
</tr>
<tr>
<td>EWR July 18th</td>
<td>1.07</td>
<td>0.99</td>
</tr>
<tr>
<td>EWR Aug 14th</td>
<td>1.06</td>
<td>0.98</td>
</tr>
<tr>
<td>(\mu)</td>
<td>0.78</td>
<td></td>
</tr>
<tr>
<td>(\Sigma)</td>
<td>0.37</td>
<td></td>
</tr>
<tr>
<td>(\mu/\sigma)</td>
<td>2.14</td>
<td></td>
</tr>
</tbody>
</table>

The slope for the same + previous day is lower (0.66, shown in the Table 27). The reason is that if more passengers are rebooked on the previous day, some of them have extra (unplanned) expenses for one more night (for hotel, transportation etc.) For corporates standpoint same day early is the best choice.

Case Study 3: \(H_0\): Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked

Based on the results that number of overnight passengers is decreased by 54% and PTD is decreased up 53% (upper bound for 70% passengers willing to accept preemptively rebooking), it is shown that hypothesis \(H_0\): Preemptive rebooking does not
and result in reducing PTD and number of overnight and passenger which cannot be rebooked is rejected.

The results for PTD are shown in the table 28.

<table>
<thead>
<tr>
<th>PTD</th>
<th>% Improvement</th>
<th>% Improvement</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Same Day</td>
<td>Same + Prev Day</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Den Feb 3</td>
<td>0</td>
<td>1663</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>1614</td>
</tr>
<tr>
<td></td>
<td>30</td>
<td>1530</td>
</tr>
<tr>
<td></td>
<td>50</td>
<td>1451</td>
</tr>
<tr>
<td></td>
<td>70</td>
<td>1378</td>
</tr>
<tr>
<td>Den Jun 7</td>
<td>0</td>
<td>1015</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>945</td>
</tr>
<tr>
<td></td>
<td>30</td>
<td>808</td>
</tr>
<tr>
<td></td>
<td>50</td>
<td>695</td>
</tr>
<tr>
<td></td>
<td>70</td>
<td>603</td>
</tr>
<tr>
<td>EWR 18 Jul</td>
<td>0</td>
<td>1530</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>1413</td>
</tr>
<tr>
<td></td>
<td>30</td>
<td>1147</td>
</tr>
<tr>
<td></td>
<td>50</td>
<td>925</td>
</tr>
<tr>
<td></td>
<td>70</td>
<td>753</td>
</tr>
<tr>
<td>EWR 14 Aug</td>
<td>0</td>
<td>918</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>814</td>
</tr>
<tr>
<td></td>
<td>30</td>
<td>629</td>
</tr>
<tr>
<td></td>
<td>50</td>
<td>494</td>
</tr>
<tr>
<td></td>
<td>70</td>
<td>396</td>
</tr>
</tbody>
</table>

As noted above, preemptive rebooking has the effect of reducing the number of overnight passengers and decreasing PTD. The relationship between stranded passengers
(next day passengers and passengers that cannot be rebooked) and preemptive rebooking is represented by the following equation:

\[ \% \text{Decrease in SPAX} = 0.78 \times \% \text{Pax Pre-emptively Rebooked} \]

And the relationship between PTD and preemptive rebooking is represented by the following equation:

\[ \% \text{Decrease in PTD} = 0.53 \times \% \text{Pax Pre-emptively Rebooked} \]

<table>
<thead>
<tr>
<th>Table 29 Overnight passengers improvement</th>
</tr>
</thead>
<tbody>
<tr>
<td>SPAX IMPROVEMENT</td>
</tr>
<tr>
<td>% Improvement</td>
</tr>
<tr>
<td>% of pax willing to be preemptively rebooked</td>
</tr>
<tr>
<td>Den Feb 3</td>
</tr>
<tr>
<td>0</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>30</td>
</tr>
<tr>
<td>50</td>
</tr>
<tr>
<td>70</td>
</tr>
<tr>
<td>Den Jun 7</td>
</tr>
<tr>
<td>0</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>30</td>
</tr>
<tr>
<td>50</td>
</tr>
<tr>
<td>70</td>
</tr>
<tr>
<td>EWR 18 Jul</td>
</tr>
<tr>
<td>0</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>30</td>
</tr>
<tr>
<td>50</td>
</tr>
<tr>
<td>70</td>
</tr>
<tr>
<td>EWR 14 Aug</td>
</tr>
<tr>
<td>0</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>--------</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>SPAX - slope</th>
<th>same</th>
<th>prev</th>
</tr>
</thead>
<tbody>
<tr>
<td>Den Feb 3</td>
<td>0.30</td>
<td>0.60</td>
</tr>
<tr>
<td>Den Jun 7</td>
<td>0.68</td>
<td>0.83</td>
</tr>
<tr>
<td>EWR 18 Jul</td>
<td>1.07</td>
<td>1.08</td>
</tr>
<tr>
<td>EWR 14 Aug</td>
<td>1.06</td>
<td>1.18</td>
</tr>
<tr>
<td>µ</td>
<td>0.78</td>
<td>0.92</td>
</tr>
<tr>
<td>σ</td>
<td>0.37</td>
<td>0.26</td>
</tr>
<tr>
<td>µ/σ</td>
<td>2.12</td>
<td>3.54</td>
</tr>
</tbody>
</table>

**Case Study 3: UA, Summary of results**

Case Study 3 results are summarized in the table 30. Compared to Case Study 2, the percent passengers rebooked statistics is nearly the same as the same day preemptive, and within 0.2 for all other statistics.

<table>
<thead>
<tr>
<th>Analysis</th>
<th>Same Day Preemptive</th>
<th>Same Day+Previous Day Preemptively</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>mµ</td>
<td>mσ</td>
</tr>
<tr>
<td>Slope for % Pax Rebooked</td>
<td>0.69</td>
<td>0.18</td>
</tr>
<tr>
<td>Slope for ANR</td>
<td>0.78</td>
<td>0.36</td>
</tr>
<tr>
<td>Slope for CTES</td>
<td>0.78</td>
<td>0.36</td>
</tr>
</tbody>
</table>

Case Study 4
A Case Study 4 was conducted for the scheduled domestic flights for Southwest Airlines operating from Chicago Midway International Airport (MDW) for all one day cancellation events in 2012.

As is shown in Figure 31, Southwest didn’t have a big number of cancellations during 2012. Out of 1,140,535 flights, 9580 flights were cancelled (0.84%) For MDW that statistics was similar: out of 158,515, 1368 were cancelled flights (0.86%).
Figure 31 Cancellations per day, Southwest

Top cancellations events are shown in the table 31, the biggest event was on December 20\textsuperscript{th}, when 160 flights were cancelled:

Table 31 SW, MDW, Top cancellation events

<table>
<thead>
<tr>
<th>month</th>
<th>day</th>
<th>Number of cancelled flights</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>----</td>
<td>----</td>
<td>----</td>
</tr>
<tr>
<td>2</td>
<td>24</td>
<td>30</td>
</tr>
<tr>
<td>7</td>
<td>18</td>
<td>30</td>
</tr>
<tr>
<td>10</td>
<td>31</td>
<td>30</td>
</tr>
<tr>
<td>12</td>
<td>26</td>
<td>30</td>
</tr>
<tr>
<td>11</td>
<td>21</td>
<td>40</td>
</tr>
<tr>
<td>6</td>
<td>28</td>
<td>60</td>
</tr>
<tr>
<td>10</td>
<td>30</td>
<td>80</td>
</tr>
<tr>
<td>1</td>
<td>20</td>
<td>100</td>
</tr>
<tr>
<td>10</td>
<td>29</td>
<td>100</td>
</tr>
<tr>
<td>1</td>
<td>12</td>
<td>130</td>
</tr>
<tr>
<td>2</td>
<td>23</td>
<td>150</td>
</tr>
<tr>
<td>12</td>
<td>20</td>
<td>160</td>
</tr>
</tbody>
</table>

**Baseline Rebooking (After Cancelled Flight)**

The same as for Case Studies 2 and 3, for the existing re-booking paradigm, no passengers will be booked on preemptive flights, so the percent of passengers rebooked early will automatically be zero. The result is a slope of zero in the equation calculating the percent of passengers rebooked early. For similar reasons, the ANR and CTES slopes will also be zero.
Case study 4: \( H_0: \) Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%)

Based on the results that a projected 0.56%, as the upper bound of the passengers willing to be preemptively rebooked, will be accommodated it is shown that the hypothesis, Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%) is rejected.

During 2012 MDW had one rare event which had extremely big impact on the results. That was for November 21\textsuperscript{st}. Because of the fact that this was an unusual event, it will be observe separately. The results for all days are given in the table 32.

<table>
<thead>
<tr>
<th>Table 32 Preemptively rebooked passengers (m, R2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Preemptively rebooked passengers</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Same day</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Early</td>
</tr>
<tr>
<td>m</td>
</tr>
<tr>
<td>20-Jan</td>
</tr>
<tr>
<td>23-Feb</td>
</tr>
<tr>
<td>7-Jun</td>
</tr>
<tr>
<td>18-Jul</td>
</tr>
<tr>
<td>21-Nov</td>
</tr>
<tr>
<td>20-Dec</td>
</tr>
<tr>
<td>26-Dec</td>
</tr>
<tr>
<td>(\mu)</td>
</tr>
<tr>
<td>(\sigma)</td>
</tr>
<tr>
<td>(\mu/\sigma)</td>
</tr>
</tbody>
</table>
Case study 4: H₀: The number of additional pax accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated

For Case Study 4 this hypothesis is not rejected. November 21st started with a big number of cancelled flights early in the morning (5am). The cancellations during the day are at the noise level, as in the next figure.

![Cancellations during the day](image)

**Figure 32** Cancellations during February 24th

In total, 33 cancelled flights affected 4664 passengers. The last cancelled flight was at 11.30. Because of this situation, just a maximum of 37% of cancelled passengers willing to accept preemptive rebooking are able to be accommodated on the same day early (Figure 33). On the other hand, previous day and same day later show good results.
Case Study 4: H0: Preemptive rebooking does not result in additional revenue to airlines

Based on the results that a projected 69% for upper bound for passengers accommodated on the same day which is the same like on the same + previous day (table 33), it is shown that hypothesis H0 :The number of additional passengers accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated is rejected.

Table 33 SW, MDW, Airfares Not Refund

<table>
<thead>
<tr>
<th>Airfares Not Refund</th>
<th>Same day Early</th>
<th>Same +Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Case Study 4

**H0: Preemptive rebooking does not result in additional saving to Corporates**

This hypothesis can be rejected just for the previous + same day preemptive rebooking (0.69%). For the same day early it cannot be rejected, because of the situation that passengers cannot be rebooked for the same day early. CTES depends of extra cost for overnight stays. The CTES for November 21st is shown in the figure 34.
The result for CTES for 2012 for MDW is shown in the Table 34.

<table>
<thead>
<tr>
<th>Table 34 SW, MDW, CTES</th>
</tr>
</thead>
<tbody>
<tr>
<td>Same day Early</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>20-Jan</td>
</tr>
<tr>
<td>23-Feb</td>
</tr>
<tr>
<td>7-Jun</td>
</tr>
<tr>
<td>18-Jul</td>
</tr>
<tr>
<td>21-Nov</td>
</tr>
<tr>
<td>20-Dec</td>
</tr>
<tr>
<td>26-Dec</td>
</tr>
<tr>
<td>μ</td>
</tr>
<tr>
<td>σ</td>
</tr>
<tr>
<td>μ/σ</td>
</tr>
</tbody>
</table>
Case Study 4: H₀: Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked

Based on the results that number of overnight passengers is decreased by 54% and PTD is decreased by up to 53% (upper bound for 70% passengers willing to accept preemptively rebooking), it is shown that hypothesis H₀: Preemptive rebooking does not and result in reducing PTD and number of overnight and passenger which cannot be rebooked is rejected.

As noted earlier, preemptive rebooking has the effect of reducing the number of overnight passengers and decreasing PTD. The relationship between stranded passengers (next day passengers and passengers that cannot be rebooked) and preemptive rebooking is represented by the following equation:

$$\%\text{Decrease in SPAX} = 0.76 \times \%\text{Pax Pre-emptively Rebooked}$$

Case Study 4: SW, MDW, Summary of results

This case study shows that if heavy cancellations occur early in the day, the number which can be rebooked same-day preemptive is minimal. In this case the biggest difference in the results show in the % of passengers accommodated same day early and CTES. The results are summarized in Table 35.

<table>
<thead>
<tr>
<th>Table 35 Summary of results, Case Study 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Analysis</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td></td>
</tr>
</tbody>
</table>
Relation of Preemptive rebooking acceptance rate to pax rebooked, ANR and CTES without Nov 21st (Table 36):

The strong impact of November 21st event shows that a big factor is the type of event (e.g. number of cancellations, when they start and how they are grouped).
Case Study 5

A case-study was conducted for the scheduled domestic flights for Southwest for all other hubs (except MDW, which is analyzed in the Case Study 4) for all one day cancellation events in 2012. In the study 10 airports are observed:

Airports:
- MDW
- LAS
- BWI
- DAL
- PHX
- DEN
- HOU
- ATL
- LAX
- MCO

For these 10 airports the biggest cancellation events are shown in the Table 37.

<table>
<thead>
<tr>
<th>Table 37 Top cancelled event, SW</th>
</tr>
</thead>
<tbody>
<tr>
<td>Month</td>
</tr>
<tr>
<td>-------</td>
</tr>
<tr>
<td>1</td>
</tr>
<tr>
<td>1</td>
</tr>
<tr>
<td>2</td>
</tr>
</tbody>
</table>
Because of the big number of airports in the study just the events with more than 100 cancelled flights are observed.

**Baseline Rebooking (After Cancelled Flight)**

As in the previous studies, for the existing re-booking paradigm, no passengers will be booked on preemptive flights, so the percent of passengers rebooked early will automatically be zero. The result is a slope of zero in the equation calculating the percent of passengers rebooked early. For similar reasons, the ANR and CTES slopes will also be zero.

**Case study 5: H₀: Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%)**

Based on the results that a projected 62% • % Pax Pre-emptively Rebooked for upper bound of the passengers willing to be preemptively rebooked, will be
accommodated it is shown that the hypothesis Preemptive rebooking does not accommodate more than 10% passengers on cancelled flights (Preemptive passengers / total cancel passengers < 10%) is rejected. The relationship between percentage passengers accommodated and preemptive rebooking is represented by the following equation:

\[
\%\text{Pax Accomodated} = 0.62 \times \% \text{ Pax Pre-emptively Rebooked}
\]

**Case study 5: H_0: The number of additional pax accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated**

Based on the results from the table below the upper bound for passengers accommodated on the same day is not greater than 25%, it is shown that hypothesis H_0 :The number of additional passengers accommodated by adding an additional day of preemptive rebooking is greater than 25% of the original number of accommodated is rejected (Table 38). Same day preemptively can accommodate 0.62% \times \% \text{ Pax Pre-emptively Rebooked} and with Same+Previous day option for 0.17 is accommodate on the previous day (with the same, 0.62 on the same day).

<table>
<thead>
<tr>
<th>Preemptively rebooked passengers</th>
<th>Same day Early</th>
<th>Same +Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Previous day</td>
<td>Same Day</td>
</tr>
<tr>
<td>M</td>
<td>M</td>
<td>M</td>
</tr>
<tr>
<td>R</td>
<td>R</td>
<td>R</td>
</tr>
</tbody>
</table>
Case Study 5: H0: Preemptive rebooking does not result in additional revenue to airlines:

Based on the results that airlines can increase their savings (ANF) shown in the table 39 it is shown that hypothesis H0 Preemptive rebooking does not result in additional revenue to airlines is rejected.

Table 39 ANR, SW, 10 hubs (slope and R2)

<table>
<thead>
<tr>
<th></th>
<th>Airfares Not Refund</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Same day Early</td>
</tr>
<tr>
<td></td>
<td>m</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.99</td>
</tr>
<tr>
<td>3-Feb</td>
<td>0.17</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.80</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.87</td>
</tr>
<tr>
<td>17-Dec</td>
<td>0.59</td>
</tr>
</tbody>
</table>
The relationship between Airfares Not Refunded (ANR) and preemptive rebooking is represented by the following equation:

\[ \% \text{Decrease in Airfares Refunded} = 0.69 \times \% \text{ Pax Pre-emptively Rebooked} \]

**Case Study 5 H0: Preemptive rebooking does not result in additional saving to Corporates:**

Based on the results that increase is Corporate Travel Expense savings for 74\% for the same day early, it is shown that hypothesis H₀ : Preemptive rebooking does not result in additional savings to corporates is rejected.

As it is said, preemptive rebooking has the effect of reducing the number of overnight passengers that could accrue unbudgeted Corporate Travel Expenses. The relationship between Corporate Travel Expense Savings (CNES) and preemptive rebooking is represented by the following equation:

\[ \% \text{Increase in CNES} = 0.74 \times \% \text{ Pax Pre-emptively Rebooked} \]

<table>
<thead>
<tr>
<th></th>
<th>CTES</th>
<th>Same day Early</th>
<th>Same +Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>m R</td>
<td>m R</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.94 0.99</td>
<td>0.84 0.98</td>
<td></td>
</tr>
<tr>
<td>3-Feb</td>
<td>0.17 1.00</td>
<td>3.37 1.00</td>
<td></td>
</tr>
</tbody>
</table>
Case Study 5: $H_0$: Preemptive rebooking does not result in reducing PTD and number of overnight and passenger which cannot be rebooked:

Based on the results that show PTD is decreased up to 56% • % Pax Pre-emptively Rebooked (upper bound for 70% passengers willing to accept preemptively rebooking), it is shown that hypothesis $H_0$: Preemptive rebooking does not and result in reducing PTD and number of overnight and passenger which cannot be rebooked is rejected.

Preemptive rebooking has the effect of reducing the number of overnight passengers and decreasing PTD (Table 41).

The relationship between stranded passengers (next day passengers and passengers that cannot be rebooked) and preemptive rebooking is represented by the following equation:

$$\%\text{Decrease in SPAX} = 0.64 \times \% \text{Pax Pre-emptively Rebooked}$$

The relationship between PTD and preemptive rebooking is represented by the following equation:

$$\%\text{Decrease in SPAX} = 0.56 \times \% \text{Pax Pre-emptively Rebooked}$$
### Table 41 Overnight and Remaining passengers and PTD statistics

<table>
<thead>
<tr>
<th>Overnight &amp; Remaining Passengers</th>
<th>Same day Early</th>
<th>Same + Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>m</td>
<td>R</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.94</td>
<td>0.99</td>
</tr>
<tr>
<td>3-Feb</td>
<td>0.17</td>
<td>1.00</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.79</td>
<td>1.00</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.86</td>
<td>1.00</td>
</tr>
<tr>
<td>17-Dec</td>
<td>0.39</td>
<td>1.00</td>
</tr>
<tr>
<td>20-Dec</td>
<td>0.72</td>
<td>1.00</td>
</tr>
<tr>
<td>µ</td>
<td>0.64</td>
<td></td>
</tr>
<tr>
<td>σ</td>
<td>0.30</td>
<td></td>
</tr>
<tr>
<td>µ/σ</td>
<td>2.15</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>PTD</th>
<th>Same day Early</th>
<th>Same + Previous day</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>m</td>
<td>R</td>
</tr>
<tr>
<td>12-Jan</td>
<td>0.79</td>
<td>1.00</td>
</tr>
<tr>
<td>3-Feb</td>
<td>0.17</td>
<td>1.00</td>
</tr>
<tr>
<td>23-Feb</td>
<td>0.73</td>
<td>1.00</td>
</tr>
<tr>
<td>18-Jul</td>
<td>0.72</td>
<td>1.00</td>
</tr>
<tr>
<td>17-Dec</td>
<td>0.37</td>
<td>1.00</td>
</tr>
<tr>
<td>20-Dec</td>
<td>0.57</td>
<td>1.00</td>
</tr>
<tr>
<td>µ</td>
<td>0.56</td>
<td></td>
</tr>
<tr>
<td>Σ</td>
<td>0.24</td>
<td></td>
</tr>
<tr>
<td>µ/σ</td>
<td>2.30</td>
<td></td>
</tr>
</tbody>
</table>

---

**Case Study 5: SW, Summary of results**

Results for the case study are summarized in the Table 42. Comparing with the summary of the case study 3 (UA, all hubs) statistics for the same day preemptive are
within 0.07 for percentage of passengers rebooked, 0.09 for ANR and 0.04 for CTES.

Signal to noise ratio is similar to the statistics for the same day preemptively case study 3.

<table>
<thead>
<tr>
<th>Analysis</th>
<th>Same Day Preemptive</th>
<th>Same Day+Previous Day Preemptively</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>m$\mu$</td>
<td>m$\sigma$</td>
</tr>
<tr>
<td>Slope for % Pax Rebooked</td>
<td>0.62</td>
<td>0.18</td>
</tr>
<tr>
<td>Slope for ANR</td>
<td>0.69</td>
<td>0.29</td>
</tr>
<tr>
<td>Slope for CTES</td>
<td>0.74</td>
<td>0.30</td>
</tr>
</tbody>
</table>
5. CONCLUSION AND FUTURE WORK

Although the large scale airline flight cancellations events are infrequent, they have significant impact on airline revenue, corporate travel expenses and passengers travel costs and inconvenience. In many cases these big events are reliably forecast enough in advance, so that it is possible to facilitate a pro-active response from the airlines. In this dissertation a new rebooking methodology, preemptive rebooking, was proposed as an option to accommodate airline passengers in case of flight cancellations. A Monte Carlo simulator was develop by enhancements of an existing code and additions of significant new modules for realistic modeling of airline rebooking, and then used to perform comprehensive analysis, using actual airline data, to validate the expected benefits of the proposed approach.

Significant contributions are described in chapter III, a PTD Calculator embedded in a Monte Carlo Simulation, which analyzes the impact of preemptive rebooking strategies. In the model passengers are divided into two big groups: passengers on the cancelled flights, and passengers on not cancelled flights. This model analyzes the first group, which is then divided into: passengers that can be rebooked (traditional rebooking model: rebooked on same day later or on the first available flight on the next day; and preemptive model: same day before the scheduled departure time, previous + same day before, same day later and next day). The rebooking algorithm takes the available itinerary information and uses it to make rebooking decisions. It selects disrupted itineraries one by one, over all of the passengers that need to be rebooked.
Those passengers are split into groups depending on whether they are willing to accept preemptive rebooking or not. For as long as passengers remain in need of rebooking, and for as long as alternate flights remain, remaining passengers are placed in empty seats.

Rebooking alternatives are considered in order beginning with flights departing prior to, but on the same day as, the original scheduled departure of the flight, followed by flights departing on the previous day, followed by flights departing after, but on the same day, followed by flights departing on the next day. Additionally, in each group, alternate flights on different airlines were considered after all alternates on the same airline were exhausted. The capability to handle the order is new to the algorithm.

Due to the fact that some passengers rebook preemptively and some do not, a binomial distribution calculator was added to the code, and used to randomly determine the specific passengers which would accept preemptive rebooking for a given percentage of passengers willing to be rebooked. Passengers are rebooked First Come Fist Serve and randomization is done in Monte Carlo simulation.

In Chapter IV detailed case studies with realistic industrial data showed that preemptive rebooking in advance of forecast large-scale cancellation events is feasible and that decrease in PTD and number of overnight and passenger which cannot be rebooked is up to 50%, and a decrease in Remaining passengers is up to 55 %. Previous day rebooking is not required as a majority of the passengers (>90%) can be accommodated earlier on the same day.

The number of rebooked passenger on previous day, and same and next day, depends on several factors:
1) How many cancellations were on the previous/same day;

2) When the cancellations occur (morning, midday, afternoon);

3) The size of an event (1 day or more days event);

4) Number of flights previous day/same day/next day;

5) Number of available itineraries and the type of itineraries (number of “2-day” itineraries);

6) Load factors;

7) Frequency;

There is also an opportunity for airlines to generate additional revenue by offering a for-fee option that would move passengers to the front of the preemptive rebooking queue in the event of a large scale event (Raiteri, 2015).

In a future work it would be analyzed complexities that maybe applied in developing the model, for example rare extreme multi day events, how different load factors can influence results, and also considering the case that only partial information on cancelled flights is available (e.g. preemptive rebooking with possibility of unexpected cancellations) and increase fidelity of airfares refund model and corporate travel expenditure model.

The general premise that the ability to mass communicate with individual clients has become feasible is one that can be implied in more industries than in aviation. Is worth investigating whether concept similar to preemptive rebooking can be used with significant gain in other areas.
APPENDIX I

1) Stored procedure Binomial Random, which selects binomially distributed random values:

USE [PaxDelay]
GO
/****** Object: StoredProcedure [dbo].[binomial_rand] ******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ------------------------------
-- Author:  <Sanja,Avramovic>
-- Create date: <12/6/14>
-- Description: <This function selects binomially distributed random value>
/*
 Binomial sampler function, SQL version derived from R library
 + Copyright (C) 1998 Ross Ihaka
 + Copyright (C) 2000-2014 The R Core Team
 + Copyright (C) 2007 The R Foundation
 + Copyright (C) 2014 Sanja Avramovic
 +
 + This program is free software; you can redistribute it and/or modify
 + it under the terms of the GNU General Public License as published by
 + the Free Software Foundation; either version 2 of the License, or
 + (at your option) any later version.
 +
 + This program is distributed in the hope that it will be useful,
 + but WITHOUT ANY WARRANTY; without even the implied warranty of
 + MERCHANTABILITY or FITNESS FOR A PARTICULAR PURPOSE. See the
 + GNU General Public License for more details.
 +
 + You should have received a copy of the GNU General Public License
 + along with this program; if not, a copy is available at
 + http://www.r-project.org/Licenses/
 +
 + SYNOPSIS
 +
 + binomial_rand @nin int, @pp float
 + returns int
 +
 + DESCRIPTION
 +
 + Random variates from the binomial distribution.
 +
 + REFERENCE
 +
 + Binomial random variate generation.
 + Communications of the ACM 31, 216-222.
 + (Algorithm BTPEC).
 +*/
-- ------------------------------

Create PROCEDURE [dbo].[binomial_rand]
(
    @nin int,
    @pp  float
as
begin

set nocount on;

declare @c float;
declare @fm float;
declare @npq float;
declare @p1 float;
declare @p2 float;
declare @p3 float;
declare @p4 float;
declare @qn float;
declare @xl float;
declare @xll float;
declare @xlr float;
declare @xm float;
declare @xr float;
declare @psave float;
declare @nsave int;
declare @m int;

set @psave = -1;
set @nsave = -1;

declare @f float;
declare @f1 float;
declare @f2 float;
declare @u float;
declare @v float;
declare @w float;
declare @w2 float;
declare @x float;
declare @x1 float;
declare @x2 float;
declare @z float;
declare @z2 float;

declare @p float;
declare @q float;
declare @np float;
declare @g float;
declare @r float;
declare @a1 float;
declare @al float;
declare @alv float;
declare @ffm float;
declare @ynorm float;

declare @i int;
declare @ix int;
declare @k int;
declare @n int;

--if (!R_FINITE(nin)) ML_ERR_return_NAN;
set @r = @nin;
--if (r != nin) ML_ERR_return_NAN;
--if (!R_FINITE(pp) ||
  /* n=0, p=0, p=1 are not errors <TSL>*/
  if (@r < 0 OR @pp < 0. OR @pp > 1.) return -1;

if (@r = 0 OR @pp = 0.) return 0;
if (@pp = 1.) return @nin;

--if (r >= INT_MAX)/* evade integer overflow, and r == INT_MAX gave only even values */
--return qbinom(unif_rand(), r, pp, /*lower_tail*/ 0, /*log_p*/ 0);
/* else */
set @n = @nin;

set @p = @pp;
if (1 - @pp < @pp) set @p = 1 - @pp;
set @q = 1 - @p;
set @np = @n * @p;
set @r = @p / @q;
set @g = @r * (@q + 1);

/* Setup, perform only when parameters change [using static (globals): */
/* FIXING: Want this thread safe
-- use as little (thread globals) as possible */
if (@np != @psave || @n != @nsave)
  set @pp = @psave;
  set @nsave = @n;
end -- if (@np < 30.0) begin
  /* inverse cdf logic for mean less than 30 */
  set @qn = power(@q, @n);
goto L_np_small;
end -- if (@np < 30.0) else
else begin
  set @ffm = @np + @p;
  set @nm = cast(@ffm as int);
  set @fm = @m;
  set @npq = @np * @q;
  set @p1 = cast ((@v * @v); 6.6 * @q) as int) + 0.5;
  set @xm = @ff; + 0.5;
  set @x1 = @xm - @p1;
  set @xr = @xm + @p1;
  set @l = 0.134 + 20.5 / (15.3 + @ff);
  set @al = (@ff - @x1) / (@ff - @x1 * @p);
  set @xll = @al + (1.0 + 0.5 * @al);
  set @al = (@m) - (&xm) / (&xm - @x1 * @p);
  set @xlr = @al * (1.0 + 0.5 * @al);
  set @p2 = @p1 * (1.0 + @al + @al);
  set @p3 = @p2 + @al / @x1;
  set @p4 = @p3 + @al / @xlr;
end -- if (@np < 30.0) else
end -- if (@np != @nsave)
else if (@@np < 30.0) goto L_np_small;
end -- if (@n = @nsave)

/*----------------------------- np = n*pp >= 30: ------------------------ */
while 1=1 begin
  set @u = rand() * @p4;
  set @v = rand();
/* triangular region */
  if (@u <= @p1) begin
    set @x = cast((@xm - @p1 * @v + @u) as int);
goto finis;
  end -- if (@u <= @p1)
/* parallelogram region */
  if (@u <= @p2) begin
    set @x = @x1 + (@u - @p1) / @c;
    set @v = @v * @c + 1.0 - abs(@xm - @x) / @p1;
    if (@v > 1.0 OR @v <= 0.) continue;
    set @x = cast(@x as int);
end -- if (@u <= @p2)
else begin
    if (@u > @p3) begin  /* right tail */
        set @ix = cast((@vr - log(@v) / @xlr) as int);
        if (@ix > @m) continue;
        set @v = @v * (@u - @p3) * @xlr;
    end -- if (@u > @p3)
    else begin  /* left tail */
        set @ix = cast((@xl + log(@v) / @xll) as int);
        if (@ix < 0) continue;
        set @v = @v * (@u - @p2) * @xll;
    end -- if (@u > @p3) else
end -- if (@u <= @p2) else
/* determine appropriate end
for the final acceptance/rejection test */
if (@k <= 20 || @k >= @npq / 2) begin
    /* explicit evaluation */
    set @f = 1.0;
    if (@m < @ix) begin
        set @i = @m + 1;
        while @i <= @ix begin
            set @ff = @f * (@g / @i - @r);
            set @i = @i + 1;
        end -- while @i <= @ix
    end -- if (@m < @ix)
else if (@m != @ix) begin
    set @i = @ix + 1;
    while @i <= @m begin
        set @f = @f / (@g / @i - @r);
        set @i = @i + 1;
    end -- while @i <= @m
end -- if (@m != @ix)
if (@v <= @f) goto finis;
end -- if (@k <= 20 || @k >= @npq / 2 - 1)
else begin
    /* squeezing using upper and lower bounds on log(f(x)) */
    set @amaxp = (@k / @npq) * ((@k * (@k / 3.0 + 0.625) + 0.16666666666666667 / @npq + 0.5));
    set @ynorm = -@k * @k / (2.0 * @npq);
    set @alv = log(@v);
    if (@alv < @ynorm - @amaxp) goto finis;
    if (@alv <= @ynorm + @amaxp) begin
        /* stirling's formula to machine accuracy */
        /* for the final acceptance/rejection test */
        set @x1 = @ix + 1;
        set @f1 = @ff + 1.0;
        set @z = @m + 1 - @f1;
        set @w = @m - @ix + 1.0;
        set @al = @r * @z;
        set @al2 = @x1 * @al;
        set @f2 = @f1 * @f1;
        set @w2 = @w * @w;
        if (@al <= @x1 * log(@f1 / @x1) + (@m - @m + 0.5) * log(@al2)
            / @f1) + (@ix - @m) * log(@al2 * @p / (@x1 * @al)) + (13860.0 - (46.0 - (132.0 - (99.0 + 140.0)
            / @f2) / @f2) / @f2) / @f1 / (13860.0 - (46.0 - (132.0 - (99.0 + 140.0)
            / @z2) / @z2) / @z2) / @z2) / @z2) / @x2) / @x2) / @x2) / @x2) / @x2) / @x2) / @x2) / @x2) / @x2)
            / @x2) / @x2 / @x2 / @w2 / @w2) / @w2) / @w2 / @w2) / 166320.0)
            goto finis;
        end -- if (@al <= @ynorm + @amaxp)
    end -- if (@k <= 20 || @k >= @npq / 2 - 1) else
end -- while @i=1
L_np_small:
/*----------------------------- np = n*p < 30 :----------------------------- */
while $1=1$ begin
    set @ix = 0;
    set @f = @qn;
    set @u = rand();
    while $2=2$ begin
        if (@u < @f) goto finis;
        if (@ix > 110) break;
        set @u = @u - @f;
        set @ix = @ix + 1;
        set @f = @f * (@g / @ix - @r);
    end -- while 2=2
end -- while 1=1

finis:
    if (@psave > 0.5)
        set @ix = @n - @ix;
    return @ix;

end

2) Sored procedure Compute Total Pax, this procedure computes and returns the total number of passengers on a chosen origin/destination’s flight:

USE [PaxDelay]
GO
/****** Object:  StoredProcedure [dbo].[compute_total_pax]    Script Date: 3/15/2015 2:39:19 PM
******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO
-- ===========================
-- Author:  <Sanja Avramovic>
-- Create date: <10/22/2014>
-- Description: computes and returns the total number of pax on a chosen origin/dest/flight
-- on a chosen day.
-- Based on Guermo's and Ashwin's code
-- ===========================
CREATE PROCEDURE [dbo].[compute_total_pax]
(  
    @origin  varchar(3),
    @dest    varchar(3),
    @airline varchar(3),
    @fl_num  smallint,
    @start_date smalldatetime
)
AS
BEGIN
    set nocount on

    -- Declare the return variable here
    DECLARE @result int

    -- Compute Total Pax
    DECLARE @ix int = 0;
    DECLARE @f float = @qn;
    DECLARE @u float = rand();
    WHILE @ix < 110
    BEGIN
        IF @u < @f GOTO finis;
        IF @ix > 110 BREAK;
        SET @u = @u - @f;
        SET @ix = @ix + 1;
        SET @f = @f * (@g / @ix - @r);
    END -- WHILE

    RETURN @ix;

    finis:
    IF (@psave > 0.5)
        SET @ix = @n - @ix;
    RETURN @ix;

END

GO
-- count pax for one day, so the end date is one day ahead
DECLARE @end_date smalldatetime
set @end_date = dateadd(day,1,@start_date)

-- sum all pax with a matching origin, dest, airline, and flight
-- count the totals for all possibilities (origin/hub, hub/dest, and origin/dest)
set @result =
(select sum(a.total)+b.total+c.total)
      from
        ( select isnull(sum(max_pax),0) as total
              from MasterTable
              where
                origin = @origin
                and dest = @dest
                and airline = @airline
                and fl_1_num = @fl_num
                and hub is null
                and fl_1_sch_dep >= @start_date
                and fl_1_sch_dep < @end_date
          ) a,
        ( select isnull(sum(max_pax),0) as total
              from MasterTable
              where
                hub = @origin
                and dest = @dest
                and airline = @airline
                and (fl_1_num = @fl_num or fl_2_num = @fl_num)
                and fl_1_sch_dep >= @start_date
                and fl_1_sch_dep < @end_date
          ) b,
        ( select isnull(sum(max_pax),0) as total
              from MasterTable
              where
                origin = @origin
                and hub = @dest
                and airline = @airline
                and (fl_1_num = @fl_num or fl_2_num = @fl_num)
                and fl_1_sch_dep >= @start_date
                and fl_1_sch_dep < @end_date
          ) c
      )
      -- Return the result of the function
RETURN @result
END

3) Stored Procedure Create AOTP matches pairs of flights to get 2-leg itineraries
-- Author: <Sanja Avramovic>
-- Description: create table ##aotp
-- Matches pairs of flights to get 2-leg itineraries.
-- Based on Guermo's and Ashwin's code

CREATE PROCEDURE [dbo].[Create_aotp]
AS
BEGIN

SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.

CREATE TABLE ##aotp
(origin varchar(3) not null,
hub varchar(3),
dest varchar(3) not null,
airline varchar(10) not null,
Flight_1_number int not null,
Flight_1_Sch_Dep smalldatetime,
Flight_1_Sch_Arr smalldatetime,
Flight_1_Act_Dep smalldatetime,
Flight_1_Act_Arr smalldatetime,
Flight_1_Status bit not null, Flight_1_Div bit not null,
Flight_1_Avg_Seat int, Flight_1_LF float,
Flight_2_number int,
Flight_2_Sch_Dep smalldatetime, Flight_2_Sch_Arr smalldatetime,
Flight_2_Act_Dep smalldatetime, Flight_2_Act_Arr smalldatetime,
Flight_2_Status bit, Flight_2_Div bit,
Flight_2_Avg_Seat int, Flight_2_LF float,
Total_flights int)

-- Total_flights int

-- populate the table from #ontime_lf, using pairs of flights
INSERT INTO ##aotp
(origin, hub, dest, airline,
Flight_1_number,
Flight_1_Sch_Dep, Flight_1_Sch_Arr,
Flight_1_Act_Dep, Flight_1_Act_Arr,
Flight_1_Status, Flight_1_Div,
Flight_1_Avg_Seat, Flight_1_LF,
Flight_2_number,
Flight_2_Sch_Dep, Flight_2_Sch_Arr,
Flight_2_Act_Dep, Flight_2_Act_Arr,
Flight_2_Status, Flight_2_Div,
Flight_2_Avg_Seat, Flight_2_LF)

SELECT DISTINCT m.origin, m.dest, k.dest, m.airline,
m.flight_num,
m.sch_dep, m.sch_arr,
m.act_dep, m.act_arr,
m.cancelled, m.div_delay,
m.avgseat, m.lf,
k.flight_num,
k.sch_dep, k.sch_arr,
k.act_dep, k.act_arr,
k.cancelled, k.div_delay,
146

```
k.avgseat,k.lf
--count (*) -- Remove total_flights if we don't need it for connecting flights
from ##ontime_lf m, ##ontime_lf k
where -- Here I am setting the conditions
  k.airline = m.airline
  and k.origin = m.dest -- Origin of the second flight = Destination of the first flight.
  and k.dest <> m.origin -- Destination of the second flight cannot be origin of the first flight.
  and k.sch.dep > = m.sch_arr_1
  and k.sch_Dep <= m.sch_arr_2

  group by m.origin,m.dest,k.dest,m.airline,
  m.flight_num,m.sch_dep,m.sch_arr,m.act_dep,m.act_arr,m.cancelled,m.div_delay,m.avgseat,m.lf

order by m.origin,m.dest,m.airline,m.sch_dep,m.sch_arr,m.sch_dep,k.sch_arr,k.sch_arr;
```

**END**

```GO
```

4) Stored Procedure create the table DB1B, which contains passengers' statistics for origin/hub/destination tuples

```
USE [PaxDelay]
GO

/*******/ Object:  StoredProcedure [dbo].[Create_DB1B] Script Date: 3/15/2015 7:11:39 PM *******/
SET ANSI_NULLS ON
GO

SET QUOTED_IDENTIFIER ON
GO
-- ==============================================================
-- Author:  <Sanja Avramovic>
-- Create date: <10/15/2014>
-- Description: create the table DB1B, which contains pax statistics for origin/hub/destination tuples.
-- Based on Guerom's and Ashwin's code
-- ==============================================================
create PROCEDURE [dbo].[Create_DB1B]
 @start_date smalldatetime
 AS
 BEGIN

```
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

-- extract the year and quarter of the year from the date
declare @year as smallint
declare @quarter as tinyint

set @year = year(@start_date)
set @quarter = (month(@start_date) - 1) / 3 + 1

------------------------------------------CREATING #DB1B------------------------------------------

-- drop the table if it already exists
IF OBJECT_ID('dbo.db1b', 'U') IS NOT NULL
DROP TABLE dbo.db1b;

-- create the table
create table db1b(
    origin varchar(3) not null,
    dest varchar(3) not null,
    hub varchar(3),
    airport_group nvarchar(60),
    reporting_carrier nvarchar(3),
    total_pax int,
    pax_perday int,
    market_coupons int
);

-- populate the table from db1b market
insert into db1b (origin,dest,hub,airport_group,reporting_carrier,total_pax,pax_perday,market_coupons)
select b.origin, b.dest,
    case when len(b.airport_group) > 7
        then -- this is a dirty trick to get the hub! remember to do it better later
            substring(right(b.airport_group,
                len(b.airport_group)-charindex(';',b.airport_group)),
                    1,
            charindex(';',b.airport_group)-1)
            else '.'
        end as hub,
    b.airport_group,
    b.reporting_carrier,
    --b.year,
    --b.quarter,
    b.ten_percent_total_pax_quaterly *'10,
    b.avg_pax_perday_perairline_rounded,b.market_coupons
    as,b.avg_pax_perday_perairline
from
select a.market_coupons,a.origin,a.dest,
    a.airport_group,a.reporting_carrier,
    a.year,a.quarter,
    sum(a.passengers) as ten_percent_total_pax_quaterly,
    (sum(a.passengers)*'10/(3^30)) as avg_pax_perday_perairline_rounded
from db1b_market a
where
    a.market_coupons <=2
    and a.year = @year
    and a.quarter = @quarter
group by 
a.origin,a.dest,a.airport_group,a.reporting_carrier,a.year,a.quarter,a.market_coupons 

) b

where 
  b.avg_pax_perday_perairline_rounded > 0.5

order by origin, dest, hub, reporting_carrier;

-- create indexes
create index IDX_ORIGIN on db1b(origin);
create index IDX_DEST on db1b(DEST);
create index IDX_HUB on db1b(hub);
create index IDX_AIRLINES on db1b(reporting_carrier);
create index IDX_MARKET_COUPONS on db1b(MARKET_COUPONS);

END

GO

5) Create stored procedure Create Final which create the table Final, to store the final itinerary/PTD results

USE [PaxDelay]
GO

/****** Object:  StoredProcedure [dbo].[Create_Final]    Script Date: 3/15/2015 7:19:07 PM ******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ================================
-- Author:  <Sanja Avramovic>
-- Create date: <10/15/2014>
-- Description: create the table Final,
-- to store the final itinerary/PTD results
-- Based on Guermo's and Ashwin's code
-- ================================
create PROCEDURE [dbo].[Create_Final]
  -- Add the parameters for the stored procedure here
AS
BEGIN
  -- SET NOCOUNT ON added to prevent extra result sets from
  -- interfering with SELECT statements.
  SET NOCOUNT ON;

  -- drop the table if it already exists
  IF OBJECT_ID('dbo.Final', 'U') IS NOT NULL
    DROP TABLE dbo.Final;

  -- create the table
  create table Final(
    ORIGIN varchar (3) not null,
    DEST varchar(3) not null,
    HUB varchar(3),
    AIRLINE varchar(3) not null,
    MAX_PAX smallint,
    ...
6) Create stored procedure CreateMasterTable, which contains itinerary and PTD information.

USE [PaxDelay]
GO

/****** Object:  StoredProcedure [dbo].[Create_MasterTable]    Script Date: 3/15/2015 7:24:03 PM
******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO
-- ===============================================================
-- Author:  <Sanja Avramovic>
-- Create date: <10/15/2014>
-- Description: create table MasterTable,
-- which contains itinerary and PTD info.
-- Requires a date, and the number of days to run.
-- Based on Guerro's and Ashwin's code
-- ==============================================================
create PROCEDURE [dbo].[Create_MasterTable]
@start_date smalldatetime,
@num_days tinyint
AS
BEGIN
SET NOCOUNT ON;

-- set the end date based on the start date and the number of days to run
declare @end_date as smalldatetime
declare @days as int
set @end_date = dateadd(dd,@num_days,@start_date)
set @days = @num_days --DATEDIFF(dd, @start_date, @end_date) -- Set the number of days according to the data needed to be collected

-- a parameter specifying the number of dummy pax to add
declare @additional_pax as int
set @additional_pax = 0

-- drop the table if it already exists
IF OBJECT_ID('dbo.MasterTable', 'U') IS NOT NULL
DROP TABLE dbo.MasterTable;

-- create the master table
create table MasterTable('dbo.MasterTable'
    ORIGIN varchar(3) not null, DEST varchar(3) not null, HUB varchar(3), AIRLINE varchar(3) not null, MAX_PAX smallint, FL_1_NUM smallint, FL_1_SCH_DEP smalldatetime, FL_1_SCH_ARR smalldatetime, FL_1_ACT_DEP smalldatetime, FL_1_ACT_ARR smalldatetime, FL_1_CANCELLED bit, FL_1_DIVERTED bit, FL_1_AVG_SEAT smallint, FL_1_LF float, FL_2_NUM smallint, FL_2_SCH_DEP smalldatetime, FL_2_SCH_ARR smalldatetime, FL_2_ACT_DEP smalldatetime, FL_2_ACT_ARR smalldatetime, FL_2_CANCELLED bit, FL_2_DIVERTED bit, FL_2_AVG_SEAT smallint, FL_2_LF float, ORIGIN varchar(3), DEST varchar(3), HUB varchar(3), AIRLINE varchar(3), MAX_PAX smallint, FL_1_NUM smallint, FL_1_SCH_DEP smalldatetime, FL_1_SCH_ARR smalldatetime, FL_1_ACT_DEP smalldatetime, FL_1_ACT_ARR smalldatetime, FL_1_CANCELLED bit, FL_1_DIVERTED bit, FL_1_AVG_SEAT smallint, FL_1_LF float, FL_2_NUM smallint, FL_2_SCH_DEP smalldatetime, FL_2_SCH_ARR smalldatetime, FL_2_ACT_DEP smalldatetime, FL_2_ACT_ARR smalldatetime, FL_2_CANCELLED bit, FL_2_DIVERTED bit, FL_2_AVG_SEAT smallint, FL_2_LF float)

-- populate all connecting itineraries by merging AOTP and DB1B
insert into MasterTable
(ORIGIN,DEST,HUB,AIRLINE,MAX_PAX,FL_1_NUM,FL_1_SCH_DEP,FL_1_SCH_ARR,FL_1_ACT_DEP,FL_1_ACT_ARR,FL_1_CANCELLED,FL_1_DIVERTED,FL_1_AVG_SEAT,FL_1_LF,FL_2_NUM,FL_2_SCH_DEP,FL_2_SCH_ARR,FL_2_ACT_DEP,FL_2_ACT_ARR,FL_2_CANCELLED,FL_2_DIVERTED,FL_2_AVG_SEAT,FL_2_LF)
select
    d.origin, d.dest, d.hub, d.reporting_carrier, pax_perday as max_pax,
    d.Flight_1_number, d.Flight_1_Sch_Dep, d.Flight_1_Sch_Arr, d.Flight_1_Act_Dep,
    d.Flight_1_Act_Arr,
    d.Flight_1_Status, d.Flight_1_Div, d.Flight_1_Avg_Seat, d.Flight_1_LF,
    d.Flight_2_number, d.Flight_2_Sch_Dep, d.Flight_2_Sch_Arr, d.Flight_2_Act_Dep,
    d.Flight_2_Act_Arr,
    d.Flight_2_Status, d.Flight_2_Div, d.Flight_2_Avg_Seat, d.Flight_2_LF
from

    select
db1b.origin, db1b.hub, db1b.dest, db1b.reporting_carrier,
aotp.Flight_1_number,
aotp.Flight_1_Sch_Dep, aotp.Flight_1_Sch_Arr, aotp.Flight_1_Act_Dep,
aotp.Flight_1_Act_Arr,
aotp.Flight_1_Status, aotp.Flight_1_div,
aotp.Flight_1_Avg_Seat, aotp.Flight_1_lf,
aotp.Flight_2_number,
aotp.Flight_2_Sch_Dep, aotp.Flight_2_Sch_Arr, aotp.Flight_2_Act_Dep,
aotp.Flight_2_Act_Arr,
aotp.Flight_2_Status, aotp.Flight_2_div,
aotp.Flight_2_Avg_Seat, aotp.Flight_2_lf,
    db1b.pax_perday
from

    select origin,hub,dest,reporting_Carrier,pax_perday
from db1b
where market_coupons = 2
) db1b,
   #aotp aotp
where
db1b.origin = aotp.Origin
and db1b.dest = aotp.Dest
and db1b.hub = aotp.Hub
and db1b.reporting_carrier = aotp.airline
)
d;

---------------------------------------------------------------
--------------------
----------------------
DIRECT FLIGHTS--------------------------------------------------
--------------------
-- add all flights without a connection
insert into MasterTable
(
   ORIGIN, DEST, HUB, AIRLINE, MAX_PAX,
   FL_1_NUM, FL_1_SCH_DEP, FL_1_SCH.ARR, FL_1_ACT_DEP, FL_1_ACT_ARR,
   FL_1_CANCELLED, FL_1_DIVERTED, FL_1_AVG_SEAT, FL_1_LF,
   FL_2_NUM, FL_2_SCH_DEP, FL_2_SCH_ARR, FL_2_ACT_DEP, FL_2_ACT_ARR,
   FL_2_CANCELLED, FL_2_DIVERTED, FL_2_AVG_SEAT, FL_2_LF
)
select
db1b.origin, db1b.dest, null, db1b.reporting_carrier, (pax_perday/total_flights) + @additional_pax as max_pax,
aotp.Flight_number as Flight_1_number, aotp.sch_dep as Flight_1_Sch_Dep,
aotp.sch_arr as Flight_1_Sch_Arr,
aotp.act_dep as Flight_1_Act_Dep, aotp.act_arr as Flight_1_Act_Arr,
aotp.Flight_1_Status, aotp.Flight_1_div as Flight_1_div,
aotp.Flight_1_Avg_Seat, aotp.Flight_1_LF,
null, null, null, null, null, null, null, null,
from
(
   select *
   from db1b
   where market_coupons = 1
) db1b,
(
   select
      m.origin as Origin, m.dest, m.airline, m.flight_num as Flight_number,
      m.sch_dep as Sch_Dep, m.sch_arr as Sch_Arr, m.act_dep as Act_Dep, m.act_arr
as Act_Arr,
      m.cancelled as Flight_1_Status, m.div_delay as Flight_1_div,
      m.avgseat as Flight_1_Avg_Seat, m.lf as Flight_1_LF
   from #ontime_lf m
   group by m.origin, m.dest, m.airline, m.flight_num,
   m.sch_dep, m.sch_arr, m.act_dep, m.act_arr, m.avgseat, m.avgpax, m.lf,
   m.cancelled, m.div_delay
) aotp,
(
   --select m.origin, m.dest, m.airline,(count (*)/(3*30)) as total_flights
   select m.origin, m.dest, m.airline,(count (*)/@days) as total_flights
   from #ontime_lf m
   group by m.origin, m.dest, m.airline
) aotp_1
where
db1b.origin = aotp.Origin
and db1b.dest = aotp.Dest
and db1b.reporting_carrier = aotp.airline
and db1b.origin = aotp_1.origin
and db1b.dest = aotp_1.dest
and db1b.reporting_carrier = aotp_1.airline
and aotp_1.total_flights > 0;

-- create indexes on the table, once it has been populated
create index IDX_ORIGIN on MasterTable(origin);
create index IDX_HUB on MasterTable(hub);
create index IDX_DEST on MasterTable(Dest);
create index IDX_AIRLINES on MasterTable(airline);
create index IDX_FLNUM_1 on MasterTable(fl_1_num);
create index IDX_FL_1_SCH_DEP on MasterTable(fl_1_sch_dep);
create index IDX_FL_1_SCH_ARR on MasterTable(fl_1_sch_arr);
create index IDX_FLNUM_2 on MasterTable(fl_2_num);
create index IDX_FL_2_SCH_DEP on MasterTable(fl_2_sch_dep);
create index IDX_FL_2_SCH_ARR on MasterTable(fl_2_sch_arr);
create index IDX_T100_MAX_PAX on MasterTable(t100_max_pax);

END
GO

7) Stored Procedure Create Ontime_lf: Creates a temp table with on time statistics
and load factors

USE [PaxDelay]
GO
/****** ******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO
-- ==============================================================
-- Author: <Sanja Avramovic>
-- Create date: <10/15/2015>
-- Description: <create table ontime_lf>
-- Table contains seat/LF and flight status of
-- each flight.
-- Based on Guermo's and Ashwin's code
-- ==============================================================
create PROCEDURE [dbo].[Create_ontime_lf]
/*I PUT WN/AA as an airline FOR TESTING*/

     @start_date smalldatetime,
     @num_days tinyint

AS BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

-- determine the year and the time range based on start date.
declare @year as smallint
declare @end_date as smalldatetime
set @year = year(@start_date)
set @end_date = dateadd(dd,@num_days,@start_date)
-- create the table.
create table #ontime_lf
(
    origin varchar(3) not null, dest varchar(3) not null, airline varchar(6) not null,
    flight_num int not null,
    sch_dep smalldatetime, act_dep smalldatetime,
    sch_arr smalldatetime, act_arr smalldatetime,
    sch_arr_1 smalldatetime, sch_arr_2 smalldatetime,
    cancelled bit not null, div_delay bit not null,
    avgseat int, avgpax int, lf float
)

-- use ontime and load factor data to populate the table.
insert into #ontime_lf
(
    origin, dest, airline, flight_num,
    sch_dep, act_dep, sch_arr, act_arr,
    sch_arr_1, sch_arr_2,
    cancelled, div_delay,
    avgseat, avgpax, lf
)

    select a.origin, a.dest, a.airline,
    a.flight_num,
    a.sch_dep, dateadd (minute, a.dep_delay, a.sch_dep),
    a.sch_arr, dateadd (minute, a.arr_delay, a.sch_arr),
    dateadd (minute, 30, a.sch_arr), dateadd (minute, 180, a.sch_arr),
    a.cancelled, a.div_delay,
    b.avgseat, b.avgpax, b.avgpax / (1.0 * b.avgseat)

    from ON_TIME a, LOAD_FACTORS b

    where
    a.AIRLINE = 'AA' and b.CARRIER = 'AA' and
    a.sch_dep >= @start_date and a.sch_dep < @end_date

    and a.origin = 'LGA'
    and a.origin = b.origin
    and a.dest = b.dest
    and a.airline = b.carrier
    and international = 0
    and b.year = @year
    and b.month = month(a.sch_dep)
    and b.AVGSEAT != 0

    order by a.origin, a.dest, a.airline, a.flight_num, a.sch_dep, a.sch_arr

-- build indexes.
create index IDX_ORIGIN on #ontime_lf(origin);
create index IDX_DEST on #ontime_lf(DIST);
create index IDX_AIRLINES on #ontime_lf(airline);
create index IDX_FLNUM on #ontime_lf(flight_num)
create index IDX_SCH_DEP on #ontime_lf(sch_dep);
create index IDX_SCH_ARR on #ontime_lf(sch_arr);
8) Stored procedure

USE [PaxDelay]
GO

/****** Object:  StoredProcedure [dbo].[Create_OptionsBase]    Script Date: 5/27/2015
12:29:59 AM ******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ==============================================================
-- Author:  <Sanja Avramovic>
-- Create date: <3/6/2015>
-- Description: <create table options table>
-- Create a table of alternate flight options
-- for each itinerary.
-- Based on Guermo's and Ashwin’s code
-- ==============================================================
CREATE PROCEDURE [dbo].[Create_OptionsBase]
@startdate smalldatetime,
@numdays int,
@max_wait int = 600,
@if_adj float = 0
AS
BEGIN
SET NOCOUNT ON added to prevent extra result sets from
interfering with SELECT statements.
SET NOCOUNT ON;

-- check if the table exists already, and drop it
if OBJECT_ID('tempdb..##options_base') is not null drop table ##options_base;

-- create the table, before populating it
create table ##options_base
(   Itin_status smallint not null,origin varchar(3) not null,dest varchar(3)
not null,Original_Hub varchar(3) null,New_Hub varchar(3),
Total_Pax_on_Flight int,
Remaining_pax int,
Original_Airline varchar(10) not null,New_Airline varchar(10),
Original_Flight_Status bit not null,
Original_Flight_1 smallint not null,Original_Sch_Dep smalldatetime not
null,Original_Flight_1_Sch_Arr smalldatetime,Original_Flight_1_Act_Arr smalldatetime,
Original_Flight_2 smallint,Original_Flight_2_sch_dep
smalldatetime,Original_Flight_2_sch_arr smalldatetime,
Alternate_Flight_1 smallint,Alternate_Flight_1_Sch_Dep
smalldatetime,Alternate_Flight_1_Sch_Arr smalldatetime, Alternate_Flight_1_LF
float,Alternate_Flight_1_TotalSeats smallint,Alternate_Flight_1_AvaSeats
smallint,Initial_Alternate_Flight_1_AvaSeats smallint, Alternate_Flight_1_Act_Arr
smalldatetime,
Alternate_Flight_2 smallint,Alternate_Flight_2_Sch_Dep
smalldatetime,Alternate_Flight_2_Sch_Arr smalldatetime, Alternate_Flight_2_LF
float,Alternate_Flight_2_TotalSeats smallint,Alternate_Flight_2_AvaSeats
smallint

...
smallint, Initial_Alternate_Flight_2_AvaSeats, smallint, Alternate_Flight_2_Act_Arr
smalldatetime,
    Total_Rebooked_Pax, int, PTD_in_mins, int, unit_trip_delay, int,
unit_trip_time, int, unit_trip_time, int,
alt_priority, smallint, -- the priority level of this alternative option
    (for example, alts on the same airline have better priority than those that don't)
priority_metric, smallint, -- reordering of alt_priority, used for ordering results
    alt1dep_metric, int, -- ordering metric based on the departure time of
alternate flight 1
    alt2dep_metric, int, -- ordering metric based on the departure time of
alternate flight 1
)

-- populate the table for each class of flight disruption
EXEC [dbo].[Populate_Cancellations] @startdate, @numdays, @max_wait, @lf_adj ;
EXEC [dbo].[Populate_Diverted] @startdate, @numdays, @max_wait, @lf_adj ;
EXEC [dbo].[Populate_Delayed] @startdate, @numdays, @max_wait, @lf_adj ;
EXEC [dbo].[Populate_Cancellations2] @startdate, @numdays, @max_wait, @lf_adj ;

-- update the trip delays per pax per alt itinerary
update ##options_base
set unit_trip_delay =
    datediff(mi, alternate_flight_2_sch_arr, alternate_flight_2_act_arr) - 15
    where alternate_flight_2_act_arr is not null
    and alternate_flight_2_act_arr > alternate_flight_2_sch_arr
update ##options_base
set unit_trip_delay =
    datediff(mi, alternate_flight_1_sch_arr, alternate_flight_2_act_arr) - 15
    where alternate_flight_2_act_arr is null
    and alternate_flight_2_act_arr > alternate_flight_1_sch_arr
update ##options_base
set unit_trip_time =
    datediff(mi, alternate_flight_1_sch_dep, alternate_flight_2_sch_arr)
    where alternate_flight_1_sch_dep is not null
    and alternate_flight_2_sch_arr is not null
    and alternate_flight_1_sch_dep < alternate_flight_2_sch_arr
update ##options_base
set unit_trip_time =
    datediff(mi, alternate_flight_1_sch_dep, alternate_flight_2_act_arr)
    where alternate_flight_1_sch_dep is not null
    and alternate_flight_2_act_arr is null
    and alternate_flight_1_sch_dep < alternate_flight_2_act_arr
update ##options_base
set unit_trip_time =
    datediff(mi, alternate_flight_1_sch_dep, alternate_flight_1_sch_arr)
    where alternate_flight_1_sch_dep is not null
    and alternate_flight_2_sch_arr is not null
    and alternate_flight_1_sch_dep < alternate_flight_2_sch_arr
update ##options_base
set unit_sch_time =
    datediff(mi, alternate_flight_1_sch_dep, alternate_flight_2_sch_arr)
    where alternate_flight_1_sch_dep is not null
    and alternate_flight_2_sch_arr is not null
    and alternate_flight_1_sch_dep < alternate_flight_2_sch_arr

and alternate_flight_1_sch_dep < alternate_flight_1_sch_arr

update ##options_base
set unit_sch_time =
datediff(mi, alternate_flight_2_sch_dep, alternate_flight_2_sch_arr)
where alternate_flight_1_sch_dep is null
and alternate_flight_2_sch_arr is not null
and alternate_flight_2_sch_dep < alternate_flight_2_sch_arr

/* stranded pax was initially tracked in its own table, before it was merged into
The main results:
-- check if the table exists already, and drop it
if OBJECT_ID('tempdb..##Stranded_pax') is not null drop table ##Stranded_pax;
-- create a table for keeping track of the number of stranded pax
select year, month, day, sum(Stranded_pax) as Stranded_pax, sum(Stranded_pax) as
Remaining_pax into ##Stranded_pax
from
( select year(Original_Sch_Dep) as year, month(Original_Sch_Dep) as month,
day(Original_Sch_Dep) as day, Stranded_pax
from
( select Original_Flight_1, Original_Flight_2, Original_Sch_Dep,
Original_Sch_Arr, max(Remaining_pax) as Stranded_pax
from ##options_temp
group by Original_Flight_1, Original_Flight_2, Original_Sch_Dep,
Original_Sch_Arr
) a
) a
group by year, month, day;
*/
create index IDX_ITIN_STATUS on ##options_base(itin_status);
create index IDX_ORIGIN on ##options_base(origin);
create index IDX_OG_HUB on ##options_base(original_hub);
create index IDX_NW_HUB on ##options_base(new_hub);
cREATE INDEX IDX_DEST on ##options_base(DEST);
create index IDX_AIRLINES on ##options_base(original_airline);
create index IDX_OG_FL_1 on ##options_base(original_flight_1);
create index IDX_OG_FL_1_SCH_DEP on ##options_base(original_Sch_Dep);
create index IDX_OG_FL_1_SCH_ARR on ##options_base(original_Flight_1_Sch_Arr);
create index IDX_OG_FL_1_ACT_ARR on ##options_base(original_Flight_1_Act_Arr);
create index IDX_OG_FL_2 on ##options_base(original_flight_2);
create index IDX_OG_FL_2_SCH_DEP on ##options_base(original_Flight_2_sch_dep);
create index IDX_OG_FL_2_SCH_ARR on ##options_base(original_Sch_Arr);
create index IDX_OG_FL_2_ACT_ARR on ##options_base(original_Flight_2_Act_Arr);
create index IDX_ALT_FL_1 on ##options_base(alternate_flight_1);
create index IDX_ALT_FL_1_SCH_DEP on ##options_base(Alternate_Flight_1_Sch_Dep);
create index IDX_ALT_FL_1_ACT_ARR on ##options_base(Alternate_Flight_1_Act_Arr);
create index IDX_ALT_FL_2 on ##options_base(alternate_flight_2);
create index IDX_ALT_FL_2_SCH_DEP on ##options_base(Alternate_Flight_2_Sch_Dep);
create index IDX_ALT_FL_2_SCH_ARR on ##options_base(Alternate_Sch_Arr);
create index IDX_ALT_FL_2_ACT_ARR on ##options_base(Alternate_Flight_2_Act_Arr);
cREATE INDEX IDX_ALT_PRIORITY on ##options_base(alt_priority);

END
GO

9) Stored procedure
USE [PaxDelay]
GO
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO
-- -------------------------------------------------------------
-- Author:         <Sanja Avramovic>
-- Create date:   <10/23/2014>
-- Description:   <create table options table>
-- Make a working copy of the options base table
-- in ##options_temp.
-- Based on Guermo’s and Ashwin’s code
-- -------------------------------------------------------------
CREATE PROCEDURE [dbo].[Create_OptionsTemp] /*
@startdate smalldatetime,
@numdays int,
@max_wait int = 600,
@lf_adj float = 0
*/
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.
SET NOCOUNT ON;

-- check if the table exists already, and drop it
if OBJECT_ID('tempdb..##options_temp') is not null drop table ##options_temp;
select * into ##options_temp from ##options_base;

-- create the table, before populating it
create table ##options_temp
(  Itin_status smallint not null,origin varchar(3) not null,dest varchar(3) not null,Original_Hub  varchar(3) null,New_Hub  varchar(3),
   Total_Pax_on_Flight  int,
   Remaining_pax int,
   Original_Airline  varchar(10) not null,New_Airline  varchar(10),
   Original_Flight_Status  bit not null,Original_Flight_1  smallint not null,Original_Sch_Dep  smalldatetime not null,Original_Flight_1_Sch_Arr smalldatetime,Original_Flight_1_Act_Arr smalldatetime,
   Original_Flight_2  smallint,Original_Flight_2_sch_dep smalldatetime,Original_Flight_2_Sch_Arr smalldatetime,Alternate_Flight_1  smallint,Alternate_Flight_1_Sch_Dep smalldatetime,Alternate_Flight_1_Sch_Arr smalldatetime,Alternate_Flight_1_LF float,Alternate_Flight_1_TotalSeats smallint,Alternate_Flight_1_AvaSeats smallint,Initial_Alternate_Flight_1_AvaSeats smallint,Alternate_Flight_1_Act_Arr smalldatetime,
   Alternate_Flight_2  smallint,Alternate_Flight_2_sch_dep smalldatetime,Alternate_Flight_2_Sch_Arr smalldatetime,Alternate_Flight_2_LF float,Alternate_Flight_2_TotalSeats smallint,Alternate_Flight_2_AvaSeats smallint,Initial_Alternate_Flight_2_AvaSeats smallint,Alternate_Flight_2_Act_Arr smalldatetime,
   Total_Rebooked_Pax int,PTD_in_mins int,unit_trip_delay int,unit_trip_time int,unit_sch_time int,
   alt_priority smallint, -- the priority level of this alternative option (for example, alts on the same airline have better priority than those that don’t)
   early bit default 0 -- if the alternative is used for early rebooking rather than normal rebooking (flights up to 30 min after the flight still count as early rebooking, because advanced warning is needed for them)
)
-- populate the table for each class of flight disruption
EXEC [dbo].[Populate_Cancellations] @startdate, @numdays, @max_wait, @lf_adj;
EXEC [dbo].[Populate_Diverted] @startdate, @numdays, @max_wait, @lf_adj;
EXEC [dbo].[Populate_Delayed] @startdate, @numdays, @max_wait, @lf_adj;
EXEC [dbo].[Populate_Cancellations2] @startdate, @numdays, @max_wait, @lf_adj;

-- update the trip delays per pax per alt itinerary
update #options_temp
set unit_trip_delay = datediff(mi,alternate_flight_2_sch_arr,alternate_flight_2_act_arr)-15
where alternate_flight_2_act_arr is not null and
datediff(mi,alternate_flight_2_sch_arr,alternate_flight_2_act_arr)>15;
update #options_temp
set unit_trip_delay = datediff(mi,alternate_flight_1_sch_arr,alternate_flight_1_act_arr)-15
where alternate_flight_2_act_arr is null and
datediff(mi,alternate_flight_1_sch_arr,alternate_flight_1_act_arr)>15;

update #options_temp
set unit_trip_time = datediff(mi,alternate_flight_1_sch_dep,alternate_flight_2_act_arr)
where alternate_flight_1_sch_dep is not null
and alternate_flight_2_act_arr is not null
and alternate_flight_1_sch_dep<alternate_flight_2_act_arr;
update #options_temp
set unit_trip_time = datediff(mi,alternate_flight_1_sch_dep,alternate_flight_1_act_arr)
where alternate_flight_1_sch_dep is not null
and alternate_flight_2_act_arr is null
and alternate_flight_1_sch_dep<alternate_flight_1_act_arr;

update #options_temp
set unit_trip_time = datediff(mi,alternate_flight_2_sch_dep,alternate_flight_2_act_arr)
where alternate_flight_1_sch_dep is null
and alternate_flight_2_act_arr is not null
and alternate_flight_2_sch_dep<alternate_flight_2_act_arr;
update #options_temp
set unit_trip_time = datediff(mi,alternate_flight_1_sch_dep,alternate_flight_2_sch_arr)
where alternate_flight_1_sch_dep is not null
and alternate_flight_2_sch_arr is not null
and alternate_flight_1_sch_dep<alternate_flight_2_sch_arr;
update #options_temp
set unit_trip_time = datediff(mi,alternate_flight_1_sch_dep,alternate_flight_1_sch_arr)
where alternate_flight_1_sch_dep is not null
and alternate_flight_2_sch_arr is null
and alternate_flight_1_sch_dep<alternate_flight_1_sch_arr;
update #options_temp
set unit_trip_time = datediff(mi,alternate_flight_2_sch_dep,alternate_flight_2_sch_arr)
where alternate_flight_1_sch_dep is null
and alternate_flight_2_sch_arr is not null
and alternate_flight_2_sch_dep<alternate_flight_2_sch_arr;

create index IDX_ITIN_STATUS on #options_temp(itin_status);
create index IDX_ORIGIN on #options_temp(origin);
create index IDX_OG_HUB on #options_temp(original_hub);
create index IDX_NW_HUB on #options_temp(new_hub);
create index IDX_DEST on #options_temp(DEST);
create index IDX_AIRLINES on #options_temp(original_airline);
create index IDX_OG_FL_1 on #options_temp(original_flight_1)
create index IDX_OG_FL_1_SCH_DEP on ##options_temp(Original_Sch_Dep);
create index IDX_OG_FL_1_SCH_ARR on ##options_temp(Original_Flight_1_Sch_Arr);
create index IDX_OG_FL_1_ACT_ARR on ##options_temp(Original_Flight_1_Act_Arr);
create index IDX_OG_FL_2 on ##options_temp(Original_flight_2);
create index IDX_OG_FL_2_SCH_DEP on ##options_temp(Original_Flight_2_Sch_Dep);
create index IDX_OG_FL_2_SCH_ARR on ##options_temp(Original_Sch_Arr);
create index IDX_ALT_FL_1 on ##options_temp(Alternate_flight_1);
create index IDX_ALT_FL_1_SCH_DEP on ##options_temp(Alternate_Flight_1_Sch_Dep);
create index IDX_ALT_FL_1_ACT_ARR on ##options_temp(Alternate_Flight_1_Act_Arr);
create index IDX_ALT_FL_2 on ##options_temp(Alternate_flight_2);
create index IDX_ALT_FL_2_SCH_DEP on ##options_temp(Alternate_Flight_2_Sch_Dep);
create index IDX_ALT_FL_2_SCH_ARR on ##options_temp(Alternate_Sch_Arr);
create index IDX_ALT_FL_2_ACT_ARR on ##options_temp(Alternate_Flight_2_Act_Arr);
create index IDX_ALT_PRIORITY on ##options_temp(alt_priority);
END

GO

10) Stored procedure

USE [PaxDelay]
GO
******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ---------------------------------------------
-- Author:    <Sanja Avramovic>
-- Create date: <11/18/2014>
-- Description: <create table OriginDest>
-- Creates a table with all possible origin-dest pairs
-- ---------------------------------------------
CREATE PROCEDURE [dbo].[Create_OriginDest]
AS
BEGIN
    -- SET NOCOUNT ON added to prevent extra result sets from
    -- interfering with SELECT statements.
    SET NOCOUNT ON;

    -- make sure the table does not already exist
    IF OBJECT_ID('dbo.OriginDest', 'U') IS NOT NULL
    DROP TABLE dbo.OriginDest;

    -- create the table (origin-dest pair, plus an id used as a key for each row)
    CREATE TABLE OriginDest(
        id int not null,
        origin varchar(3) not null,
        dest varchar(3) not null
    );

    -- create a list of all airports that are used, using origin/dest/hub info from DB1B
    SELECT distinct x.airport into #airports from
(select distinct a.origin as airport from dbo.db1b a
union
select distinct a.dest as airport from dbo.db1b a
union
select distinct a.hub as airport from dbo.db1b a)
x
where x.airport is not null;

-- populate the table, by generating a list of all airport-airport pairs
insert into OriginDest (id,origin,dest)
select row_number() over (order by a.airport,b.airport) as id, a.airport as origin,
b.airport as dest
from #airports a, #airports b;

-- clean up
drop table #airports;

-- create indexes for the table
create unique index IDX_ID on OriginDest(id);
create index IDX_ORIGIN on OriginDest(origin);
create index IDX_DEST on OriginDest(dest);

END
GO

11) Stored procedure

USE [PaxDelay]
GO

SET ANSI_NULLS ON
GO

SET QUOTED_IDENTIFIER ON
GO

-- ****************************
-- Author: <Sanja Avramovic>
-- Create date: <10/23/2014>
-- Description: <populate table options table - cancellations>
-- Based on Guermo’s and Ashwin’s code
-- TESTING NOTE: the range of flights to select was temporarily changed from a @max_wait minutes
-- range to a +/-1 day range. Change commented lines to change back.
-- ****************************
CREATE PROCEDURE [dbo].[Populate_Cancellations]
@startdate smalldatetime,
@numdays int,
@max_wait int = 600,
@lf_adj float = 0
-- @early_max_wait = 600

AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;
declare @enddate smalldatetime
declare @wait_afterCancelled int
set @wait_afterCancelled = 30
set @enddate = dateadd(dd, @numdays, @startdate)
declare @lf_mult float;
set @lf_mult = 1+@lf_adj;

--insert into #options_temp
insert into #options_base(
Itin_status,
origin.dest,Original_Hub,New_Hub,
Total_Pax_on_Flight,
Remaining_pax,
Original_Airline,New_Airline,
Original_Flight_Status,Original_Flight_1,Original_Sch_Dep,Original_Flight_1_Sch_Arr,Original_Flight_1_Act_Arr,
Original_Flight_2,Original_Flight_2_sch_dep,Original_Sch_Arr,
Alternate_Flight_1,Alternate_Flight_1_Sch_Dep,Alternate_Flight_1_Sch_Arr,Alternate_Flight_1_lf,Alternate_Flight_1_TotalSeats,Alternate_Flight_1_AvaSeats,Initial_Alternate_Flight_1_AvaSeats,Alternate_Flight_1_Act_Arr,
Alternate_Flight_2,Alternate_Flight_2_Sch_Dep,Alternate_Flight_2_Sch_Arr,Alternate_Flight_2_lf,Alternate_Flight_2_TotalSeats,Alternate_Flight_2_AvaSeats,Initial_Alternate_Flight_2_AvaSeats,Alternate_Flight_2_Act_Arr,
Total_Rebooked_Pax,PTD_in_mins,unit_trip_delay,unit_trip_time,unit_sch_time,alt_priority,priority_metric,alt1dep_metric,alt2dep_metric)

select 1,
/*
1st - rebook on flights from the same airline (and it's subsidiaries) on that day
*/
1
2nd rebook on flights from other airlines on that day
3rd rebook on flights for the same airline (and its subsidiaries) on the next day
4th rebook on flights from other airlines on the next day.
5th passengers not rebooked after any of these, would get assigned a 900 minute delay */

\[
\text{case when } b.f1_1\_sch\_dep \text{ is null then } -1 \quad \text{else}
\]

\[
\text{case when } \text{datediff}(\text{mi}, a.f1_1\_sch\_dep, b.f1_1\_sch\_dep) <
\]

\[
@\text{wait\_after\_cancelled} \text{ then}
\]

\[
(\text{case when } \text{day}(a.f1_1\_sch\_dep) \neq \text{day}(b.f1_1\_sch\_dep) \text{ and }
\]

\[
\text{datediff}(\text{mi},a.f1_1\_sch\_dep,b.f1_1\_sch\_dep)\neq 0
\]

\[
\text{then (case when } a.\text{airline}=b.\text{airline then 1 else 2)
\]

\[
\text{end)-- 1. early, day before, same airline, 2. early, not same day, different airline}
\]

\[
\text{else (case when } a.\text{airline}=b.\text{airline then 3 else 4)
\]

\[
\text{end) end)-- 3. early, same day, same airline, 4. early, same day, different airline}
\]

\[
\text{else (case when } a.\text{airline}\neq b.\text{airline then 5 else 6 end)-- 5. normal, same day, same air, 6. normal, same day, different air.}
\]

\[
\text{else (case when } a.\text{airline}\neq b.\text{airline then 7 else 8)
\]

\[
\text{end end) --7. normal, next day, same air, 8. normal, next day, different air}
\]

\[
\text{end as alt\_priority,
\]

/* ordering metrics */

\[
\text{case when } b.f1_1\_sch\_dep \text{ is null then } -1 \quad \text{else}
\]

\[
\text{case when } \text{datediff}(\text{mi}, a.f1_1\_sch\_dep, b.f1_1\_sch\_dep) <
\]

\[
@\text{wait\_after\_cancelled} \text{ then}
\]

\[
(\text{case when } \text{day}(a.f1_1\_sch\_dep) \neq \text{day}(b.f1_1\_sch\_dep) \text{ and }
\]

\[
\text{datediff}(\text{mi},a.f1_1\_sch\_dep,b.f1_1\_sch\_dep)\neq 0
\]

\[
\text{then (case when } a.\text{airline}=b.\text{airline then 3 else 4)
\]

\[
\text{end)-- 1. early, day before, same airline, 2. early, not same day, different airline}
\]

\[
\text{else (case when } a.\text{airline}=b.\text{airline then 1 else 2)
\]

\[
\text{end) end)-- 3. early, same day, same airline, 4. early, same day, different airline}
\]

\[
\text{else (case when } a.\text{airline}\neq b.\text{airline then 5 else 6 end)-- 5. normal, same day, same air, 6. normal, same day, different air.}
\]

\[
\text{else (case when } a.\text{airline}\neq b.\text{airline then 7 else 8)
\]

\[
\text{end end) --7. normal, next day, same air, 8. normal, next day, different air}
\]

\[
\text{end as priority\_metric,
\]

\[
\text{case when } b.f1_1\_sch\_dep \text{ is null then null else}
\]

\[
\text{case when } \text{datediff}(\text{mi}, a.f1_1\_sch\_dep, b.f1_1\_sch\_dep) <
\]

\[
@\text{wait\_after\_cancelled} \text{ then}
\]

\[
-\text{datediff}(\text{mi},a.f1_1\_sch\_dep,b.f1_1\_sch\_dep)
\]

\[
\text{else } \text{datediff}(\text{mi},a.f1_1\_sch\_dep,b.f1_1\_sch\_dep)+(5\text{^24}\text{^60})
\]

\[
\text{end as alt1\_dep\_metric,
\]

\[
\text{case when } b.f1_2\_sch\_dep \text{ is null then null else}
\]

\[
\text{case when } \text{datediff}(\text{mi}, a.f1_1\_sch\_dep, b.f1_1\_sch\_dep) <
\]

\[
@\text{wait\_after\_cancelled} \text{ then}
\]

\[
-\text{datediff}(\text{mi},a.f1_2\_sch\_dep,b.f1_2\_sch\_dep)
\]

\[
\text{else } \text{datediff}(\text{mi},a.f1_2\_sch\_dep,b.f1_2\_sch\_dep)+(5\text{^24}\text{^60})
\]
end
end as alt2dep_metric

/*
   case when a.airline=b.airline and day(a.fl_1_sch_dep) = day (b.fl_1_sch_dep) then 1
   when a.airline != b.airline and day(a.fl_1_sch_dep) = day (b.fl_1_sch_dep) then 2
   when a.airline=b.airline
   then 3
   else 4
   end as alt_priority,
   case when datediff(mi, a.fl_1_sch_dep, b.fl_1_sch_dep) < @wait_after_cancelled
   then 1 else 0 end as early
*/
from
(select *
from Final
-- Set of cancelled flights
where
fl_1_cancelled = 1
-- ADD START AND END DATE with sch dep and sch arr times comparison
--and fl_1_diverted = 0
) a
left join
-- Set of possible alternate flights (candidate) where pax can be rebooked
(select *
from Final
-- ADD START AND END DATE with sch dep and sch arr times comparison
fl_1_cancelled = 0
and
((((fl_2_cancelled = 0) or (fl_2_cancelled is null))
and (dateadd(minute,30,fl_1_act_arr) <= fl_2_sch_dep) or (fl_2_sch_dep is null)) -- Actual Arrival of the first flight should be less than Sch Dep of second flight - 30 minutes(connection time)
and (fl_1_avg_seat*(1-fl_1_lf)) > 0 -- Sets the condition to check if the flight is full or not
and (((fl_2_avg_seat*(1-fl_2_lf)) > 0) or (fl_2_avg_seat is null))
and fl_1_diverted = 0
and (fl_1_diverted = 0 or fl_2_diverted is null)
)b
on
a.fl_1_SCH_DEP >= @startdate and a.fl_1_SCH_DEP < @enddate and
a.origin = b.origin
and a.dest = b.dest
--and a.airline = b.airline
--and b.fl_1_sch_dep >=

dateadd(minute,@wait_after_cancelled,a.fl_1_sch_dep) -- need to wait 30 minutes before the next flight

/*if the time range is 24 hours:*/
--and datediff(mi, a.fl_1_sch_dep, b.fl_1_sch_dep) / @max_wait = 0 -- the flight must be within @max_wait minutes, before or after the original, to be considered
/*if the time range is the whole prev. day*/

and
abs(datediff(dd,convert(smalldatetime,convert(date,a.fl_1_sch_dep)),b.fl_1_sch_dep))<=1
--and datediff(mi, a.fl_1_sch_dep, b.fl_1_sch_dep) <= @max_wait and

datediff(mi, a.fl_1_sch_dep, b.fl_1_sch_dep) > @early_max_wait
--and b.fl_1_sch_dep <=dateadd(minute,@max_wait,a.fl_1_sch_dep)
--and year (a.fl_1_sch_dep) = year (b.fl_1_sch_dep)
--and month(a.fl_1_sch_dep) = month(b.fl_1_sch_dep)
--and day(a.fl_1_sch_dep) = day (b.fl_1_sch_dep)
--and b.fl_1_sch_dep
<=dateadd(minute,@max_waitForAlternateFlt+@wait_after_cancelled,a.fl_1_sch_dep)
--and b.fl_2_sch_arr >=a.fl_2_sch_arr
order by a.fl_1_sch_dep, a.fl_1_num, a.fl_2_sch_arr, a.fl_2_num, b.fl_1_sch_dep, b.fl_2_sch_arr

END

GO

12) Stored procedure

USE [PaxDelay]
GO

12:34:52 AM */
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ==============================================-- Author: <Sanja Avramovic>-- Create date: <10/23/2014>-- Description: <populate table options table - cancellations on the second flight>-- Based on Guermo's and Ashwin's code-- TESTING NOTE: the range of flights to select was temporarily changed from a @max_wait minutes range to a +1 day range. Change commented lines to change back.-- ==============================================CREATE PROCEDURE [dbo].[Populate_Cancellations2]
    @startdate smalldatetime,
    @numdays int,
    @max_wait int = 600,
    @lf_adj float = 0
AS
BEGIN
    -- SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.
    SET NOCOUNT ON;
    declare @enddate smalldatetime
    declare @waitAfterCancel int
    set @waitAfterCancel = 30
    set @enddate = dateadd(dd, @numdays, @startdate)

    declare @lf_mult float;
    set @lf_mult = 1+@lf_adj;

    --insert into #options_temp
    insert into #options_base
        (Itin_Status, origin, dest, Original_Hub, New_Hub, Total_Pax_on_Flight, Remaining_pax, Original_Airline, New_Airline, Original_Flight_Status, Original_Flight_1, Original_Sch_Dep, Original_Flight_1_Sch_Arr, Original_Flight_2, Original_Flight_2_sch_dep, Original_Sch_Arr, Alternate_Flight_1, Alternate_Flight_1_Sch_Dep, Alternate_Flight_1_Sch_Arr, Alternate_Flight_2, Alternate_Flight_2_Sch_Dep, Alternate_Flight_2_Sch_Arr)
SELECT 2, -- Flight 2 is cancelled. So itin_status is auto cancelled
    a.origin, a.dest,
    a.hub as Hub, null,
    a.max_pax as Total_Pax_on_Flight,
    a.max_pax as Remaining_Pax,
    a.airline as Original_Airline, b.airline as New_Airline,
    a.f1_1_diverted as Original_Flight_Status, a.f1_1_num as
    Alternate_Flight_1, a.f1_1_sch_dep as Original_Sch_Dep, null, a.f1_1_act_arr as
    Original_Flight_1_Act_Arr,
    a.f1_2_num as Original_Flight_2, a.f1_2_sch_dep as
    Original_Flight_2_Sch_Dep, null, a.f1_2_sch_arr as
    Original_Flight_2_Sch_Arr, a.f1_2_lf as
    Alternate_Flight_2_LF, a.f1_2_avg_seat * @lf_mult as
    Alternate_Flight_2_TotalSeats, (1 - a.f1_2_lf) * b.f1_2_avg_seat * @lf_mult as
    Alternate_Flight_2_AvaSeats,
    Initial_Alternate_Flight_2_AvaSeats,
    b.f1_2_act_arr as Alternate_Flight_2_Act_Arr,
    0 as Total_Rebooked_Pax,
    0 as PTD_in_mins,
    0 as unit_trip_delay,
    0 as unit_trip_time,
    0 as unit_sch_time,
    /*
     1st - rebook on flights from the same airline (and it's
     subsidiaries) on that day
     2nd rebook on flights from other airlines on that day
     3rd rebook on flights for the same airline (and it's
     subsidiaries) on the next day
     4th rebook on flights from other airlines on the next day.
     5th passengers not rebooked after any of these, would get
     assigned a 900 minute delay */
    case when b.f1_2_sch_dep is null then -1 else
        case when
        datediff(mi, a.f1_2_sch_dep, b.f1_2_sch_dep) < @waitAfterCancel
        then
            (case when day(a.f1_2_sch_dep) != day(b.f1_2_sch_dep) and
            datediff(mi, a.f1_2_sch_dep, b.f1_2_sch_dep) > 0
            then (case when a.airline = b.airline then 1 else 2
            end) -- 1. early, day before, same airline, 2. early, not same day, different airline
            else (case when a.airline = b.airline then 3 else 4
            end) end) -- 3. early, same day, same airline, 4. early, same day, different airline
        else
            0
        end
    end

    /*
    1_LF, Alternate_Flight_1_TotalSeats, Alternate_Flight_1_AvaSeats, Initial_Alternate_Flight_1_AvaSeats,
    Alternate_Flight_1_Act_Arr,
    Alternate_Flight_2, Alternate_Flight_2_Sch_Dep, Alternate_Flight_2_Sch_Arr, Alternate_Flight_2_LF,
    Alternate_Flight_2_TotalSeats, Alternate_Flight_2_AvaSeats, Initial_Alternate_Flight_2_AvaSeats,
    Alternate_Flight_2_Act_Arr,
    Total_Rebooked_Pax, PTD_in_mins, unit_trip_delay, unit_trip_time, unit_sch_time,
    alt_priority, priority_metric, alt1dep_metric, alt2dep_metric
    */
(case when day(a.fl_2_sch_dep) = day (b.fl_2_sch_dep) then
(case when a.airline=b.airline then 5 else 6 end) -- 5. normal, same day, same air, 6. normal, same day, different air.
else (case when a.airline=b.airline then 7 else 8 end) end) -- 7. normal, next day, same air, 8. normal, next day, different air
end end as alt_priority,
*/
case when a.airline=b.airline and day(a.fl_2_sch_dep) = day
(b.fl_2_sch_dep) then 1
when a.airline != b.airline and day(a.fl_2_sch_dep) = day
(b.fl_2_sch_dep) then 2
else
(case when a.airline=b.airline then 3
else 4
end end as alt_priority,
*/
/* ordering metrics */
case when b.fl_2_sch_dep is null then -1 else
case when datediff(mi, a.fl_2_sch_dep, b.fl_2_sch_dep) < @waitAfterCancel
then
(case when day(a.fl_2_sch_dep) != day (b.fl_2_sch_dep) and
datediff(mi,a.fl_2_sch_dep,b.fl_2_sch_dep)=0
then (case when a.airline=b.airline then 3 else 4
end)-- 1. early, day before, same airline, 2. early, not same day, different airline
else (case when a.airline=b.airline then 1 else 2
end) end)-- 3. early, same day, same airline, 4. early, same day, different airline
else (case when day(a.fl_2_sch_dep) = day (b.fl_2_sch_dep) then
(case when a.airline=b.airline then 5 else 6 end) -- 5. normal, same day, same air, 6. normal, same day, different air.
else (case when a.airline=b.airline then 7 else 8
end) end) -- 7. normal, next day, same air, 8. normal, next day, different air
end end as priority_metric,
null as alt1dep_metric,
case when b.fl_2_sch_dep is null then null else
case when datediff(mi, a.fl_2_sch_dep, b.fl_2_sch_dep) < @waitAfterCancel
then
-datediff(mi,a.fl_2_sch_dep,b.fl_2_sch_dep)
else datediff(mi,a.fl_2_sch_dep,b.fl_2_sch_dep)*(5*24*60)
end end as alt2dep_metric
from
-- SET OF ALL ITINERARIES WITH FLIGHT 2 CANCELLED
select *
from Final
where
-- fl_1_diverted = 0
fl_1_cancelled = 0
and fl_2_cancelled = 1
-- and dateadd(minute,30,fl_1_act_arr) > fl_2_act_dep
a left join

-- SET OF FLIGHTS in which Flight 2 is not cancelled or diverted and have seats remaining
select distinct
fl_2_num, airline, fl_2_sch_arr, origin, dest, hub,
fl_2_lf, fl_2_avg_seat, fl_2_act_arr
from Final
where
fl_2_diverted = 0 -- Flight 2 cannot be diverted
and fl_2_cancelled = 0 -- Flight 2 cannot be cancelled
and fl_2_lf < 1 -- Flight 2 cannot be full

on
a.FL_1_SCH_DEP >= @startdate and a.FL_1_SCH_DEP < @enddate and
a.origin = b.origin
and a.dest = b.dest
and a.hub = b.hub
-- and a.airline = b.airline
-- the alternate flight 2 should be scheduled atleast 30 minutes after the arrival of the cancelled flight:
and b.fl_2_sch_dep > dateadd(minute, @waitAfterCancel, a.fl_1_act_arr)

/*if the time range is 24 hours:*/
-- and datediff(mi, a.fl_2_sch_dep, b.fl_2_sch_dep) <= @max_wait
/*if the time range is the whole prev. day*/
and
abs(datediff(dd, convert(smalldatetime, convert(date, a.fl_2_sch_dep)), b.fl_2_sch_dep)) <= 1
-- and year (a.fl_2_sch_dep) = year (b.fl_2_sch_dep)
-- and month(a.fl_2_sch_dep) = month(b.fl_2_sch_dep)
-- and day(a.fl_2_sch_dep) = day (b.fl_2_sch_dep)

END
GO

13) Stored procedure

USE [PaxDelay]
GO

******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ==============================================================
-- Author: <Sanja Avramovic>
-- Create date: <10/23/2014>
-- Description: < Populate for Delayed flights>
-- Based on Guermo’s and Ashwin’s code
-- TESTING NOTE: the range of flights to select was temporarily changed from a @max_wait minutes range to a +/-1 day range. Change commented lines to change back.
-- ==============================================================
CREATE PROCEDURE [dbo].[Populate_Delayed]
BEGIN

-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

declare @enddate smalldatetime
set @enddate = dateadd(dd, @numdays, @startdate)

set @waitAfterDelay = 30
set @lf_mult = 1 + @lf_adj;

insert into ##options_temp
insert into ##options_base

( Itin_status, origin, dest, Original_Hub, New_Hub, Total_Pax_on_Flight, Remaining_pax, Original_Airline, New_Airline,
  Original_Flight_Status, Original_Flight_1, Original_Flight_1_sch_dep, Original_Flight_1_sch_arr, Original_Flight_1_act_arr,
  Alternate_Flight_1, Alternate_Flight_1_sch_dep, Alternate_Flight_1_sch_arr, Alternate_Flight_1_lf,
  Alternate_Flight_1_total_seats, Alternate_Flight_1_ava_seats, Initial_Alternate_Flight_1_ava_seats,
  Alternate_Flight_1_alt_priority, Alternate_Flight_1_alt1dep_metric, Alternate_Flight_1_alt2dep_metric,
  Total_Rebooked_Pax, PTD_in_mins, unit_trip_delay, unit_trip_time, unit_sch_time, alt_priority, priority_metric, alt1dep_metric, alt2dep_metric )

select 0,-- For missed connections
  a.origin, a.dest,
  a.airline as Original_Airline, b.airline as New_Airline,
  a.max_pax as Total_Pax_on_Flight,
  a.max_pax as Remaining_pax,
  a.fl_1_num as Original_Flight_1_num,
  a.fl_1_diverted as Original_Flight_Status,
  a.fl_1_sch_dep as Original_Flight_1_sch_dep,
  a.fl_1_act_arr as Original_Flight_1_act_arr,
  a.fl_1_ava as Initial_Alternate_Flight_1_ava_seats,
  a.fl_1_avg_seat * @lf_mult as Alternate_Flight_1_total_seats,
  (1 - a.fl_1_ava) * a.fl_1_avg_seat * @lf_mult as Alternate_Flight_1_ava_seats,
  a.fl_1_lf as Alternate_Flight_1_lf,
  a.fl_1_alt as Alternate_Flight_1_alt_priority,
  a.fl_1_alt1dep as Alternate_Flight_1_alt1dep_metric,
  a.fl_1_alt2dep as Alternate_Flight_1_alt2dep_metric
FROM
  #options_temp a
  FULL OUTER JOIN #options_base b ON a.origin = b.origin AND a.dest = b.dest
  AND a.fl_1_sch_dep = b.fl_2_sch_dep
ORDER BY
  b.origin, b.dest, b.fl_2_sch_dep

-- For connections with delays
set @waitAfterDelay = 30
set @lf_mult = 1 + @lf_adj;

insert into ##options_temp
insert into ##options_base

( Itin_status, origin, dest, Original_Hub, New_Hub, Total_Pax_on_Flight, Remaining_pax, Original_Airline, New_Airline,
  Original_Flight_Status, Original_Flight_1, Original_Flight_1_sch_dep, Original_Flight_1_sch_arr, Original_Flight_1_act_arr,
  Alternate_Flight_1, Alternate_Flight_1_sch_dep, Alternate_Flight_1_sch_arr, Alternate_Flight_1_lf,
  Alternate_Flight_1_total_seats, Alternate_Flight_1_ava_seats, Initial_Alternate_Flight_1_ava_seats,
  Alternate_Flight_1_alt_priority, Alternate_Flight_1_alt1dep_metric, Alternate_Flight_1_alt2dep_metric,
  Total_Rebooked_Pax, PTD_in_mins, unit_trip_delay, unit_trip_time, unit_sch_time, alt_priority, priority_metric, alt1dep_metric, alt2dep_metric )

select 0,-- For missed connections
  a.origin, a.dest,
  a.airline as Original_Airline, b.airline as New_Airline,
  a.max_pax as Total_Pax_on_Flight,
  a.max_pax as Remaining_pax,
  a.fl_1_num as Original_Flight_1_num,
  a.fl_1_diverted as Original_Flight_Status,
  a.fl_1_sch_dep as Original_Flight_1_sch_dep,
  a.fl_1_act_arr as Original_Flight_1_act_arr,
  a.fl_1_ava as Initial_Alternate_Flight_1_ava_seats,
  a.fl_1_avg_seat * @lf_mult as Alternate_Flight_1_total_seats,
  (1 - a.fl_1_ava) * a.fl_1_avg_seat * @lf_mult as Alternate_Flight_1_ava_seats,
  a.fl_1_lf as Alternate_Flight_1_lf,
  a.fl_1_alt as Alternate_Flight_1_alt_priority,
  a.fl_1_alt1dep as Alternate_Flight_1_alt1dep_metric,
  a.fl_1_alt2dep as Alternate_Flight_1_alt2dep_metric
FROM
  #options_temp a
  FULL OUTER JOIN #options_base b ON a.origin = b.origin AND a.dest = b.dest
  AND a.fl_1_sch_dep = b.fl_2_sch_dep
ORDER BY
  b.origin, b.dest, b.fl_2_sch_dep

-- For connections with delays
set @waitAfterDelay = 30
set @lf_mult = 1 + @lf_adj;

insert into ##options_temp
insert into ##options_base

( Itin_status, origin, dest, Original_Hub, New_Hub, Total_Pax_on_Flight, Remaining_pax, Original_Airline, New_Airline,
  Original_Flight_Status, Original_Flight_1, Original_Flight_1_sch_dep, Original_Flight_1_sch_arr, Original_Flight_1_act_arr,
  Alternate_Flight_1, Alternate_Flight_1_sch_dep, Alternate_Flight_1_sch_arr, Alternate_Flight_1_lf,
  Alternate_Flight_1_total_seats, Alternate_Flight_1_ava_seats, Initial_Alternate_Flight_1_ava_seats,
  Alternate_Flight_1_alt_priority, Alternate_Flight_1_alt1dep_metric, Alternate_Flight_1_alt2dep_metric,
  Total_Rebooked_Pax, PTD_in_mins, unit_trip_delay, unit_trip_time, unit_sch_time, alt_priority, priority_metric, alt1dep_metric, alt2dep_metric )

select 0,-- For missed connections
  a.origin, a.dest,
  a.airline as Original_Airline, b.airline as New_Airline,
  a.max_pax as Total_Pax_on_Flight,
  a.max_pax as Remaining_pax,
  a.fl_1_num as Original_Flight_1_num,
  a.fl_1_diverted as Original_Flight_Status,
  a.fl_1_sch_dep as Original_Flight_1_sch_dep,
  a.fl_1_act_arr as Original_Flight_1_act_arr,
  a.fl_1_ava as Initial_Alternate_Flight_1_ava_seats,
  a.fl_1_avg_seat * @lf_mult as Alternate_Flight_1_total_seats,
  (1 - a.fl_1_ava) * a.fl_1_avg_seat * @lf_mult as Alternate_Flight_1_ava_seats,
  a.fl_1_lf as Alternate_Flight_1_lf,
  a.fl_1_alt as Alternate_Flight_1_alt_priority,
  a.fl_1_alt1dep as Alternate_Flight_1_alt1dep_metric,
  a.fl_1_alt2dep as Alternate_Flight_1_alt2dep_metric
FROM
  #options_temp a
  FULL OUTER JOIN #options_base b ON a.origin = b.origin AND a.dest = b.dest
  AND a.fl_1_sch_dep = b.fl_2_sch_dep
ORDER BY
  b.origin, b.dest, b.fl_2_sch_dep
b.fl_2.act_arr as Alternate_Flight_2_Act_Arr,
0 as Total_Rebooked_Pax,
0 as PTD_in_mins,
0 as unit_trip_delay,
0 as unit_trip_time,
0 as unit_sch_time,

/*
1st - rebook on flights from the same airline (and it's
subsidiaries) on that day
2nd rebook on flights from other airlines on that day
3rd rebook on flights for the same airline (and it's
subsidiaries) on the next day
4th rebook on flights from other airlines on the next day.
5th passengers not rebooked after any of these, would get
assigned a 900 minute delay
*/
case when b.fl_2.sch_dep is null then -1 else
    case when datediff(mi, a.fl_2.sch_dep, b.fl_2.sch_dep) < @waitAfterDelay
        then (case when day(a.fl_2.sch_dep) != day (b.fl_2.sch_dep) and
            datediff(mi,a.fl_2.sch_dep,b.fl_2.sch_dep)<0
            then (case when a.airline=b.airline then 1 else 2
            end)-- 1. early, day before,same airline, 2. early, not same day, different airline
        else (case when a.airline=b.airline then 3 else 4
        end) end)-- 3. early, same day, same airline, 4. early, same day,different airline
    else (case when day(a.fl_2.sch_dep) = day (b.fl_2.sch_dep) then
            (case when a.airline=b.airline then 5 else 6 end)-- 5. normal, same day, same air,  6. normal,
            same day, different air.
            else (case when a.airline=b.airline then 7 else 8
            end) end)-- 7. normal, next day, same air , 8. normal, next day, different air
        end as alt_priority,
/* case when a.airline=b.airline and day(a.fl_2.sch_dep) = day
(b.fl_2.sch_dep) then 1
(b.fl_2.sch_dep) then 2
when a.airline != b.airline and day(a.fl_2.sch_dep) = day
when a.airline=b.airline
then 3
else
4
end as alt_priority,*/

/* ordering metrics */
case when b.fl_2.sch_dep is null then -1 else
    case when datediff(mi, a.fl_2.sch_dep, b.fl_2.sch_dep) < @waitAfterDelay
        then (case when day(a.fl_2.sch_dep) != day (b.fl_2.sch_dep) and
            datediff(mi,a.fl_2.sch_dep,b.fl_2.sch_dep)<0
            then (case when a.airline=b.airline then 3 else 4
            end)-- 1. early, day before,same airline, 2. early, not same day, different airline
        else (case when a.airline=b.airline then 1 else 2
        end) end)-- 3. early, same day, same airline, 4. early, same day,different airline
    else

(case when day(a.fl_2_sch_dep) = day(b.fl_2_sch_dep) then
  (case when a.airline=b.airline then 5 else 6 end)  -- 5. normal, same day, same air,   6. normal, same day, different air.
else (case when a.airline=b.airline then 7 else 8 end) end)  --7. normal, next day, same air, 8. normal, next day, different air
end
end as priority_metric,
null as alt1dep_metric,
case when b.fl_2_sch_dep is null then null else
  case when a.airline=b.airline then 5 else 6 end
end as alt2dep_metric
from

(  -- SET OF ALL DELAYED FLIGHTS with missed connections-- Flight 1 is DELAYED
  select *
  from Final
  where
    fl_1_diverted = 0
    and fl_1_cancelled = 0
    and dateadd(minute,30,fl_1_act_arr) > fl_2_act_dep
  )a left join

(  -- SET OF FLIGHTS in which Flight 2 is not cancelled or diverted and have seats remaining
  select distinct
    fl_2_num,airline,fl_1_sch_dep,fl_2_sch_arr,origin,dest,hub,
    fl_2_lf,fl_2_avg_seat,fl_2_act_arr
  from Final
  --from pax_trip_delay
  where
    fl_2_diverted = 0  -- Flight 2 cannot be diverted
    and fl_2_cancelled = 0  -- Flight 2 cannot be cancelled
    and fl_2_lf < 1  -- Flight 2 cannot be full
  )b

 on
  a.fl_1_sch_dep >=startdate and a.fl_1_sch_dep < enddate and
  a.origin = b.origin
  and a.dest = b.dest
  and a.hub = b.hub
  --and a.airline = b.airline
  -- the alternate flight 2 should be scheduled at least 30 minutes after the arrival of the diverted flight:
  and b.fl_2_sch_dep => dateadd(minute,waitAfterDelay,a.fl_1_act_arr) /*if the time range is 24 hours:*/
  and datediff(mi,a.fl_2_sch.dep,b.fl_2_sch_dep) <= @max_wait /*if the time range is the whole prev. day*/
  and abs(datediff(dd,convert(smalldatetime,convert(date,a.fl_2_sch_dep)),b.fl_2_sch_dep))<=1
  --and year(a.fl_1_sch_dep) = year(b.fl_2_sch_dep)
  --and month(a.fl_1_sch_dep) = month(b.fl_2_sch_dep)
  --and day(a.fl_1_sch_dep) = day(b.fl_2_sch_dep)

END
14) Stored procedure

USE [PaxDelay]
GO

******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ---------------------------------------------------------
-- Author:        <Sanja Avramovic>
-- Create date:  <10/23/2014>
-- Description: <Finding candidate flights for missed connections due to diversions and delays>
-- Based on Guermo's and Ashwin's code
-- TESTING NOTE: the range of flights to select was temporarily changed from a @max_wait minutes range to a +/-1 day range. Change commented lines to change back.
-- ---------------------------------------------------------
CREATE PROCEDURE [dbo].[Populate_Diverted]
    @startdate smalldatetime,
    @numdays int,
    @max_wait int = 600,
    @lf_adj float = 0
AS
BEGIN
    -- SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.
    SET NOCOUNT ON;
    declare @enddate smalldatetime;
    declare @waitAfterDiverted int
    set @waitAfterDiverted = 30
    set @enddate = dateadd(dd, @numdays, @startdate)
    declare @lf_mult float;
    set @lf_mult = 1+@lf_adj;
    --insert into ##options_temp(
    insert into ##options_base(
        Itin_status,
        origin.dest,Original_Hub,New_Hub,
        Total_Pax_on_Flight,
        Remaining_pax,
        Original_Airline,New_Airline,
        Original_Flight_Status,Original_Flight_1,Original_Sch_Dep,Original_Flight_1_Sch_Arr,
        Alternate_Flight_1,Alternate_Flight_1_Sch_Dep,Alternate_Flight_1_Sch_Arr,Alternate_Flight_1_LF,
        Alternate_Flight_1_TotalSeats,Alternate_Flight_1_AvaSeats,Initial_Alternate_Flight_1_AvaSeats
    )
    ON DUPLICATE KEY UPDATE
        Itin_status = values(Itin_status),
        origin.dest = values(origin.dest),Original_Hub = values(Original_Hub),New_Hub = values(New_Hub),
        Total_Pax_on_Flight = values(Total_Pax_on_Flight),
        Remaining_pax = values(Remaining_pax),
        Original_Airline = values(Original_Airline),New_Airline = values(New_Airline),
        Original_Flight_Status = values(Original_Flight_Status),Original_Flight_1 = values(Original_Flight_1),
        Original_Sch_Dep = values(Original_Sch_Dep),Original_Flight_1_Sch_Arr = values(Original_Flight_1_Sch_Arr),
        Alternate_Flight_1 = values(Alternate_Flight_1),Alternate_Flight_1_Sch_Dep = values(Alternate_Flight_1_Sch_Dep),
        Alternate_Flight_1_Sch_Arr = values(Alternate_Flight_1_Sch_Arr),Alternate_Flight_1_LF = values(Alternate_Flight_1_LF),
        Alternate_Flight_1_TotalSeats = values(Alternate_Flight_1_TotalSeats),Alternate_Flight_1_AvaSeats = values(Alternate_Flight_1_AvaSeats),
        Initial_Alternate_Flight_1_AvaSeats = values(Initial_Alternate_Flight_1_AvaSeats)
select 0, -- For missed connections
  a.origin, a.dest,
  a.hub as Hub, null,
  a.max_pax as Total_Pax_on_Flight,
  a.max_pax as Remaining_Pax,
  a.airline as Original_Airline, b.airline as New_Airline,
  a.fl_1_diverted as Original_Flight_Status, a.fl_1_num as
  Original_Flight_1, a.fl_1_sch as Original_Sch_Dep, null, a.fl_1_act as
  Original_Flight_1_Act_Arr,
  a.fl_2_num as Original_Flight_2, a.fl_2_sch as Original_Sch_Arr,
  null, null, null, null, null, null, null, null,
  b.fl_2_num as Alternate_Flight_2, b.fl_2_sch as
  Alternate_Flight_2_Sch_Dep, b.fl_2_sch as Alternate_Sch_Arr, b.fl_2_lf as
  Alternate_Flight_2_LF, b.fl_2_avg * @lf_mult as Alternate_Flight_2_TotalSeats,
  (1-b.fl_2_lf)*b.fl_2_avg * @lf_mult as Alternate_Flight_2_AvaSeats,
  (1-b.fl_2_lf)*b.fl_2_avg * @lf_mult as
  Initial_Alternate_Flight_2_AvaSeats,
  b.fl_2_act as Alternate_Flight_2_Act_Arr,
  0 as Total_Rebooked_Pax,
  0 as PTD_in_mins,
  0 as unit_trip_delay,
  0 as unit_trip_time,
  0 as unit_sch_time,

/ *
  1st - rebook on flights from the same airline (and it's
  subsidiaries) on that day
  2nd rebook on flights from other airlines on that day
  3rd rebook on flights for the same airline (and it's
  subsidiaries) on the next day
  4th rebook on flights from other airlines on the next day.
  5th passengers not rebooked after any of these, would get
  assigned a 900 minute delay
  */
  case when b.fl_2_sch is null then -1 else
  case when datediff(mi.a.fl_2_sch, b.fl_2_sch) <
  @waitAfterDiverted then
  case when day(a.fl_2_sch) != day(b.fl_2_sch) and
  datediff(mi.a.fl_2_sch, b.fl_2_sch) > 0
  then (case when a.airline = b.airline then 1 else 2
  end)-- 1. early, day before, same airline, 2. early, not same day, different airline
  else (case when a.airline = b.airline then 3 else 4
  end) end)-- 3. early, same day, same airline, 4. early, same day, different airline
  else (case when day(a.fl_2_sch) = day(b.fl_2_sch) then
  (case when a.airline = b.airline then 5 else 6
  end)-- 5. normal, same day, same airline, 6. normal,
  same day, different airline.
  else (case when a.airline = b.airline then 7 else 8
  end) end)-- 7. normal, next day, same airline, 8. normal, next day, different airline
  end end as alt_priority,
/*case when a.airline=b.airline and day(a.fl_2_sch_dep) = day(b.fl_2_sch_dep) then 1 
(b.fl_2_sch_dep) then 2 
when a.airline != b.airline and day(a.fl_2_sch_dep) = day(b.fl_2_sch_dep) then 2 
when a.airline=b.airline then 3 
else 4 
end as alt_priority,*/

/* ordering metrics */

 case when b.fl_2_sch_dep is null then -1 
   else case when datediff(mi, a.fl_2_sch_dep, b.fl_2_sch_dep) < @waitAfterDiverted then 
   (case when day(a.fl_2_sch_dep) != day(b.fl_2_sch_dep) and 
   datediff(mi,a.fl_2_sch_dep,b.fl_2_sch_dep)=0
    then (case when a.airline=b.airline then 3 else 4 
    end)-- 1. early, day before, same airline, 2. early, not same day, different airline 
    end end)-- 3. early, same day, same airline, 4. early, same day, different airline 
    else 
    (case when day(a.fl_2_sch_dep) = day(b.fl_2_sch_dep) then 
    (case when a.airline=b.airline then 5 else 6 
    end)-- 5. normal, same day, same airline, 6. normal, same day, different airline. 
    end end)-- 7. normal, next day, same airline, 8. normal, next day, different airline 
  end end as priority_metric,

  null as alt1dep_metric,

  case when b.fl_2_sch_dep is null then null 
    else case when datediff(mi, a.fl_2_sch_dep, b.fl_2_sch_dep) < 
    @waitAfterDiverted then 
    -datediff(mi,a.fl_2_sch_dep,b.fl_2_sch_dep)
    else datediff(mi,a.fl_2_sch_dep,b.fl_2_sch_dep)+(5*24*60) 
    end 
    end as alt2dep_metric

from

(-- SET OF ALL DIVERTED FLIGHTS with missed connections-- Flight 1 is Diverted
select * 
--from #temp_pax_trip_delay
from Final
where 
   fl_1_diverted = 1 
   and fl_1_cancelled = 0 
   and dateadd(minute,30,fl_1_act_arr) > fl_1_act_dep 
)a left join

(-- SET OF FLIGHTS in which Flight 2 is not cancelled or diverted and have seats
remaining
select distinct fl_2_num,airline,fl_2_sch_dep,fl_2_sch_arr,origin,dest,hub, 
fl_2_lf,fl_2_avg_seat,fl_2_act_arr
from Final
--from pax_trip_delay
where 
   fl_2_diverted = 0 -- Flight 2 cannot be diverted

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and fl_2_cancelled = 0 -- Flight 2 cannot be cancelled
and fl_2 lf < 1 -- Flight 2 cannot be full
)

b

on

a.FL_1_SCH_DEP > @startdate and a.FL_1_SCH_DEP < @enddate and a.origin = b.origin

and a.dest = b.dest
and a.hub = b.hub
-- and a.airline = b.airline
-- the alternate flight 2 should be scheduled at least 30 minutes after the
arrival of the diverted flight:
and b.fl_2_sch_dep = dateadd(minute, @waitAfterDiverted, a.fl_1_act_arr)
/*if the time range is 24 hours:*/
-- and datediff(m, a.fl_2_sch_dep, b.fl_2_sch_dep) <= @max_wait
/*if the time range is the whole prev. day*/
and abs(datediff(dd, convert(smalldatetime, convert(date, a.fl_2_sch_dep)), b.fl_2_sch_dep)) <= 1
-- and year(a.fl_1_sch_dep) = year(b.fl_2_sch_dep)
-- and month(a.fl_1_sch_dep) = month(b.fl_2_sch_dep)
-- and day(a.fl_1_sch_dep) = day(b.fl_2_sch_dep)

END
GO

15) Stored procedure

USE [PaxDelay]
GO

*******/
SET ANSI_NULLS ON
GO

SET QUOTED_IDENTIFIER ON
GO

-- -=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=
-- Author:   Guillermo Calderón-Meza
-- Create date: not recorded (Modified on May 18, 2009)
-- Description: Computes the number of available seats for the records in #on_time_tmp based on
load_factors data
-- -=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=-=
CREATE PROCEDURE [dbo].[Update_Avaseat]
    @year int,
    @cur_month int
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

declare @start_date smalldatetime, @end_date smalldatetime

set @start_date = dbo.parts2date(@year, @cur_month, 1)
set @end_date = dateadd(month, 1, @start_date)

update #on_time_tmp set avaseat = null
    where sch_dep >= @start_date and
    sch_dep < @end_date

print 'computing available seats not cancelled'
-- set value for column avaseat for non-cancelled flight, note I use carrier instead of
airline,
-- so mq and aa has different seat or pax info
update #on_time_tmp set avaseat = (select avaseat from load_factors
where year = datepart(year,#on_time_tmp.sch_dep)
and month = datepart(month,#on_time_tmp.sch_dep)
and international = 0 -- consider only domestic
and origin = #on_time_tmp.origin
and dest = #on_time_tmp.dest
and carrier = #on_time_tmp.airline)
where cancelled <> 1
and sch_dep >= @start_date
-- Do not include @end_date in the range because it is actually out of the intended range
and sch_dep < @end_date

print 'computing available seats not cancelled, still null'
-- some carriers might not be included in load_factors table, then their avaseat value will remain null,
-- so their avaseat are set to be the avg for all carriers
update #on_time_tmp set avaseat = (select avg(avaseat) from load_factors
where year = datepart(year,#on_time_tmp.sch_dep)
and month = datepart(month,#on_time_tmp.sch_dep)
and international = 0 -- consider only domestic
and origin = #on_time_tmp.origin
and dest = #on_time_tmp.dest
and cancelled <> 1
and avaseat is null
and sch_dep >= @start_date
-- Do not include @end_date in the range because it is actually out of the intended range
and sch_dep < @end_date

print 'Setting a default of 50 - 50 * 0.7 available seats to records with avaseat null (assuming small airport)'
update #on_time_tmp set avaseat = 50 - 50 * 0.7
where cancelled <> 1
and avaseat is null
and sch_dep >= @start_date
-- Do not include @end_date in the range because it is actually out of the intended range
and sch_dep < @end_date

print 'computing available seats cancelled, all = 0'
-- set value for column avaseat for cancelled flight
update #on_time_tmp set avaseat = 0
where cancelled = 1
and sch_dep >= @start_date
-- Do not include @end_date in the range because it is actually out of the intended range
and sch_dep < @end_date

END
GO

16) Stored procedure

USE [PaxDelay]
GO

/****** Object:  StoredProcedure [dbo].[Update_Avaseats1]    Script Date: 5/27/2015 12:38:40 AM
******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- -------------------------------------------------------------------------
-- Author: <Sanja Avramovic>
-- Create date: <11/29/2014>
-- Description: update the number of available seats on an alternative flight.
-- Based on Guermo's and Ashwin's code
-- -------------------------------------------------------------------------
CREATE PROCEDURE [dbo].[Update_Avaseats1]
    @alt_fl_1_avaseats smallint,
    @alt_fl_1_lf float,
    @alt_fl_1_num smallint,
    @alt_fl_1_sch_dep smalldatetime
AS
BEGIN
    -- SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.
    SET NOCOUNT ON;
    update #options_temp
    set Alternate_Flight_1_AvaSeats = @alt_fl_1_avaseats,
    Alternate_Flight_1_LF = @alt_fl_1_lf
    where
    Alternate_Flight_1 = @alt_fl_1_num
    and Alternate_Flight_1_Sch_Dep = @alt_fl_1_sch_dep;
END
GO

17) Stored procedure

USE [PaxDelay]
GO

******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- -------------------------------------------------------------------------
-- Author: <Sanja Avramovic>
-- Create date: <11/29/2014>
-- Description: update the number of available seats on an alternative flight.
-- Based on Guermo's and Ashwin's code
-- -------------------------------------------------------------------------
CREATE PROCEDURE [dbo].[Update_Avaseats2]
    @alt_fl_2_avaseats smallint,
    @alt_fl_2_lf float,
    @alt_fl_2_num smallint,
    @alt_fl_2_sch_arr smalldatetime
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;
update #options_temp
set Alternate_Flight_2_AvaSeats = @alt_fl_2_avaseats, Alternate_Flight_2_LF = @alt_fl_2_lf
where
  Alternate_Flight_2 = @alt_fl_2_num
  and Alternate_Flight_2_Sch_Arr = @alt_fl_2_sch_arr
END
GO

18) Stored procedure

USE [PaxDelay]
GO
/****** Object:  StoredProcedure [dbo].[Update_Avgpax]    Script Date: 5/27/2015 12:40:00 AM
******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ==============================================--
-- Author:  Guillermo Calderon-Maza
-- Create date: Not recorder (Modified on May 18, 2009)
-- Description: Computes the average number of passengers (pax) in the flights based on the T100
-- information
-- It assumes that there is a temporary table called
#on_time_tmp and that it was initialized
-- with some (most) of the values for avgpax, but
there are still some avgpax field with null value
-- Checking for dates is not necessary since #on_time_tmp is
correctly populated with the relevant records
-- ==============================================--
CREATE PROCEDURE [dbo].[Update_Avgpax]
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

print 'computing average pax still null'
-- for carrier not included in load_factors table, they use avg value of all carriers
update #on_time_tmp set avgpax = ( 
  select avg(avgpax)
  from load_factors
  where year = datepart(year,#on_time_tmp.sch_dep)
    and month = datepart(month,#on_time_tmp.sch_dep)
    and international = 0 -- consider only domestic
    and origin = #on_time_tmp.origin
    and dest = #on_time_tmp.dest
  group by origin, dest)
where avgpax is null

print 'Setting a default avgpax of 0.7 * 50 for the remaining flights (assuming they correspond to small airports)'
update #on_time_tmp set avgpax = 0.7 * 50
19) Stored procedure

USE [PaxDelay]
GO
******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- =============================================
-- Author:  <Sanja Avramovic>
-- Create date: <10/29/2014>
-- Description: <UPDATING TABLE FOR DELAYS AND DIVERSIONS-->
-- Based on Guermo's and Ashwin's code
-- =============================================
CREATE PROCEDURE [dbo].[Update_DelDiv]
    -- Add the parameters for the stored procedure here
@extraDivDelay int
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.
SET NOCOUNT ON;

update Final
set
PTD = max_pax*(datediff(mi,fl_2_sch_arr,fl_2_act_arr)-15),
trip_delay = max_pax*(datediff(mi,fl_1_sch_arr,fl_1_act_arr)-15),
trip_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_2_act_arr) as bigint),
sch_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_sch_arr) as bigint)

where
-- Setting the conditions for flights not being cancelled or diverted
fl_1_cancelled = 0
and fl_1_2_cancelled = 0
and fl_1_diverted = 0
and fl_2_diverted = 0
-- and fl_1_sch_dep >=@start_date
-- and fl_2_sch_arr <= dateadd(minute,30,fl_1_act_arr) -- Sets the condition for catching a connecting flight.
-- A flight is termed a delayed flight only if its Actual time of arrival is more than 15 minutes than that of its scheduled arrival
and fl_2_act_arr > dateadd(minute,15,fl_1_sch_arr) -- Sets the condition for calculating delay beyond 15 minutes

-- Delayed Flights
-- Direct
update Final
set
PTD = max_pax*(datediff(mi,fl_1_sch_arr,fl_1_act_arr)-15),
trip_delay = max_pax*(datediff(mi,fl_1_sch_arr,fl_1_act_arr)-15),
trip_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_act_arr) as bigint),
sch_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_sch_arr) as bigint)

where
  hub is null
  and fl_1_cancelled = 0
  and fl_1_diverted = 0
  and datediff(mi,fl_1_sch_arr,fl_1_act_arr)-15 >0

-- Diverted Flights
-- CONSIDERING THAT ONLY FLIGHT 2 (CONNECTING FLIGHTS) ARE DIVERTED----

update Final
set PTD =
max_pax*(cast(paxdelay.dbo.estimate_diverted_delay(fl_1_sch_dep,hub,dest,default,0[rand() as int]) as int)+extraDivDelay),
  trip_delay =
max_pax*(cast(paxdelay.dbo.estimate_diverted_delay(fl_1_sch_dep,hub,dest,default,0[rand() as int]) as int)+extraDivDelay),
  trip_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_2_act_arr) as bigint),
  sch_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_2_sch_arr) as bigint)
where
  fl_1_cancelled = 0
  and fl_2_cancelled = 0
  and fl_1_diverted = 0
  and fl_2_diverted = 1
  --and fl_1_sch_dep >@start_date
  --and fl_1_sch_dep <@end_date
  and fl_2_act_dep > dateadd(minute,30,fl_1_act_arr) -- Sets the condition for catching a connecting flight

-- Direct Flights

update Final
set PTD =
max_pax*(cast(paxdelay.dbo.estimate_diverted_delay(fl_1_sch_dep,origin,dest,default,default,0[rand() as int]) as int)+extraDivDelay),
  trip_delay =
max_pax*(cast(paxdelay.dbo.estimate_diverted_delay(fl_1_sch_dep,origin,dest,default,default,0[rand() as int]) as int)+extraDivDelay),
  trip_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_act_arr) as bigint),
  sch_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_sch_arr) as bigint)
where
  fl_1_cancelled = 0
  and fl_1_diverted = 1
  and fl_2_num is null
  and fl_2_sch_dep is null

-- On time flights

update Final
set trip_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_act_arr) as bigint),
  sch_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_1_sch_arr) as bigint)
where
  fl_2_num is null
  and fl_2_sch_dep is null
  and fl_1_cancelled = 0
  and fl_1_diverted = 0
  and datediff(mi,fl_1_sch_arr,fl_1_act_arr) <= 15;

update Final
set trip_time = cast(max_pax*datediff(mi,fl_1_sch_dep,fl_2_act_arr) as bigint),

179
sch_time = cast(max_pax * datediff(mi, fl_1_sch_dep, fl_2_sch_arr) as bigint)

where

fl_1_cancelled = 0
and fl_1_2_cancelled = 0
and fl_1_diverted = 0
and fl_1_2_diverted = 0
and fl_1_2_act_dep > dateadd(minute, 30, fl_1_act_arr)
and fl_1_2_act_arr < dateadd(minute, 15, fl_1_2_sch_arr);

END
GO

20) Stored procedure

USE [PaxDelay]
GO

_DATABASEisplayed_script_date: 5/27/2015 12:41:12
 DATABASEisplayed_script_author: Sanja Avramovic
 DATABASEisplayed_script_description: Update PTDs in Final table for a
 chosen itinerary.
 DATABASEisplayed_script_based_on: Guermo's and Ashwin's code

CREATE PROCEDURE [dbo].[Update_Itinerary_PTDs]

-- Add the parameters for the stored procedure here
@last_fl_1_num smallint,
@last_fl_2_num smallint,
@last_fl_1_sch_dep smalldatetime,
@last_fl_2_sch_arr smalldatetime
AS
BEGIN

-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

update Final set PTD = (select sum(PTD_in_mins)
from #options_temp
where @last_fl_1_num = original_flight_1
and @last_fl_1_sch_dep

original_flight_2
or (@last_fl_2_num is null and original_flight_2 is null))

and (@last_fl_2_num =
original_flight_2 or (@last_fl_2_num is null and original_flight_2 is null))

and fl_1_2_diverted = 0

and fl_1_2_cancelled = 0
and fl_1_2_act_dep > dateadd(minute, 30, fl_1_act_arr)
and fl_1_2_act_arr < dateadd(minute, 15, fl_1_2_sch_arr);

END
GO

update Final set PTD = (select sum(PTD_in_mins)
from #options_temp
where @last_fl_1_num =
original_flight_1
and @last_fl_1_sch_dep

original_flight_2
or (@last_fl_2_num is null and original_flight_2 is null))

and (@last_fl_2_num =
original_flight_2 or (@last_fl_2_num is null and original Flight_2 is null))

and fl_1_2_diverted = 0
and fl_1_2_cancelled = 0
and fl_1_2_act_dep > dateadd(minute, 30, fl_1_act_arr)
and fl_1_2_act_arr < dateadd(minute, 15, fl_1_2_sch_arr);

END
GO
trip_delay = (select sum(total_rebooked_pax*unit_trip_delay) from #options_temp where @last_fl_1_num = original_flight_1 and @last_fl_1_sch_dep = Original_Sch_Dep and original_flight_2 or (@last_fl_2_num is null and original_flight_2 is null)) and (@last_fl_2_sch_arr = original_sch_arr or (@last_fl_2_sch_arr is null and original_sch_arr is null)) },

trip_time = (select sum(cast(total_rebooked_pax*unit_trip_time as bigint)) from #options_temp where @last_fl_1_num = original_flight_1 and @last_fl_1_sch_dep = Original_Sch_Dep and original_flight_2 or (@last_fl_2_num is null and original_flight_2 is null)) and (@last_fl_2_sch_arr = original_sch_arr or (@last_fl_2_sch_arr is null and original_sch_arr is null)) },

sch_time = (select sum(cast(total_rebooked_pax*unit_sch_time as bigint)) from #options_temp where @last_fl_1_num = original_flight_1 and @last_fl_1_sch_dep = Original_Sch_Dep and original_flight_2 or (@last_fl_2_num is null and original_flight_2 is null)) and (@last_fl_2_sch_arr = original_sch_arr or (@last_fl_2_sch_arr is null and original_sch_arr is null)) )

where fl_1_num = @last_fl_1_num and fl_1_sch_dep = @last_fl_1_sch_dep and (fl_2_num = @last_fl_2_num or (@last_fl_2_num is null and fl_2_sch_arr is null)) and (fl_2_sch_arr = @last_fl_2_sch_arr or (@last_fl_2_sch_arr is null and fl_2_num is null))

END
GO

21) Stored procedure

USE [PaxDelay]
GO

SET ANSI_NULLS ON
GO

SET QUOTED_IDENTIFIER ON
GO
CREATE PROCEDURE [dbo].[Update_MTable]
-- Add the parameters for the stored procedure here
@last_fl_1_num smallint,
@last_fl_2_num smallint,
@last_fl_1_sch_dep smalldatetime,
@last_fl_2_sch_arr smalldatetime
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

update Final
set PTD = (select sum(PTD_in_mins)
from ##options_temp
where @last_fl_1_num = original_flight_1
  and @last_fl_1_sch_dep = Original_Sch_Dep
  and (@last_fl_2_num is null and original_flight_2 is null))
  and (@last_fl_2_sch_arr = original_sch_arr or (@last_fl_2_sch_arr is null))

where fl_1_num = @last_fl_1_num
  and fl_1_sch_dep = @last_fl_1_sch_dep
  and (@last_fl_2_num is null and fl_2_num is null)
  or (@last_fl_2_num is null and fl_2_sch_arr is null))

END

GO

22) Stored procedure

USE [PaxDelay]
GO

SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO
CREATE PROCEDURE [dbo].[Update_PTDs]
-- Add the parameters for the stored procedure here

@total_rebooked_pax int,
@PTD int,
@remaining_pax int,
@alt_fl_1_avaseats smallint,
@alt_fl_1_lf float,
@alt_fl_2_avaseats smallint,
@alt_fl_2_lf float,
@og_fl_1_num smallint,
@og_fl_1_sch_dep smalldatetime,
@alt_fl_1_num smallint,
@alt_fl_1_sch_dep smalldatetime,
@og_fl_2_num smallint,
@og_fl_2_sch_arr smalldatetime,
@alt_fl_2_num smallint,
@alt_fl_2_sch_arr smalldatetime

AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

update ##options_temp
set Total_Rebooked_Pax = @total_rebooked_pax,
    PTD_in_mins = @PTD
where
    Original_Flight_1 = @og_fl_1_num
    and (Original_Flight_2 = @og_fl_2_num or
          (@og_fl_2_num is null and Original_Flight_2 is null))
    and Original_Sch_Dep = @og_fl_1_sch_dep
    and (Original_SchArr = @og_fl_2_sch_arr or
          (@og_fl_2_num is null and Original_Sch_arr is null))
    and Alternate_Flight_1 = @alt_fl_1_num
    and (Alternate_Flight_2 = @alt_fl_2_num or
          (@alt_fl_2_num is null and Alternate_Flight_2 is null))
    and Alternate_Flight_1_Sch_Dep = @alt_fl_1_sch_dep
    and (Alternate_Flight_2_Sch_Arr = @alt_fl_2_sch_arr or
          (@alt_fl_2_num is null and Alternate_Flight_2_sch_arr is null))

update ##options_temp
set Remaining_pax = @remaining_pax
where
    Original_Flight_1 = @og_fl_1_num
update #options_temp
  set Alternate_Flight_1_AvaSeats =
@alt_fl_1_avaseats, Alternate_Flight_1_LF = @alt_fl_1_lf
  where
Alternate_Flight_1 = @alt_fl_1_num
  and Alternate_Flight_1_Sch_Dep =
@alt_fl_1_schdep

update #options_temp
  set Alternate_Flight_2_AvaSeats =
@alt_fl_2_avaseats, Alternate_Flight_2_LF = @alt_fl_2_lf
  where
Alternate_Flight_2 = @alt_fl_2_num
  and Alternate_Flight_2_Sch_Arr =
@alt_fl_2_scharr

END
GO

23) Stored procedure

USE [PaxDelay]
GO

******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ========================================================
-- Author:  <Sanja Avramovic>
-- Create date: <10/29/2014>
-- Description:  <Update table - MasterTable for the PTDs 2>
-- Based on Guermo's and Ashwin's code
-- ========================================================
CREATE PROCEDURE [dbo].[Update_PTDsII]
  @total_rebooked_pax  int,
  @PTD  int,
  @remaining_pax  int,
  @alt_fl_1_avaseats  smallint,
  @alt_fl_1_lf  float,
  @og_fl_1_num  smallint,
  @og_fl_1_sch_dep  smalldatetime ,
  @og_fl_2_num  smallint,
  @og_fl_2_sch_arr  smalldatetime,
  @alt_fl_2_num  smallint,
  @alt_fl_2_sch_arr  smalldatetime
AS
BEGIN
  -- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

update ##options_temp
set Total_Rebooked_Pax = @total_rebooked_pax,
    PTD_in_mins = @PTD
where
    Original_Flight_1 = @og_fl_1_num
and Original_Flight_2 = @og_fl_2_num
and Original_Sch_Dep = @og_fl_1_sch_dep
and Original_Sch_Arr = @og_fl_2_sch_arr
and Alternate_Flight_2 = @alt_fl_2_num
and Alternate_Flight_2_Sch_Arr = @alt_f1_2_sch_arr

update ##options_temp
set Remaining_pax = @remaining_pax
where
    Original_Flight_1 = @og_fl_1_num
and Original_Flight_2 = @og_fl_2_num
and Original_Sch_Dep = @og_fl_1_sch_dep
and Original_Sch_Arr = @og_fl_2_sch_arr

update ##options_temp
set Alternate_Flight_2_AvaSeats = @alt_fl_2_avaseats,
    Alternate_Flight_2_LF = @alt_fl_2_lf
where
    Alternate_Flight_2 = @alt_fl_2_num
and Alternate_Flight_2_Sch_Arr = @alt_f1_2_sch_arr

END
GO

24) Stored procedure

USE [PaxDelay]
GO

SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ===============================================================================
-- Author:  <Sanja Avramovic>
-- Create date: <11/29/2014>
-- Description: Update options table to reflect PTD as a result of rebooking
-- Based on Guermo's and Ashwin's code
-- ===============================================================================
CREATE PROCEDURE [dbo].[Update_Rebook_PTD]
    @total_rebooked_pax int,
    @PTD int,
    @og_fl_1_num smallint,
AS
BEGIN
-- SET NOCOUNT ON added to prevent extra result sets from
-- interfering with SELECT statements.
SET NOCOUNT ON;

-- set the number of pax rebooked on the selected option, and the PTD resulting from rebooking
update #options_temp
set Total_Rebooked_Pax = @total_rebooked_pax, 
PTD_in_mins = @PTD

where
Original_Flight_1 = @og_fl_1_num
and Original_Sch_Dep = @og_fl_1_sch_dep
and (Original_Flight_2 = @og_fl_2_num or (@og_fl_2_num is null and
Original_Flight_2 is null))
and (Original_Sch_Arr = @og_fl_2_sch_arr or (@og_fl_2_num is null and
Original_Sch_arr is null))
and (Alternate_Flight_1 = @alt_fl_1_num or @alt_fl_1_num is null)
and (Alternate_Flight_1_Sch_Dep = @alt_fl_1_sch_dep or @alt_fl_1_sch_dep is null)
and (Alternate_Flight_2 = @alt_fl_2_num or @alt_fl_2_num is null and
Alternate_Flight_2_Sch_Dep = @alt_fl_2_sch_dep or @alt_fl_2_sch_dep is null)
and (Alternate_Flight_2_Sch_Arr = @alt_fl_2_sch_arr or @alt_fl_2_num is null and
Alternate_Flight_2_sch_arr is null));
END
GO

25) Stored procedure

USE [PaxDelay]
GO

AM ******/
SET ANSI_NULLS ON
GO
SET QUOTED_IDENTIFIER ON
GO

-- ================================
-- Author:  <Sanja Avramovic>
-- Create date: <10/29/2014>
-- Description: Update the number of remaining
-- pax on an itinerary (after rebooking).
-- Based on Guermo's and Ashwin's code
-- ================================
CREATE PROCEDURE [dbo].[Update_Remaining_Pax]
    @remaining_pax int,
    @og_fl_1_num smallint,
    @og_fl_1_sch_dep smalldatetime ,
    @og_fl_2_num smallint,
SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.

-- update every entry which matches the original itinerary with the new remaining pax value for that itinerary
update #options_temp
set Remaining_pax = @remaining_pax
where
    Original_Flight_1 = @og_fl_1_num
    and Original_Sch_Dep = @og_fl_1_sch_dep
    and (Original_Flight_2 = @og_fl_2_num or (@og_fl_2_num is null and Original_Flight_2 is null))
    and (Original_Sch_Arr = @og_fl_2_sch_arr or (@og_fl_2_num is null and Original_Sch_arr is null));

26) Stored procedure

USE [PaxDelay]
GO

SET ANSI_NULLS ON
GO

SET QUOTED_IDENTIFIER ON
GO

-- ========================================================
-- Author:     <Sanja Avramovic>
-- Create date: <11/08/2015>
-- Description: <update subsidiary codes>
-- Modifies MasterTable to change all subsidiary airlines to reflect the parent airline.
-- Based on Kevin’s and Ashwin’s code
-- ========================================================
CREATE PROCEDURE [dbo].[Update_Subsidiary]
    -- Add the parameters for the stored procedure here
    @start_date smalldatetime
AS
BEGIN
    -- SET NOCOUNT ON added to prevent extra result sets from interfering with SELECT statements.
    SET NOCOUNT ON;
    declare @newairline varchar(3)
    declare @oldairline varchar(3)
    declare @airport varchar(3)
    declare @status int
    declare @year as smallint
    declare @quarter as tinyint

    set @year = year(@start_date)
    set @quarter = (month(@start_date) - 1) / 3 + 1

    declare @newairline varchar(3)
declare subs cursor fast_forward
for
select newairline, oldairline, airport
from Subsidiary
where year = @year and quarter = @quarter

open subs

fetch next from subs into
@newairline, @oldairline, @airport

set @status = @@fetch_status

while (@status = 0) begin
update MasterTable
set airline = @newairline
where airline = @oldairline
and (@airport is NULL or @airport = origin or @airport = hub or @airport = dest);

fetch next from subs into
@newairline, @oldairline, @airport

set @status = @@fetch_status
end

close subs
deallocate subs
END
GO

27) Main program

Set nocount on
Use PaxDelay

declare @startdate smalldatetime;
declare @enddate smalldatetime;
declare @numdays int;
declare @extradays int; -- number of "extra" days to put on either side in order to get alts, etc
DECLARE @extstartdate smalldatetime;
declare @extenddate smalldatetime;
declare @extnumdays int;
declare @r int;
declare @preempt_same float;
declare @preempt_prev float;
declare @epoch int; -- number of times to run step 3 for each % choice
declare @lf_adj float; -- amount by which to adjust load factor for testing

set @numdays = 1;
set @startdate = '2012-07-18';
set @enddate = dateadd(dd, @numdays, @startdate);

set @extradays = 1;
set @extstartdate = dateadd(dd, @extradays, @startdate);
set @extenddate = dateadd(dd, @extradays, @enddate);
set @extnumdays = @numdays + 2 * @extradays;

set @epoch = 25;

set @preempt_prev = 0 -- 0.1; -- 10% rebooked prev. day
set @preempt_same = 0 -- 0.4; -- 40% rebooked earlier same day
set @lf_adj = -0.05 --0 --0.02 -- 0 no change to the default, so max 100% LF 0.02 is +2%, -0.02 is -2%
declare @origin varchar(3)
declare @dest varchar(3)
declare @hub varchar(3)
declare @airline varchar(3)
declare @fl_num smallint
declare @pax int
declare @status int
declare @totalpax int
declare @maxpaxperc float
declare @date smalldatetime
declare @t100mpax float
declare @prev_t100 float
declare @diff int
declare @lf float
declare @seats smallint

/*step 1 is in stored procedures (Create_DB1B, Create_ontime_lf, Create_aotp, Create_MasterTable)*/

print('Start Stebp 1' + cast(getdate() as varchar))
RAISERROR('Start Step 1',0,1) WITH NOWAIT;
exec dbo.Create_DB1B @startdate;
print('Created DB1B ' + cast(getdate() as varchar))
RAISERROR('Created DB1B',0,1) WITH NOWAIT;

-- create final table (or recreate it, if it exists
exec dbo.Create_Final;
exec dbo.Create_ontime_lf @extstartdate,@extnumdays;
print('Created ontime_lf ' + cast(getdate() as varchar))
RAISERROR('Created ontime_lf',0,1) WITH NOWAIT;
exec dbo.Create_aotp;
print('Created aotp ' + cast(getdate() as varchar))
RAISERROR('Created aotp',0,1) WITH NOWAIT;
exec dbo.Create_MasterTable @extstartdate,

/*step2.a*/

exec dbo.ConvertDB1BtoT100 @extstartdate,

while @r<@numdays
begin
print('Finished day %d - %d ' + cast(getdate() as varchar))
RAISERROR('Finished day %d - %d',0,1, @r,
exec dbo.ConvertDB1BtoT100 @extstartdate,@r;
end

--while @r<@numdays
if not exists (SELECT * FROM sys.indexes WHERE name='IDX_ORIGIN' AND object_id = OBJECT_ID('dbo.Final'))
begin
create index IDX_ORIGIN on Final(origin);
end
create index IDX_HUB on Final(hub);
create index IDX_DEST on Final(DEST);
create index IDX_AIRLINES on Final(airline);
create index IDX_FLNUM_1 on Final(f1_1_num);
create index IDX_FL_1_SCH_DEP on Final(f1_1_sch_dep);
create index IDX_FL_1_SCH_ARR on Final(f1_1_sch_arr);
create index IDX_FLNUM_2 on Final(f1_2_num);
create index IDX_FL_2_SCH_DEP on Final(f1_2_sch_dep);
create index IDX_FL_2_SCH_ARR on Final(f1_2_sch_arr);
create index IDX_T100_MAX_PAX on Final(t100_max_pax);
end

print('Finish step 2' + cast(getdate() as varchar))
RAISERROR('Finished step2b',0,1)
WITH NOWAIT;

-------------------------
print('Step 3' + cast(getdate() as varchar))
RAISERROR('Step3',0,1)
WITH NOWAIT;
declare @rebooked int
declare @ptd int
declare @ptdperfl int
declare @avaseats int
set @rebooked = 0
set @ptd = 0
set @ptdperfl = 0
set @avaseats = 0

declare @pstatus int -- return status for preemptive_percent cursor
declare @spax int
declare @remainpax int
declare @istatus smallint
declare @priority smallint
declare @mpax int
declare @rpax int
declare @rpax_1 int -- rpax for prev day preemptive rebooking
declare @rpax_2 int -- rpax for same day preemptive rebooking
declare @rpax_3 int -- rpax for normal rebooking[dbo].[FinalDec1_2_600]
declare @rpax_orig int
declare @og_airline varchar(10)
declare @og_fl_1_num smallint
declare @og_fl_1_sch_dep smalldatetime
declare @og_fl_1_sch_arr smalldatetime
declare @og_fl_1_act_arr smalldatetime
declare @og_fl_2_num smallint
declare @og_fl_2_sch_dep smalldatetime
declare @og_fl_2_sch_arr smalldatetime
declare @og_fl_2_act_arr smalldatetime
declare @alt_fl_1_num smallint
declare @alt_fl_1_sch_dep smalldatetime
declare @alt_fl_1_sch_arr smalldatetime
declare @alt_fl_1_act_arr smalldatetime
declare @alt_fl_2_num smallint
declare @alt_fl_2_sch_dep smalldatetime
declare @alt_fl_2_sch_arr smalldatetime
declare @alt_fl_2_act_arr smalldatetime

declare @t1 float
declare @totseats1 smallint
declare @avaseats1 smallint
declare @alt_fl_1_act_arr smalldatetime
declare @alt_fl_1_2_num smallint
declare @alt_fl_1_2_sch_dep smalldatetime
declare @alt_fl_1_2_sch_arr smalldatetime
declare @l1f float
declare @t2 float
declare @totseats2 smallint
declare @avaseats2 smallint
declare @alt_fl_1_2_act_arr smalldatetime

drop table ##results if OBJECT_ID('tempdb..##results') is not null

-- table to store the results
create table ##results
(
    runID int, status varchar(3), legs int,
    total_flight_itins int, total_pax int,
    PTD_in_Mins int, trip_delay int, trip_time bigint, sch_time bigint,
    alt_priority int,
    perc_prev float, perc_same float,
    year int, month int, day int
);
-- table of percents, to generate the preemptive percents
create table #percents
(
    perc float
);
insert into #percents
(perc)
values
(0),(0.1),(0.3),(0.5),(0.7);--,(0.9),(1);

declare preemptive_percents cursor fast_forward
for
-- below selects when one of the two days is 0
select a.perc as prev, b.perc as same from #percents a, #percents b
where a.perc=0 or b.perc=0;

-- below selects all combinations
/*select a.perc as prev, b.perc as same from #percents a, #percents b
where a.perc+b.perc <= 1;*/

-- below selects only the preset value
--select @preempt_prev as prev, @preempt_same as same
exec dbo.Create_OptionsBase @startdate, @numdays, 1440, @lf_adj; -- the number is the window, in minutes, to look for alt flights
open preemptive_percents;
fetch next from preemptive_percents into
    @preempt_prev, @preempt_same;
set @pstatus = @@fetch_status;
while @pstatus = 0 begin
    set @r=0;
    while @r < @epoch begin
        set @r = @r + 1;
        print('Create OptionsTemp' + cast(getdate() as varchar))
        RAISERROR('Create OptionsTemp' ,0,1) WITH NOWAIT;
        --exec dbo.Create_OptionsTemp @startdate, @numdays, 1440, @lf_adj; -- the number is the window, in minutes, to look for alt flights
        exec dbo.Create_OptionsTemp;
        print('Create OptionsTemp Temp' + cast(getdate() as varchar) + 'prev=' + cast(@preempt_prev as varchar) + 'same=' + cast(@preempt_same as varchar) + ' r=' + cast(@r as varchar));
        RAISERROR('Start step 3 cursor' ,0,1) WITH NOWAIT;

        declare rebooking cursor forward_only keyset read_only --fast_forward
        for
            select itin_status, origin, dest, Total_Pax_on_Flight, Remaining_pax, Original_Airline, New_Airline,
            Original_Flight_1, Original_Sch_Dep, Original_Flight_1_Sch_Arr, Original_Flight_1_Act_Arr,
            Original_Flight_2, Original_Flight_2_Sch_dep, Original_Sch_Arr,
            Alternate_Flight_1, Alternate_Flight_1_Sch_Dep, Alternate_Flight_1_LF, Alternate_Flight_1_TotalSeats, Alternate_Flight_1_AvaSeats, Alternate_Flight_1_Act_Arr,
Alternate_Flight_2, Alternate_Flight_2_Sch_Dep, Alternate_Flight_2_Sch_Arr, Alternate_Flight_2_LF, Alternate_Flight_2_TotalSeats, Alternate_Flight_2_AvaSeats, Alternate_Flight_2_Act_Arr, Total_Rebooked_Pax, alt_priority from
#options_temp
where alternate_flight_1 is not null or alternate_flight_2 is not null
--where alt_priority > 4
/*order by Original_Flight_1, Original_Sch_Dep, Original_Flight_2, Original_Sch_Arr,
alt_priority, Alternate_Flight_1_Sch_Dep, Alternate_Flight_2_Sch_Arr, Alternate_Flight_2_Sch_Dep*/

-- ****************************new ORDER BY to get reverse order of flights******************
order by
Original_Flight_1, Original_Sch_Dep, Original_Flight_2, Original_Sch_Arr,
--case when alt_priority in (1,2) then alt_priority+2 when alt_priority in (3,4) then alt_priority-2 else alt_priority end,
--For rebooking, the code starts to rebook the pax from original departure time in the past to the first available option:
--case when alt_priority<5 then 
datediff(mi,original_sch_dep,Alternate_Flight_1_Sch_Dep) else 
datediff(mi,original_sch_dep,Alternate_Flight_1_Sch_Dep)+(5*24*60) end asc,
--case when alt_priority<5 then 
datediff(mi,original_sch_dep,Alternate_Flight_2_Sch_Dep) else 
datediff(mi,original_sch_dep,Alternate_Flight_2_Sch_Dep)+(5*24*60) end asc,

--Alternate_Flight_1_Sch_Dep,
priority_metric, alt1dep_metric, alt2dep_metric,
Alternate_Flight_2_Sch_Arr

open rebooking
fetch next from rebooking into
@istatus, @origin, @dest, @pax, @pax_orig, @og_airline, @new_airline,
@og_fl_1_num, @og_fl_1_sch_dep, @og_fl_1_sch_arr, @og_fl_1_act_arr,
@og_fl_2_num, @og_fl_2_sch_dep, @og_fl_2_sch_arr,
@alt_fl_1_num, @alt_fl_1_sch_dep, @alt_fl_1_sch_arr, @lf1, @totseats1, @avaseats1,
@alt_fl_1_act_arr,
@alt_fl_2_num, @alt_fl_2_sch_dep, @alt_fl_2_sch_arr, @lf2, @totseats2, 
@avaseats2, @alt_fl_2_act_arr,
@rebooked, @priority

set @status = @@fetch_status
declare @last_fl_1 int
declare @last_sch_dep smalldatetime
declare @last_fl_2 int
declare @last_sch_arr smalldatetime
declare @last_priority smallint

set @last_fl_1 = NULL
set @last_sch_dep = NULL
set @last_fl_2 = NULL
set @last_sch_arr = NULL
set @last_priority = -1
set @pax_1 = 0;
set @pax_2 = 0;
set @pax_3 = 0;

--set @counter =0
while (@status = 0) begin
/*set @counter = @counter+1
if (@counter % 10000)=0 begin
RAISERROR('done counter step3 steps %d' ,0,1,@counter) WITH
NOWAIT
print('counter step 3' +cast(getdate() as varchar))
*/
begin
...
end*/

-- check if the original itinerary for the current option is different
from the previous one
if (@last_fl_1 is null or (@last_fl_1 != @og_fl_1_num) or (@last_sch_dep != @og_fl_1_sch_dep)
or @last_sch_arr != @og_fl_1_sch_arr)

BEGIN

-- store the computed PTDs into the final table
exec dbo.Update_Itinerary_PTDs @last_fl_1, @last_fl_2, @last_sch_dep, @last_sch_arr;

-- save the previous remaining pax value
if @last_priority < 3 set @rpax = @rpax + @rpax_2;
if @last_priority < 5 set @rpax = @rpax + @rpax_3;
exec dbo.Update_Remaining_Pax @rpax, @last_fl_1, @last_sch_dep,

-- save the remaining pax value for the current itinerary,
including splitting rpax off for preemptive rebooking
-- set @pax_1 = floor(@preempt_prev*@rpax_orig);
exec @pax_1 = dbo.binomial_rand @rpax_orig, @preempt_prev;
-- set @pax_2 = floor(@preempt_same*@rpax_orig);
exec @pax_2 = dbo.binomial_rand @rpax_orig, @preempt_same;
set @pax_3 = @rpax_orig - @pax_1 - @pax_2;
if @pax_3 < 0 begin
    set @pax_2 = @pax_2 + @pax_3;
    set @pax_3 = 0;
end
set @rpax = @pax_1 + @pax_2;
set @last_priority = 0;

END -- if (original itinerary is different than prev original itinerary)

set @last_fl_1 = @og_fl_1_num;
set @last_sch_dep = @og_fl_1_sch_dep;
set @last_fl_2 = @og_fl_2_num;
set @last_sch_arr = @og_fl_2_sch_arr;

-- if the priority class changes to a new new preemptive rebooking
category, add all of the rpax from that category
if @priority in (1, 2) and @last_priority in (-1, 0, 3, 4) and @rpax = @rpax_1
begin
    set @pax_3 = @pax_3 + (@rpax - @pax_1);
    set @rpax = @pax_1;
end

-- if @priority = 3 and @last_priority = 5 set @rpax = @rpax_2;
if @priority = 5 and @last_priority = @priority
set @last_priority = @priority;

set @spax = 0;
set @remainpax = 0;

if istatus = 1 -- FOR ITINS HAVING 1ST FLIGHT AS CANCELLED
begin
    if @rpax > 0
    begin
        set @rebooked = dbo.pairmin(@avaseats1, @avaseats2);
        if @rpax < @rebooked set @rebooked = @rpax;
        set @rpax = @rpax - @rebooked;
    end
end

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if the alt flight is on the same day, then none of the rebooked pax are stranded anymore

```sql
if day(@og_fl_1_sch_dep) = day(@alt_fl_1_sch_dep) or
    @og_fl_1_sch_dep = @og_fl_1_sch_dep
    set @spax = -@rebooked;
set @remainpax = -@rebooked;
```

--/ direct to direct option
```sql
if (@og_fl_2_num is NULL and @alt_fl_2_num is NULL)
    set @ptd =
        datediff(mi,@og_fl_1_sch_arr,@alt_fl_1_act_arr) -15;
--/direct to connecting option
else if (@og_fl_2_num is NULL and @alt_fl_2_num is not NULL)
    set @ptd =
        datediff(mi,@og_fl_1_sch_arr,@alt_fl_2_act_arr)
            -15;
```

--/ Connecting to Direct Option
```sql
else if (@og_fl_2_num is not NULL and @alt_fl_2_num is NULL)
    set @ptd =
        datediff(mi,@og_fl_2_sch_arr,@alt_fl_1_act_arr)
            -15;
```

--/ Connecting to Connecting
```sql
else if (@og_fl_2_num is not NULL and @alt_fl_2_num is not NULL)
    set @ptd =
        datediff(mi,@og_fl_2_sch_arr,@alt_fl_2_act_arr) -15;
```

if (@ptd >0)  set @ptd = @ptd * @rebooked;
else set @ptd = 0;

set @avaseats1 = @avaseats1 - @rebooked;
set @avaseats2 = @avaseats2 - @rebooked;
set @lf1 = 1 -
convert(float,(1.0*@avaseats1)/@totseats1);
set @lf2 = 1 -
convert(float,(1.0*@avaseats2)/@totseats2);
```

--/ Need to assume that the same flight number doesn't repeat in a day--wont work for more than one day
--/ Need to adjust the load factors so that the available seats gets adjusted accordingly
```
```
```sql
end--//--if rpax > 0
else set @ptd = 0;
```

exec dbo.Update_Rebook_PTD @rebooked, @ptd, @og_fl_1_sch_dep,
@og_fl_1_sch_dep, @alt_fl_1_sch_dep, @og_fl_1_sch_dep, @og_fl_1_sch_dep,
og_fl_2_num,og_fl_2_sch_arr,alt_fl_2_num,alt_fl_1_sch_arr;
exec dbo.Update_Avaseats2
@avaseats2, @lf2, @alt_fl_2_num, @alt_fl_2_sch_arr;
exec dbo.Update_Avaseats1
@avaseats1, @lf1, @alt_fl_1_num, @alt_fl_1_sch_dep;
```

end -- // if istatus == 1
else if (@istatus = 2 or @istatus = 0)
begin
if (@rpax >0 )
begin
set @rebooked = @avaseats2;
if @rpax < @rebooked set @rebooked = @rpax;
set @rpax = @rpax - @rebooked;
```

-- if flight 2 and it's alternate are on the same day, none of the rebooked pax are stranded anymore
```
if day(@og_fl_1_2_sch_dep)=day(@alt_fl_1_2_sch_dep) or
    set @spax = -@rebooked;
set @remainpax = -@rebooked;

set @ptd = datediff(mi,@og_fl_2_sch_arr,@alt_fl_2_act_arr)-15;
if (@ptd>0) set @ptd = @ptd * @rebooked;
else set @ptd = 0;

set @avaseats2 = @avaseats2 - @rebooked;
set @lf2 = 1 - convert(float,(1.0*@avaseats2)/@totseats2);
--- Need to assume that the same flight number doesn't
repeat in a day--wont work for more than one day
--- Need to adjust the load factors so that the available
seats gets adjusted accordingly
end -- if rpax > 0
else set @ptd = 0;
exec dbo.Update_Rebook_PTD @rebooked, @ptd, @og_fl_1_num,
@og_fl_1_sch_dep, null, null,
@og_fl_2_num,@og_fl_1_sch_arr,@alt_fl_1_2_sch_arr;
exec dbo.Update_Avaseats2 @avaseats2,@lf2,@alt_fl_1_2_sch_arr;
end-- if (@istatus = 2 or @istatus = 0)

/*
   if @remainpax != 0
    update #Stranded_pax
    set Stranded_pax = Stranded_pax+spax,
    Remaining_pax = Remaining_pax+@remainpax
    where year=year(@og_fl_1_sch_dep) and
    month=month(@og_fl_1_sch_dep) and
    day=day(@og_fl_1_sch_dep);
*/
fetch next from rebooking into
@istatus, @origin, @dest, @mpax, @rpax_orig, @og_airline,
@og_fl_1_num, @og_fl_1_sch_dep, @og_fl_1_sch_arr,
@og_fl_2_num,@og_fl_1_sch_arr,@alt_fl_1_2_sch_arr,
@alt_fl_1_num, @alt_fl_1_sch_dep, @lf1, @toseats1, @avaseats1,
@toseats2, @avaseats2, @alt_fl_1_2_sch_arr,
@rebooked,@priority
set @status = @@fetch_status
end -- end while (@status = 0)
close rebooking
deallocate rebooking
exec dbo.Update_Itinerary_PTDs @last_fl_1,
@last_fl_2,@last_sch_dep,@last_sch_arr;
--if @last_priority<3 set @rpax = @rpax+@rpax_2;
if @last_priority<5 set @rpax = @rpax+@rpax_3;
exec dbo.Update_Remaining_Pax @rpax, @last_fl_1, @last_sch_dep, @last_fl_2,
@last_sch_arr;
exec dbo.Update_DelDiv 180;

exec dbo.Collect_results @preempt_prev, @preempt_same, @r;
end -- while @r < ... loop

fetch next from preemptive_percents into
@preempt_prev, @preempt_same;

set @pstatus = @@fetch_status;
end -- while @pstatus = 0

close preemptive_percents
deallocate preemptive_percents
drop table #percents
if OBJECT_ID('tempdb..##options_temp') is not null drop table ###options_temp;

print('Finished Step 3' +cast(getdate() as varchar))
RAISERROR('Finished step 3',0,1) WITH NOWAIT;
--select * from ##Stranded_pax
select * from ##results --where day = 27
/*clean up*/
/*
IF OBJECT_ID('dbo.Final', 'U') IS NOT NULL
DROP TABLE dbo.Final;
select * from Final*/

drop table DB1B;
drop table MasterTable
drop table Final;
drop table #options_temp;
drop table ##Stranded_pax
drop table #ontime_lf;
drop table ##aotp;
--select * from ##Stranded_pax
--select * into results10 from ##results

28) Statistics

use PaxDelay

select distinct year,month,day,status, legs, perc_prev, perc_same, alt_priority,
avg(cast(PTD_in_mins as float)) as median_ptd into #median_ptd
from
(
    select status, legs, perc_prev, perc_same, alt_priority, PTD_in_mins,
    year,month,day,n, max(n) over (partition by status, legs, perc_prev,
    perc_same,alt_priority, year,month,day) mxn from
    (SELECT status,legs,perc_prev,perc_same,alt_priority,
    PTD_in_mins,year,month,day,
    ROW_NUMBER() OVER(PARTITION BY status,legs,perc_prev,perc_same,alt_priority, year,month,day ORDER BY PTD_in_mins) AS n
    FROM ##results
    ) a
where (mxn%2-1 and 2^n=(maxn+1)) or (maxn%2=0 and 2^n=maxn) or (maxn%2=0 and 2^(n-1)=maxn)
group by status, legs, perc_prev, perc_same,alt_priority, year,month,day;
select distinct year,month,day,status, legs, perc_prev, perc_same, alt_priority, avg(cast(trip_delay as float)) as median_trip_delay from #median_trip_delay
from (
    select status, legs, perc_prev, perc_same, alt_priority, trip_delay, year,month,day,n, max(n) over (partition by status, legs, perc_prev, perc_same,alt_priority, year,month,day) maxn from
    (SELECT status,legs,perc_prev,perc_same, alt_priority, trip_delay,year,month,day,
      row_number() over(partition by status,legs,perc_prev,perc_same,alt_priority, year,month,day) AS a
    ) a
where (maxn%2=1 and 2*n=(maxn+1)) or (maxn%2=0 and 2*n=maxn) or (maxn%2=0 and 2*(n-1)=maxn)
group by status, legs, perc_prev, perc_same,alt_priority, year,month,day;

select distinct year,month,day,status, legs, perc_prev, perc_same, alt_priority, avg(cast(trip_time as float)) as median_trip_time from #median_trip_time
from (
    select status, legs, perc_prev, perc_same, alt_priority, trip_time, year,month,day,n, max(n) over (partition by status, legs, perc_prev, perc_same,alt_priority, year,month,day) maxn from
    (SELECT status,legs,perc_prev,perc_same, alt_priority, trip_time,year,month,day,
      row_number() over(partition by status,legs,perc_prev,perc_same,alt_priority, year,month,day) AS a
    ) a
where (maxn%2=1 and 2*n=(maxn+1)) or (maxn%2=0 and 2*n=maxn) or (maxn%2=0 and 2*(n-1)=maxn)
group by status, legs, perc_prev, perc_same,alt_priority, year,month,day;

select distinct year,month,day,status, legs, perc_prev, perc_same, alt_priority, avg(cast(sch_time as float)) as median_sch_time from #median_sch_time
from (
    select status, legs, perc_prev, perc_same, alt_priority, sch_time, year,month,day,n, max(n) over (partition by status, legs, perc_prev, perc_same,alt_priority, year,month,day) maxn from
    (SELECT status,legs,perc_prev,perc_same, alt_priority, sch_time,year,month,day,
      row_number() over(partition by status,legs,perc_prev,perc_same,alt_priority, year,month,day) AS a
    ) a
where (maxn%2=1 and 2*n=(maxn+1)) or (maxn%2=0 and 2*n=maxn) or (maxn%2=0 and 2*(n-1)=maxn)
group by status, legs, perc_prev, perc_same,alt_priority, year,month,day;

select distinct year,month,day,status, legs, perc_prev, perc_same, alt_priority, avg(cast(total_pax as float)) as median_tpax from #median_tpax
from (
    select status, legs, perc_prev, perc_same, alt_priority, total_pax,year,month,day,n, max(n) over (partition by status, legs, perc_prev, perc_same,alt_priority, year,month,day) maxn from
    (SELECT status,legs,perc_prev,perc_same, alt_priority, total_pax,year,month,day,
      row_number() over(partition by status,legs,perc_prev,perc_same,alt_priority, year,month,day) AS a
    ) a
where (maxn%2=1 and 2*n=(maxn+1)) or (maxn%2=0 and 2*n=maxn) or (maxn%2=0 and 2*(n-1)=maxn)
group by status, legs, perc_prev, perc_same,alt_priority, year,month,day;
```sql
SELECT status, legs, perc_prev, perc_same, alt_priority,
total_pax, year, month, day,
    ROW_NUMBER() OVER(PARTITION BY status, legs, perc_prev, perc_same, alt_priority, year, month, day ORDER BY total_pax) AS n
FROM
)
a
)
)
)
)
)

WHERE (maxn%2=1 and 2^n=(maxn+1)) or (maxn%2=0 and 2^n=maxn) or (maxn%2=0 and 2^(n-1)=maxn)
group by status, legs, perc_prev, perc_same, alt_priority, year, month, day;

--The code is averaging results from AllSteps (binomial distribution ~25 runs)
select a.year, a.month, a.day, a perc_prev, a perc_same, a status, a legs, a alt_priority,
    avg(cast(a.total_pax as float)) as total_pax,
    stdev(cast(a.total_pax as float)) as stdev_total_pax,
    avg(cast(a.PTD_in_mins as float)) as mean_ptd,
    stdev(cast(a.PTD_in_mins as float)) as stdev_ptd,
    avg(cast(a.trip_delay as float)) as mean_trip_delay,
    stdev(cast(a.trip_delay as float)) as stdev_trip_delay,
    avg(cast(a.trip_time as float)/60 as mean_trip_time)
) as stdev_trip_time,
    avg(cast(a.sch_time as float)/60 as mean_sch_time)
) as stdev_sch_time,
    y.median_sch_time/60 as median_sch_time

from #results a, #median_tpax n, #median_ptd p, #median_trip_delay t, #median_trip_time x, #median_sch_time y

where
    a.status=n status and a.legs=n.legs and a perc_prev=n perc_prev and
    a perc_same=n perc_same and a year=n year and a month=n month and a day=n day and
    a alt_priority=n alt_priority
    and a status=p status and a legs=p legs and a perc_prev=p perc_prev and
    a perc_same=p perc_same and a year=p year and a month=p month and a day=p day and
    a alt_priority=p alt_priority
    and a status=t status and a legs=t legs and a perc_prev=t perc_prev and
    a perc_same=t perc_same and a year=t year and a month=t month and a day=t day and
    a alt_priority=t alt_priority
    and a status=x status and a legs=x legs and a perc_prev=x perc_prev and
    a perc_same=x perc_same and a year=x year and a month=x month and a day=x day and
    a alt_priority=x alt_priority
    and a status=y status and a legs=y legs and a perc_prev=y perc_prev and
    a perc_same=y perc_same and a year=y year and a month=y month and a day=y day and
    a alt_priority=y alt_priority

group by a status, a legs, a perc_prev, a perc_same, a alt_priority, a year, a month, a day,
    n.median_tpax, p.median_ptd,
    t.median_trip_delay, x.median_trip_time, y.median_sch_time
order by a.year, a.month, a.day, a perc_prev, a perc_same, a legs, a status desc;

/*
select * from #summary
order by year, month, day, perc_same, legs, status desc;
*/
drop table #median_ptd;
drop table #median_trip_delay;
drop table #median_trip_time;
drop table #median_sch_time;
drop table #median_tpax;

select year, month, day, perc_same, perc_prev,
    min(rpSeats) as min_rpSeats, max(rpSeats) as max_rpSeats,
    min(rpLits) as min_rpLits, max(rpLits) as max_rpLits,
    min(prevpax) as min_prevpax, max(prevpax) as max_prevpax,
    min(samepax) as min_samepax, max(samepax) as max_samepax,
    min(latepax) as min_latepax, max(latepax) as max_latepax;

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\[
\begin{align*}
\text{min}(\text{nextpax}) & \text{ as } \text{min\_nextpax}, \\
\text{max}(\text{nextpax}) & \text{ as } \text{max\_nextpax}
\end{align*}
\]

\[
\text{into } \#\text{minmax}
\]

\[
\text{from (}
\]

\[
\text{select year, month, day, perc\_same, perc\_prev, runid,}
\]

\[
\sum(\text{if}(\text{alt\_priority} \in (1,2), \text{total\_pax}, 0)) \text{ as } \text{prevpax},
\]

\[
\sum(\text{if}(\text{alt\_priority} \in (3,4), \text{total\_pax}, 0)) \text{ as } \text{samepax},
\]

\[
\sum(\text{if}(\text{alt\_priority} \in (5,6), \text{total\_pax}, 0)) \text{ as } \text{latепax},
\]

\[
\sum(\text{if}(\text{alt\_priority} \in (7,8), \text{total\_pax}, 0)) \text{ as } \text{nextpax},
\]

\[
\sum(\text{if}(\text{alt\_priority} \in (0), \text{total\_pax}, 0)) \text{ as } \text{rpSeats},
\]

\[
\sum(\text{if}(\text{alt\_priority} \in (-1), \text{total\_pax}, 0)) \text{ as } \text{rpItins}
\]

\[
\text{from } \#\#\text{results}
\]

\[
\text{where status='CNX'}
\]

\[
\text{group by year, month, day, perc\_same, perc\_prev, runid}
\]

\[
) \text{ a}
\]

\[
\text{group by year, month, day, perc\_same, perc\_prev}
\]

\[
\text{select * from (}
\]

\[
\text{select year, month, day, 'same' as pretype, perc\_same as perc,}
\]

\[
\text{min\_rpSeats, max\_rpSeats,}
\]

\[
\text{min\_rpItins, max\_rpItins,}
\]

\[
\text{min\_prevpax, max\_prevpax,}
\]

\[
\text{min\_samepax, max\_samepax,}
\]

\[
\text{min\_latепax, max\_latепax,}
\]

\[
\text{min\_nextpax, max\_nextpax}
\]

\[
\text{from } \#\text{minmax}
\]

\[
\text{where perc\_prev=0}
\]

\[
\text{union}
\]

\[
\text{select year, month, day, 'prev' as pretype, perc\_prev as perc,}
\]

\[
\text{min\_rpSeats, max\_rpSeats,}
\]

\[
\text{min\_rpItins, max\_rpItins,}
\]

\[
\text{min\_prevpax, max\_prevpax,}
\]

\[
\text{min\_samepax, max\_samepax,}
\]

\[
\text{min\_latепax, max\_latепax,}
\]

\[
\text{min\_nextpax, max\_nextpax}
\]

\[
\text{from } \#\text{minmax}
\]

\[
\text{where perc\_prev=0}
\]

\[
) \text{ a}
\]

\[
\text{where day=18}
\]

\[
\text{order by year, month, day, pretype, perc;}
\]

\[
\text{drop table } \#\text{minmax}
\]

\[
\text{select * into } \#\text{ancsummary}
\]

\[
\text{from (}
\]

\[
\text{select year, month, day, 'same' as pretype, perc\_same as perc,}
\]

\[
\text{sum(\text{total\_pax}) as totpax, sum(\text{stdev\_total\_pax}) as sd\_totpax,}
\]

\[
\text{sum(\text{if}(\text{alt\_priority} \in (1,2), \text{total\_pax}, 0)) as } \text{prevpax,}
\]

\[
\text{sum(\text{if}(\text{alt\_priority} \in (3,4), \text{total\_pax}, 0)) as } \text{samepax,}
\]

\[
\text{sum(\text{if}(\text{alt\_priority} \in (5,6), \text{total\_pax}, 0)) as } \text{latепax,}
\]

\[
\text{sum(\text{if}(\text{alt\_priority} \in (7,8), \text{total\_pax}, 0)) as } \text{nextpax,}
\]

\[
\text{sum(\text{if}(\text{alt\_priority} \in (0), \text{total\_pax}, 0)) as } \text{rpSeats,}
\]

\[
\text{sum(\text{if}(\text{alt\_priority} \in (-1), \text{total\_pax}, 0)) as } \text{rpItins}
\]

\[
\text{from } \#\text{ancsummary}
\]

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```sql
sum(mean_trip_time) as trip_time, sum(stddev_trip_time) as sd_trip_time,
sum(mean_sch_time) as sch_time, sum(stddev_sch_time) as sd_sch_time
from #summary
where perc_prev=0 and status='CNX'
group by year, month, day, perc_same
union

select
  year, month, day,
  'prev' as pretype, perc_prev as perc,
  sum(total_pax) as totpax, sum(stdev_total_pax) as sd_totpax,
  sum(iif(alt_priority in (1,2),total_pax,0)) as prevpax,
  sum(iif(alt_priority in (1,2),stdev_total_pax,0)) as sd_prevpax,
  sum(iif(alt_priority in (3,4),total_pax,0)) as samepax,
  sum(iif(alt_priority in (3,4),stdev_total_pax,0)) as sd_samepax,
  sum(iif(alt_priority in (5,6),total_pax,0)) as latepax,
  sum(iif(alt_priority in (5,6),stdev_total_pax,0)) as sd_latepax,
  sum(iif(alt_priority in (7,8),total_pax,0)) as nextpax,
  sum(iif(alt_priority in (7,8),stdev_total_pax,0)) as sd_nextpax,
  sum(iif(alt_priority=1,total_pax,0)) as rpax,
  sum(iif(alt_priority=0,total_pax,0)) as sd_rpax,
  sum(iif(alt_priority=0,total_pax,0)) as sd_rpaxNoSeat,
  sum(iif(alt_priority=0,total_pax,0)) as sd_rpaxNoItins,
  sum(total_pax) as total_pax,
  sum(mean_ptd) as ptd, sum(stdev_ptd) as sd_ptd,
  sum(mean_trip_time) as trip_time, sum(stdev_trip_time) as sd_trip_time,
  sum(mean_sch_time) as sch_time, sum(stdev_sch_time) as sd_sch_time
from #summary
where perc_same=0 and status='CNX'
group by year, month, day, perc_prev
)a

/

/*
select
  year, month, day, pretype, perc, null as lf_bias,
  100*(prepax/totpax as perc_prepax, 100*(totpax-prepax-rpax)/totpax as perc_latepax, null as perc_overnight, ptd/(60*24) as paxdelay_days,
  377*rpax as rev_recoup,
  250*spax/2 as indirect_cost,
  sch_time/24 as sch_trip_time_days,
  trip_time/24 as act_trip_time_days
from #cancsummary
order by year,month,day,pretype desc,perc
*/

/*
select
  year, month, day, pretype, perc,
  totpax, prepax, samepax, latepax, nextpax, rpax,
  377*(prevpax+samepax) as rev_recoup,
  ptd as tot_ptd, nextpax as overnight, rpax as unbooked, 250*(nextpax+rpax) as ind_cost
from #cancsummary
where day =20
order by year,month,day,pretype desc,perc
select
  year, month, day, pretype, perc,
  (prevpax+samepax+latepax+nextpax) as num_rebooked_mean,
  (sd_prepax+sd_samepax+sd_latepax+sd_nextpax) as num_rebooked_sdev,
  (prevpax+samepax) as num_early_mean, (sd_prepax+sd_samepax) as num_early_sdev,
  rpaxNoSeat, sd_rpaxNoSeat, rpaxNoItins, sd_rpaxNoItins,
```
```sql
nextpax as overnight_pax,
377*(prevpax+samepax) as rev_recoup,
ptd/(24.0*60.0) as ptd_days_mean, sd_ptd/(24*60) as ptd_days_sdev,
ptd/(1.0*totpax) as avg_ptd_min
from #cancsummary
where day =20
order by year,month,day,pretype desc,perc
*/
/*
select
  pretype, perc,
  sum(totpax) as totpax, sum(prevpax) as prevpax, sum(samepax) as samepax,
  sum(latepax) as latepax, sum(nextpax) as nextpax, sum(rpax) as rpax,
  377*sum(prevpax+samepax) as rev_recoup,
  sum(ptd) as totpax, sum(nextpax) as overnight, sum(rpax) as unbooked,
  250*sum(nextpax+rpax) as ind_cost
from #cancsummary
where day in (25,26)
group by pretype, perc
order by pretype desc,perc
*/
select pretype, perc,
  PercPrepax, PercSamepax, PercLate, PercNext, PercRPAX, CorpTravCost,
  rev_recoup, num_rebooked_mean, num_rebooked_sdev, num_early_mean, num_early_sdev,
  overnight_pax, rev_recoup, ptd_days_mean, ptd_days_sdev, ptd_min/totpax as avg_ptd_min
from (
  select
    pretype, perc,
    sum(prevpax+samepax+latepax+nextpax) as num_rebooked_mean,
    sum(prevpax+samepax) as prevpax+samepax,
    sum(prevpax+samepax+latepax+nextpax) as num_rebooked_sdev,
    sum(samepax) as samepax,
    sum(latepax) as latepax,
    sum(nextpax) as nextpax,
    sum(rpax) as rpax,
    377*sum(prevpax+samepax) as rev_recoup,
    sum(totpax) as totpax,
    sum(totpax) as totpax,
    sum(totpax) as totpax,
    sum(totpax) as totpax,
    sum(totpax) as totpax,
  )
from #cancsummary
where day = 18 --in (25,26)
group by pretype, perc
) a
order by pretype desc,perc

select avg(ptpax) as tptax, legs from (
  select runID, legs, sum(total_pax) as tpa from ##results where perc_prev=0 and perc_same=0
  and status='CNX' and day=18 group by runID, legs) a group by legs

select * from #summary where day=18
select * from #cancsummary where day=18
drop table #cancsummary
drop table #summary;
select min(1f) as minlfCNXALL, max(1f) as maxlfCNXALL, avg(1f) avglfCNXALL from ##ontime_lf
where day(sch_dep) = 18 and cancelled =1
select min(1f) as minlfALL, max(1f) as maxlfALL, avg(1f) avglfALL from ##ontime_lf
where day(sch_dep) = 18
select min(1f) as minlfORCNX, max(1f) as maxlfORCNX, avg(1f) avglfORCNX from ##ontime_lf
where day(sch_dep) = 18 and (origin = 'EWR' or dest='EWR') and cancelled =1
```
```sql
select min(1f) as min1fOR, max(1f) as max1fOR, avg(1f) avg1fOR from #ontime_1f where day(sch_dep) = 18 and (origin = 'EWR' or dest='EWR')
select sum(1f50) as CNXlf50ALLCXC, sum(1f50to60) as CNXlf50to60ALLCXC, sum(1f60to70) as CNXlf60to70ALLCXC, sum(1f70to80) as CNXlf70to80ALLCXC, sum(1f80to90) as 1f80to90ALLCXC,
sum(1f90) as 1f90ALLCXC from
(
select
CASE WHEN 1f <=0.50 THEN 1 ELSE 0 END AS 1f50,
CASE WHEN 1f > 0.50 and 1f <=0.60 then 1 else 0 end as 1f50to60,
case when 1f > 0.60 and 1f <=0.70 then 1 else 0 end as 1f60to70,
case when 1f > 0.70 and 1f <=0.80 then 1 else 0 end as 1f70to80,
case when 1f>0.80 and 1f <=0.90 then 1 else 0 end as 1f80to90,
case when 1f>0.90 then 1 else 0 end as 1f90
from #ontime_1f where day(sch_dep) = 18 and cancelled = 1
)a
select sum(1f50) as 1f50ALL, sum(1f50to60) as 1f50to60ALL, sum(1f60to70) as 1f60to70ALL, sum(1f70to80) as 1f70to80ALL, sum(1f80to90) as 1f80to90ALL, sum(1f90) as 1f90ALL from
(
select
CASE WHEN 1f <=0.50 THEN 1 ELSE 0 END AS 1f50,
CASE WHEN 1f > 0.50 and 1f <=0.60 then 1 else 0 end as 1f50to60,
case when 1f > 0.60 and 1f <=0.70 then 1 else 0 end as 1f60to70,
case when 1f > 0.70 and 1f <=0.80 then 1 else 0 end as 1f70to80,
case when 1f>0.80 and 1f <=0.90 then 1 else 0 end as 1f80to90,
case when 1f>0.90 then 1 else 0 end as 1f90
from #ontime_1f where day(sch_dep) = 18
)a
```

```sql
select sum(1f50) as CNXlf50ORDC, sum(1f50to60)as CNXlf50to60ORDC, sum(1f60to70) as CNXlf60to70ORDC, sum(1f70to80) as CNXlf70to80ORDC, sum(1f80to90) as 1f80to90ORDC,
sum(1f90) as 1f90ORDC from
(
select
CASE WHEN 1f <=0.50 THEN 1 ELSE 0 END AS 1f50,
CASE WHEN 1f > 0.50 and 1f <=0.60 then 1 else 0 end as 1f50to60,
case when 1f > 0.60 and 1f <=0.70 then 1 else 0 end as 1f60to70,
case when 1f > 0.70 and 1f <=0.80 then 1 else 0 end as 1f70to80,
case when 1f>0.80 and 1f <=0.90 then 1 else 0 end as 1f80to90,
case when 1f>0.90 then 1 else 0 end as 1f90
from #ontime_1f where day(sch_dep) = 18 and (origin = 'EWR' or dest='EWR') and cancelled = 1
)a
```

```sql
select sum(1f50) as 1f50ORD, sum(1f50to60)as 1f50to60ORD, sum(1f60to70) as 1f60to70ORD, sum(1f70to80) as 1f70to80ORD, sum(1f80to90) as 1f80to90ORD, sum(1f90) as 1f90ORD from
(
select
CASE WHEN 1f <=0.50 THEN 1 ELSE 0 END AS 1f50,
CASE WHEN 1f > 0.50 and 1f <=0.60 then 1 else 0 end as 1f50to60,
case when 1f > 0.60 and 1f <=0.70 then 1 else 0 end as 1f60to70,
case when 1f > 0.70 and 1f <=0.80 then 1 else 0 end as 1f70to80,
case when 1f>0.80 and 1f <=0.90 then 1 else 0 end as 1f80to90,
case when 1f>0.90 then 1 else 0 end as 1f90
from #ontime_1f where day(sch_dep) = 18 and (origin = 'EWR' or dest='EWR')
)a
```
REFERENCES


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Raiteri, A (2015) personal conversation


Sanja Avramovic received her Bachelor of Arts and Masters of Science from University of Belgrade, Serbia in 2009. She was employed in Serbian Broadcasting Corporation for 12 years. After she moved to the U.S. she worked in ArchLab, George Mason University as a GRA on flying simulators, and after that in The Green Technology Group as a data analyst on the contract with Department of Veterans Affairs, working with Big Databases and performing data mining. She is currently working as a GRA in the Department of Health Administration and Policy GMU on the development of CDE ontology and SEER database.