## Airport Gate Assignment Optimization

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#### Abstract

Airport management comprises of a number of complex operations such as security, passenger check-ins, baggage management, taxiing, gate operations, retail, etc. These operations have many touchpoints which can be categorize into three distinct but related stakeholders - the airport operator, the airlines and the passengers. These multiple operations can affect the experience and financials of these stakeholders. Our research will focus on one specific operation, the optimization of gate management. The primary focus of the project is to determine how the passengers are affected by delays and cancellations, and develop a mathematical model based on data analysys in order to maximize the number of passengers served through gates.

## Airport Gate Assignment Optimization

#### Introduction

An international airport is the most populated and busiest travel location to move people around the world. Gate assignment is an imperative task for an airports ground operations; and the search for an optimal assignment of flights to gates is of great importance to eliminate delays, maintain operations with high efficiency and reasonable arrangement, and to maximize profits. It is significant to notice that flight delays and cancellations have a negative impact on airport performance and do not provide an excellent experience to travelers. This project aims to propose a new airport gate assignment method to effectively improve the comprehensive operations capacity and efficiency. The project will also aid in optimizing the number of gates allocated to airline carriers to improve the travelers experience.

Airplanes are parked at gates (in the apron) and where passengers wait to board and disembark from these planes. Airports managed gates though a negotiation with the airline carrier who request or advise airports to set aside a certain number of gates (called gate allocations) for dates in the upcoming seasons. In North America, seasons are divided between the summer months (May to Oct) and winter months (Nov to April). These negotiations are based on these two seasons. Airports prepare gate allocations based on the relationship with the airlines, their flight history, their delay or cancellation profiles, etc. Airports will analyze the data, assign the appropriate number of gates and advise the airline of their gate allocations. Airlines will develop their flight route schedules which will be made available to the public through their websites, travel sites, T.V. commercials, etc.; and Airports will do their resource planning according to the airlines schedule. Airports budget for personnel, get the necessary systems or services for systems, even spend for new facilities that need to be built around this schedule if needed with a CAPEX program. Sometimes, an airport may shuffle airlines around gates and when other airlines are affected, airports will work with these airlines.

## Competitive Analysis

The area of airport gates assignment optimization is well researched. Several researches were aiming to solve the problem considering multiple factors, like cost, passenger walking distances, networking of connected flights etc.

In one research (Azmi, Razali, Ibrahim, Aziz, & Aziz, 2016), in order to solve discrete optimization problem, the SKF algorithm is combined with angle modulated approach. Airport gate allocation problem refers to the search for optimal assignment of flights to gates at an airport. Assignment of flight to gates has become very complex nowadays, especially for a big size airport. In this study, the airport gate allocation problem is solved using a recently introduced angle modulated simulated Kalman filter (AMSKF). The objective of this study is to minimize the total walking distance. A small case study with 14 flights and 16 gates has been chosen. Preliminary results show that SKF is a promising algorithm for solving the airport gate allocation problem.

Another paper (Barnhart, Fearing, Odoni, & Vaze, 2012) summarizes research trends and opportunities in the area of managing air transportation demand and capacity. Capacity constraints and resulting congestion and low schedule reliability currently impose large costs on airlines and their passengers. Significant capacity increases that would solve these problems are not expected in the near- or medium-term. This paper outlines first a number of directions for effecting improvement through marginal capacity increases and better management of demand and available capacity. It describes strategic initiatives that airlines and civil aviation authorities might undertake over time horizons of months to years as well as tactical measures that may be adopted on a daily basis in response to dynamic, real-time developments like poor weather or schedule disruptions. Research challenges in these areas are identified and classified in terms of specifying, allocating, and utilizing capacity. The first two categories reject challenges faced by infrastructure providers, the last category challenges faced by airlines.

Others (Bihr, 1990) think that positioning of incoming flights should take into account

the distribution of passengers among connecting flights. The larger the number of passengers arriving on flight X for destination A, the closer flight X should be positioned to the flight departing for A. This inquiry looks conceptually at this problem, reducing the distance passengers must walk from gate to gate. The criterion selected is minimization of the total passenger-distance travel for a given arrival-departure cycle. The problem can be conveniently cast as a 0,1 Linear Programming (LP) problem].

Another paper (Ding, Lim, Rodrigues, & Zhu, n.d.) studies the over-constrained Airport Gate Assignment Problem, where the objectives are to minimize the number of ungated flights and the total walking distances or connection times. They approach the problem by first using a greedy algorithm to minimize ungated flights and by developing exchange moves which then facilitate the use of heuristics. Simulated Annealing and a hybrid of Simulated Annealing and Tabu Search are used.

Airport gate assignment is of great importance in airport operations. In the paper (Li, n.d.), the authors studied the Airport Gate Assignment Problem (AGAP), propose a new model and implement the model with Optimization Programming language (OPL). With the objective to minimize the number of conflicts of any two adjacent aircrafts assigned to the same gate, the authors build a mathematical model with logical constraints and the binary constraints, which can provide an efficient evaluation criterion for the Airlines to estimate the current gate assignment. To illustrate the feasibility of the model the authors construct experiments with the data obtained from Continental Airlines, Houston Gorge Bush Intercontinental Air-port IAH, which indicate that our model is both energetic and effective. Moreover, the authors interpret experimental results, which further demonstrate that the proposed model can provide a powerful tool for airline companies to estimate the efficiency of their current work of gate assignment.

To help researched see how airports are performing with FlightStats monthly On-time Performance Report, the report ("Airline on-time performance reports," n.d.) compares the on-time arrival performance of major and regional airports around the world. The ("Data

elements," n.d.) report presents Load Factor (passenger-miles as a proportion of available seat-miles in percent) for All Carriers and All Airports. Optimizing gate assignments at airport terminals is analysed in this article ("Optimizing gate assignments at airport terminals," 1998). Airline On-Time Statistics and Delay Causes is presented here ("OST\_R BTS Title from h2," n.d.).

The research (Riedel, 2006) presents an approach for predicting operational performance of airlines on the basis of flight schedules and aircraft assignments. The methodology uses aggregate measures of properties of aircraft assignments, called Aircraft Assignment Key Performance Indicators (KPIs), and aims to find correlations between them and the operational performance of the airline. A simulation experiment is prepared to gather a large set of data points for analysis. A motivation is given for the use of control theoretic approaches in airline operations to utilize the KPIs as a basis for initial planning and corrective actions.

#### Proposed Plan

The focus of this project will be on one United States (US) major airports - either

John F. Kennedy Airport in New York ("John f. Kennedy international airport," 2018) or

San Francisco Airport (SFO) ("San francisco international airport," 2018) depending on the
data availability. The study will analyze airline gate operations performance measures over a
1997 to 2008 of publicly available data from bureau of transportation statistics. JFK Airport
is located in the neighborhood of Jamaica in the borough of Queens, in New York, 30km (16

Miles) southeast of Midtown Manhattan. One of the busiest international air passenger
airports in North America and the sixth busiest airpots in United States with close to 60

million passengers using the facility a year (2016 data). The airport features six passenger
terminals and four runways (see Figure 1). More than ninty airlines operate at JFK Airport
using the more than 128 gates across six terminals. San Francisco International Airport
(SFO) is an international airport 21 km (13 miles) south of downtown San Francisco,

California, United States, near Millbrae and San Bruno in unincorporated San Mateo County. It has flights to points throughout North America and is a major gateway to Europe and Asia. SFO is the largest airport in Northern California and the second-busiest in California, after Los Angeles International Airport. In 2017, it was the seventh busiest in the United States and the twenty-forth busiest airport in the world by passenger count. SFO has four terminals (1, 2, 3, and International) and seven concourses (Boarding Areas A through G) with a total of 115 gates arranged alphabetically in a counterclockwise ring. (see Figure 2)

The primary focus is to determine how the passengers are affected by delays and cancellations, and develop a mathematical model based on data analysys in order to maximize the number of passengers served through gates.

### **Data Insights**

Information from 22 years has been retrieved from airlines transportation database. These datasets are from 1987 to 2008 and are relative to United State airlines performance, for the data description (see Table 1). As an example, for the year 2008, JFK airport has 118.804 flights from 11 different carriers. The top 4 most important carriers are JetBlue Airways with 45% of total traffic followed by Comair Inc. (16%), Delta Air Lines (14%) and American Airlines Inc. (10%). Number of flights per carrier at New York John F. Kennedy Airport (JFK) presented here (see Tables 2,3 and 5). For that analysis the following datasets were analyzed:

- Airline on-time performance reports ("Airline on-time performance reports," n.d.);
- 1987-2008 US Flights statistics ("Data elements," n.d.);
- SFO Gate and Stand Assignment Information ("Sfo gate and stand assignment information data. Gov," n.d.);
- 2015 US Flights statistics ("2015 flight delays and cancellations kaggle," n.d.);
- Aircraft registry ("Aircraft registry releasable aircraft database download," n.d.).

#### Model estimation

Our research will focus on the impact of this delays and cancellations to the airport.

What our research goal is to answer the following questions:

- What are the impacts of airline delays on gate management specifically on the following:
  - What is the current baseline of delay and cancelled events in terms of number of instances, flights and instances (Quantify and create a baseline)
  - Which airlines cause these delays?
  - What are the causes of these delays?
  - What is the impact of these delays into gate utilization?
  - What is the estimated number of passengers affected by delays and cancellations?
  - In the ballpark, what are the financial costs, of delays and cancellation?
- What is the future forecasted airline delays?
  - Using historical data and linear progressing, what will be the estimated future delays and cancellations in the next 5, 10, 20 years?
  - Combining historical data and using scenario analysis of varying growths for each airlines, what is the estimated future delays and cancellations in the next 5, 10,
     years using Monte Carlo Analysis;
  - What is the estimated potential impact of number of customers affected and financial costs of these future delays and cancellations?
- What should the airport consider as options to address potential future scenarios?

As the result of this analysis we are planning to build a model based on one of the variations of Linear Programming to optimize the gate allocation. The required methodology cost function will be developed after thorough data analysis where we are planning to use Linear Regression to extract behavioral parameters of airlines from the historical data.

#### The process

For the development of this project we are using CRoss-Industry Standard Process for Data Mining (CRISP-DM) (see Figure 3) was developed by Daimler Chrysler (then Daimler-Benz), SPSS (then ISL) and NCR in 1999. Compiling the current proposal document is a part of this process (see Table 2), in it's preparation our team has conducted 2 iteration of the CRISP-DM process.

#### Potential Risks

Use of assumptions including past performance to predict future performance. Simplifying cost estimates such as aggregate load factors and blended costs. Reliance of public data without management inputs. Data did not exclude normalization of data for abnormal, one-time events.

## Cost/benefit analysis

Understanding gate performance of airlines enable airport management to proper plan airline gate assignments optimally that reduces lost revenues, maximizes the use of gate facilities and defers/avoids expansion of capex due to inefficient use.

If airlines perform as what is agreed to with an airport, then the gates are managed optimally. However, there are instances when a number of flights, get delayed or even cancelled. There are various reasons for these delays. It could be the weather-related, security reason, or natural calamity or sometimes it is the airline (i.e. bad plan, route did not have enough passenger, labour strike, non-payment of leases (these planes are mostly leased financed).

When there is a delay or cancellation, the impact is two-fold from the point of view of the airport. First, customer (passenger) experience suffers. The second is that the airport would incur both actual cost (the personnel it hired, capex spent if applicable, deicing, electricity, use of systems, etc.), revenue lost (landing fees, terminal fees), and opportunity cost (airport could have given that particular gate to another airline).

From the point of the airlines, it is similar but the customer experience is more palpable than the airports. Typically, passengers view delays as the fault of airlines not airports (unlike delays in baggage retrieval is faulted to the airport when it is actually more of the airline). The airline also suffers financially such as cost of transporting the baggage and hotel accommodation, salaries for pilot and crew members, jet fuel costs, etc.

### Conclusion

Airports perform complex operations that involve close coordination between airlines, airport authority and passengers. Optimal gate assignment is an imperative task for airports ground operations to prevent potential delays, maintain operations with high efficiency and maximize profits. This project objective is to determine how passengers are affected by delays and cancellations. Using one of the variations of Linear Programming models based on data analysis, the goal is to maximize the number of passengers served through gates. Optimizing gate assignments to benefit all stakeholders can avoid direct cost of delays for Airlines, Airports are able to maximize the use of gate facilities and avoid unnecessary capital builds and passengers are able to have a better-quality airport experience.

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# Tables

 $\label{eq:continuous} \begin{tabular}{ll} Table 1 \\ Variable \ descriptions \\ \end{tabular}$ 

Name	Description
Year	1987-2008
Month	1 - 12
DayofMonth	1 - 31
DayOfWeek	1 (Monday) - 7 (Sunday)
DepTime	actual departure time (local, hhmm)
CRSDepTime	scheduled departure time (local, hhmm)
ArrTime	actual arrival time (local, hhmm)
CRSArrTime	scheduled arrival time (local, hhmm)
UniqueCarrier	unique carrier code
FlightNum	flight number
TailNum	plane tail number
${\bf Actual Elapsed Time}$	in minutes
${\it CRSE lapsed Time}$	in minutes
AirTime	in minutes
ArrDelay	arrival delay, in minutes
DepDelay	departure delay, in minutes
Origin	origin IATA airport code
Dest	destination IATA airport code
Distance	in miles
TaxiIn	taxi in time, in minutes
TaxiOut	taxi out time in minutes
Cancelled	was the flight cancelled?

Name	Description
CancellationCode	A = carrier, B = weather, C = NAS, D = security)
Diverted	1 = yes, 0 = no
CarrierDelay	in minutes
WeatherDelay	in minutes
NASDelay	in minutes
SecurityDelay	in minutes
${\bf Late Air craft Delay}$	in minutes

Table 2

Delay trend IN/OUT in minute for the top (JetBlue Airways) carrier at JFK.

JetBlue Airways	Delay IN	Delay OUT
Jan-08	12,802	35,088
Feb-08	59,040	58,630
Mar-08	55,435	62,378
Apr-08	25,965	46,178
May-08	16,423	15,769
Jun-08	96,536	57,520
Jul-08	115,981	96,075
Aug-08	115,300	109,549
Sep-08	3,036	22,868
Oct-08	-25,549	6,489
Nov-08	-13,828	26,594
Dec-08	66,061	87,718
Total	527,202	624,856

 $\begin{tabular}{ll} Table 3 \\ \hline Total \ Cancellations \\ \hline \end{tabular}$ 

Month	American Airlines	JetBlue Airways	Delta Airlines	Comair	Total
Jan-08	30	55	16	63	164
Feb-08	42	113	27	130	312
Mar-08	24	72	22	83	201
Apr-08	36	42	6	36	120
May-08	19	10	7	22	58
Jun-08	21	141	14	84	260
Jul-08	41	188	29	193	451
Aug-08	20	197	25	203	445
Sep-08	15	63	19	49	146
Oct-08	7	18	8	15	48
Nov-08	2	37	6	33	78
Dec-08	18	127	30	121	296
					<u>_</u>
Total	275	1,063	209	1,032	2,579

Table 4

Proposed Plan Phases

Phase	Task Name	Schedule	Resources	Status
Business understanding		3 days	All analysts	Completed
Data understanding		2 days	BA	In progress
	Data collection			Completed
	Data description			Completed
	Data quality check			In progress
	Exploratory analysis			In progress
Data preparation		2 days	$BA\backslash Modelers$	In progress
	Selection			2 iterations
	Cleaning			2 iterations
	Construction			2 iterations
	Integration			2 iterations
	Formatting			2 iterations
Modeling		5 days	Modelers	Not started
	Generate test design			
	Build model			
	Interpreter model results			
	Model evaluation			
	Refine and repeat		All analysts	
Evaluation		4 days	Modelers	Not started
	Assessment of results			
	Review of process			
	Determine actions			
	Refine and repeat		All analysts	

Phase	Task Name	Schedule	Resources	Status
Deployment		1 day	$\operatorname{Modelers}\backslash\operatorname{PM}$	Not started
	Deployment plan	Modelers		
	Maintenance plan	PM		
	Final report		$Modelers \backslash PM$	
	Review project		PM	

Table 5

Number of flights per carrier at New York John F. Kennedy Airport (JFK).

Month	В6	DL	ОН	AA	СО	MQ	UA	NW	US	YV	EV
Jan-08	4,540	1,275	1,518	1,068	103	667	445	154	172	81	
Feb-08	4,375	1,238	1,399	996	100	625	430	153	164	75	
Mar-08	4,928	1,380	1,372	1,035	109	645	447	176	186	78	
Apr-08	4,690	1,341	1,230	1,019	108	617	464	175	180	77	
May-08	4,425	1,422	1,117	1,043	107	639	463	182	186	79	
Jun-08	4,444	1,434	1,476	981	107	616	475	175	180	5	46
Jul-08	4,764	1,494	2,508	1,048	111	666	472	178	186		76
Aug-08	4,811	1,451	2,495	1,051	108	640	448	170	175	4	70
Sep-08	3,886	1,338	1,446	936	29	616	383	150	147	33	2
Oct-08	3,936	1,366	1,437	961		605	395	155	147	35	
Nov-08	3,935	1,306	1,450	872		687	374	142	171	85	
Dec-08	4,162	1,348	1,478	936		713	387	121	182	89	
Total	52,896	16,393	18,926	11,946	882	7,736	5,183	1,931	2,076	641	194
	45%	14%	16%	10%	1%	7%	4%	2%	2%	1%	0%

Note. JetBlue Airways (B6), Delta Air Lines Inc. (DL), Comair Inc. (OH), American Airlines Inc. (AA), Continental Air Lines Inc. (CO), American Eagle Airlines Inc. (MQ), United Air Lines Inc. (UA), Northwest Airlines Inc. (NW), US Airways Inc. (US), Mesa Airlines Inc. (YV), Atlantic Southeast Airlines (EV)

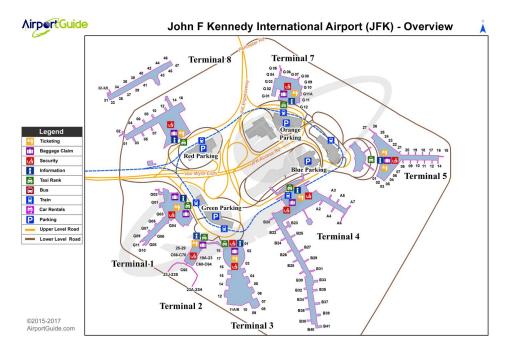
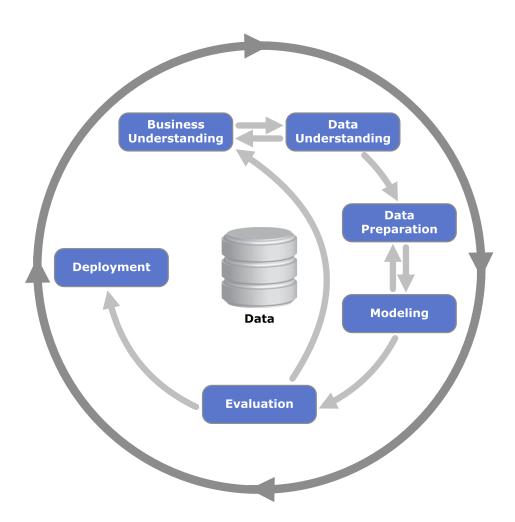


Figure 1. John F. Kennedy Airport in New York (JFK).



Figure 2. San Francisco Airport (SFO).



 $Figure~\it 3$ . CRISP-DM Process Diagramm.