

# AIRLINES DATA ANALYTICS IN AVAITION INDUSTRY



# NALAIYA THIRAN PROJECT BASED LEARNING

On

# PROFESSIONAL READINESS FOR INNOVATION, EMPLOYABILITY AND ENTREPRENEURSHIP

### A PROJECT REPORT

DHIFANI STENIKSHA D	19104048
BARATH S	19104033
BHUVANESH S	19104036
GOKUL RAJ	19104059

# **BACHELOR OF TECHNOLOGY**

IN

## COMPUTER SCIENCE AND ENGINEERING

### HINDUSTHAN COLLEGE OF ENGINEERING AND TECHOLOGY

Approved by AICTE, New Delhi, Accredited with 'A' Grade by NAAC

(An Autonomous Institution, Affiliated to Anna University, Chennai)

**COIMBATORE - 641 032** 

November 2022

### TABLE OF CONTENTS

CHAPTER NO	TITLE	PAGE NO
	ABSTRACT	I
1	INTRODUCTION	1
2	OBJECTIVE	3
3	IDEATION PHASE	5
	3.1 Literature Survey 3.2 Empathy Map	
	3.3 Ideation	
	3.4 Problem Statement	
4	PROJECT DESIGN PHASE 1	
	4.1 Proposed Solution	
	4.2 Problem Solution Fit	
	4.3 Solution Architecture	
5	PROJECT DESIGN PHASE 2	
	5.1 Customer Journey Map	
	5.2 Solution Requirements	
	5.3 Data Flow Diagrams	
	5.4 Technology Stack	
6	PROJECT PLANNING PHASE	
v	6.1 Prepare Milestone and Activity List	
	6.2 Sprint Delivery Plan	
7	PROJECT DEVELOPMENT PHASE	
	7.1 Project Development - Delivery of Sprint - 1	
	7.2 Project Development - Delivery of Sprint - 2	
	7.3 Project Development - Delivery of Sprint - 3	
	7.4 Project Development - Delivery of Sprint - 4	
8	CONCLUSION	33
9	REFERENCES	49

### **ABSTRACT**

In the contemporary world, Data analysis is a challenge in the era of varied intersdisciplines though there is a specialization in the respective disciplines. In other words, effective data analytics helps in analyzing the data of any business system. But it is the big data which helps and axial rates the process of analysis of data paving way for a success of any business intelligence system. With the expansion of the industry, the data of the industry also expands. Then, it is increasingly difficult to handle huge amount of data that gets generated no matter what's the business is like, range of fields from social media to finance, flight data, environment and health.

An Airport has huge amount of data related to number of flights, data and time of arrival and dispatch, flight routes, No. of airports operating in each country, list of active airlines in each country. The problem they faced till now it's, they have ability to analyze limited data from databases.

How can it be gathered, stored, processed and analyzed it to turn the raw data information to support decision making. In this paper Big Data is depicted in a form of case study for Airline data.

### 1.INTRODUCTION

### 1.10verall description

Researchers working in the structured data face many challenges in analyzing the data. For in\_stance the data created through social media, in blogs, in Facebook posts or Snap chat. These types of data have different structures and formats and are more difficult to store in a traditional business data base. The data in big data comes in all shapes and formats including structured. Working with big data means handling a variety of data formats and structures. Big data can be a data created from sensors which track the movement of objects or changes in the environment such as temperature fluctuations or astronomy data. In the world of the internet of things, where devices are connected and these wearables create huge volume of data. Thus big data approaches are used to manage and analyze this kind of data. Big Data include data from a whole range of fields such as flight data, population data, financial and health data such data brings as to another V, value which has been proposed by a number of researcher i.e., Veracity.

Most of the time social media is analyzed by advertisers and used to promote produces and events but big data has many other uses. It can also been used to assess risk in the insurance industry and to track reaction to products in real time. Big Data is also used to monitor things as diverse as wave movements, flight data, traffic data, financial transactions, health and crime. The challenge of Big Data is how to use it to create something that is value to the user. How to gather it, store it, process it and analyze it to turn the raw data information to support decision making.

An Airport has huge amount of data related to number of flights, data and time of arrival and dispatch, flight routes, No. of airports operating in each country, list of active airlines in each country. The problem they faced till now it's, they have ability to analyze limited data from databases. The Proposed model intension is to develop a model for the airline data to provide platform for new analytics based on the following queries.

### 1.1 Problem Statement

- Big amount of data generated on hourly basis.
- A single twin engine aircraft with an average 12 hour flight time can produce up to 844 TB of data
- There are many active users of flights
- Many flights are scheduled everyday
- User varies from common man to celebrities

### 1.2 Purpose

The main purpose of the project to explore detailed analysis on airline data sets such as listing airports operating in the India, list of airlines having zero stops, list of airlines operating with code share which country has highest airports and list of active airlines in united states. The main objective of project is the processing the big data sets using map reduce component of hadoop ecosystem in distributed environment.

### 1.3 Motivation and scope

# **Product Perspective**

The main purpose of the project to explore detailed analysis on airline data sets such as listing airports operating in the India, list of airlines having zero stops, list of airlines operating with code share which country has highest airports and list of active airlines in united states. The main objective of project is the processing the big data sets using map reduce component of Hadoop ecosystem in distributed environment.

### **Product Features**

Airline data analysis can provide a solution for businesses to collect and optimize large datasets, improve performance, improve their competitive advantage, and make faster and better decisions.

- By using airline data analysis, we can save time of users.
- The data could even be structured, semi-structured or unstructured.
- Cost savings
- Implementing new strategies
- Fraud can be detected the moment it happens

### 1.4 Assumptions and Dependencies

- Constraints are limitations which are outside the control of the project. The Project must be managed within these constraints.
- Assumptions are made about events, or facts outside the control of project. External dependencies are activities which need to be completed before an internal activity can proceed.
- Constraints, assumptions and dependencies can create risks that the project may be delayed because access is not provided to the site (assumption).
- Assumption will be that the complexity may arise due to large unstructured data set.

### 1.5 Constraints

- Hardware limitation and timing constraints.
- High feature may not correspond to semantic similarity.
- System Environment Windows subsystem for Linux with Ubuntu operating system will be required to run the application

#### 2. OBJECTIVE

- Safety
- Air navigation Capacity
- Efficiency
- Security
- Facilitation

The security seals ensure the traceability of the food at all times during transport and avoid unauthorized handling.

### 3. IDEATION PHASE

## 3.1 Literature Survey

### 1.Literature focusing on crew recovery

analytics Author: M. Selim Aktürk, Alper Atamtürk, Sinan Gürel

https://www.sciencedirect.com/science/article/pii/S0305054820302549#bb0020

The crew recovery problem (CRP) can be formulated as follows: given a flight schedule and a set of disruptions, re-assign to each (recovered) flight the necessary cabin and flight crew such that the disruption costs are minimized. For crew recovery, these disruption costs can include direct crew costs (e.g., remuneration or overtime compensation) and cost for deadheading crew. For studies that include flight cancellation as a recovery action, cancellation costs can be included in case a flight cannot be staffed. Alternatively, some authors opt to use minimizing the number of crew schedule changes as a proxy to the minimization of the crew recovery costs. The CRP is typically the second problem that is solved in the sequential solution approach. It is considered harder than the ARP since all regulations and restrictions dictated by government regulations, union agreements and airline-specific policies have to be taken into account.

# 2.Literature focusing on passenger recovery

Author: Bruno Aguiar, Jose Torres, António J M Castro

https://www.sciencedirect.com/science/article/pii/S0305054820302549#bb0010

Arguably, passenger recovery is the most relevant problem for airline disruption management since high passenger delay cost and continuous flight disruptions will lead to a potential loss of goodwill and long-term reputation damage. Passenger recovery can be formulated as follows: given a recovered flight and crew schedule and a set of disrupted passenger itineraries, re-assign to each disrupted itinerary the (recovered) flights necessary (given seat availability) to accommodate passengers from their current position to their

destination while minimizing cost. These passenger recovery costs can include both hard and soft costs. Hard costs are directly incurred when a passenger cannot complete its scheduled itinerary (e.g., compensation for delay and cancellation as stipulated by government regulations). Soft costs are the potential losses of future revenue as a result of passenger inconvenience, possibly causing the passenger to switch to a different airline in the future. These costs are approximations made by the airline and can differ per passenger class or frequent flyer status. Alternatively, these passenger disruption costs are minimized by minimizing the total number of passenger delay minutes.

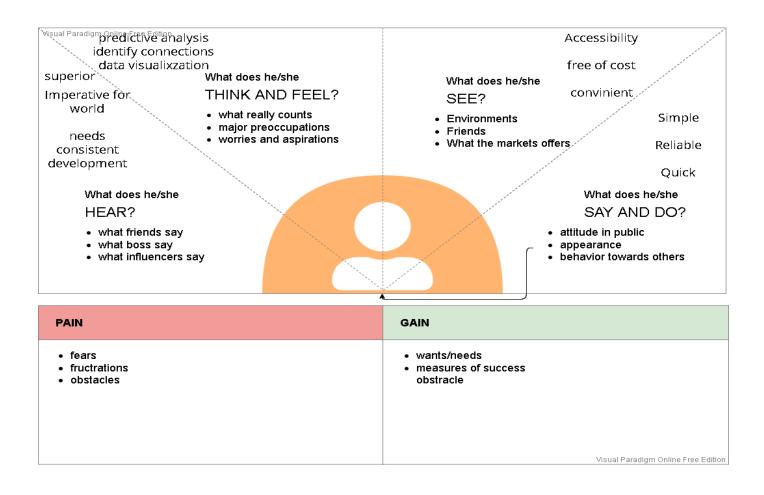
### 3.Literature focusing on integrated recovery

Industry Author: Khaled F. Abdelghany, Ahmed F. Abdelghany, Goutham Ekollu

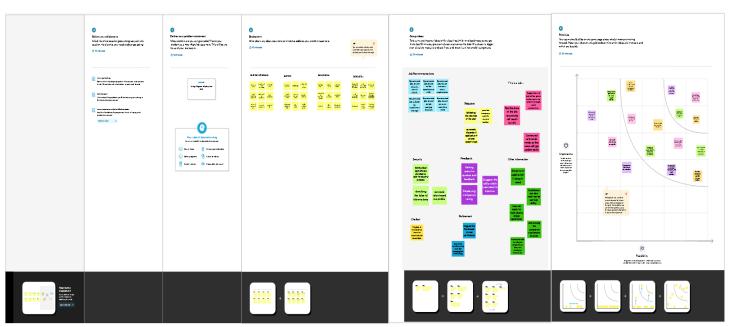
https://www.sciencedirect.com/science/article/pii/S0305054820302549#bb0005

Both from a mathematical and computational perspective, the integration of all recovery stages (aircraft, crew, and passengers) is a difficult task. The purpose of this integration is to minimize the total disruption cost. This is achieved by weighing the disruption cost related to aircraft, crew, and passengers simultaneously to find the recovery solution that overall results in the lowest cost for the airline. To the best of the authors' knowledge, the first proposal of a truly integrated approach was the PhD Thesis of Lettovsky (1997), where the author formulated the 'Airline Integrated Recovery' problem which consists of aircraft routing, crew assignment, and passenger flow. The thesis presents a linear mixed-integer mathematical problem that captures the availability of the aforementioned resources. A decomposition scheme is presented where the 'Schedule Recovery Model' master problem controls the three sub-problems known as the 'Aircraft recovery model', 'Crew recovery model', and 'Passenger flow model'. The solution is derived by applying Benders' decomposition. A limitation is that the model only considers the cockpit crew and not cabin crew

# 3.2 Empathy Map



# 3.3 Brainstroming

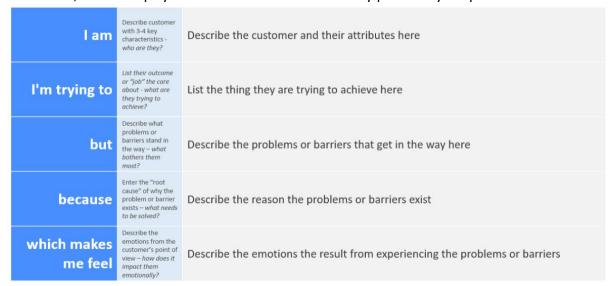


3.4 Problem Statement

#### **Customer Problem Statement Template:**

Create a problem statement to understand your customer's point of view. The Customer Problem Statement template helps you focus on what matters to create experiences people will love.

A well-articulated customer problem statement allows you and your team to find the ideal solution for the challenges your customers face. Throughout the process, you'll also be able to empathize with your customers, which helps you better understand how they perceive your product or service.



Reference: <a href="https://miro.com/templates/customer-problem-statement/">https://miro.com/templates/customer-problem-statement/</a> Example:



Problem Statement (PS)	I am (Customer)	I'm trying to	But	Because	Which makes me feel
PS-1	Facing flight delay	Evolve numerous techniques of improving the airlines transportation system	This has brought drastic change in airlines operations	Flights delay ocassinally cause inconvenience to the modern passengers	It hurts airports ,airlines,and affects a companys marketing strategies as companies rely on customer loyalty to support their

					frequent flying programs.
PS-2	Facing HAP emission from idling aircraft on ambient conditions	Improve our quantitative understanding of the largest aviation related HAPs emission source jet engines operating at low power	Potentially toxic emissions is hazardous air pollutant emissions and most important source of airport related HAPs compound at most commercia airports in idling jet engines	This will allow airport operators to utilize the latest scientific findings to construct HAPs emission estimates tailored to their specific airport	These estimates will be more defensible and better able to withstand litigation since they will be based on latest scientific findings regarding jet engine HAPs emissions and variabes that affect them.

# **4.PROJECT DESIGN PHASE 1**

# **4.1 Proposed Solution**

**Proposed Solution Template:**Project team shall fill the following information in proposed solution template.

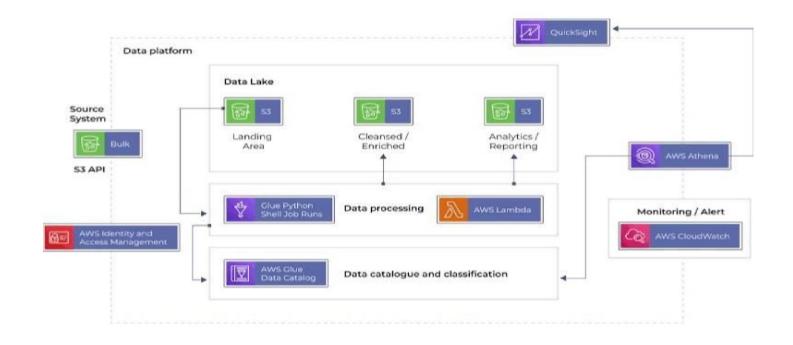
S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	The airline industry has been keeping a tab on this information since long but it needs big data to help them analyse it and make it useful for the customer
2.	Idea / Solution description	The purpose of data analytics in aviation is to examine the vast amount of data generated daily and provide useful information to airlines, airports and other aviation stakeholders so that they can improve their operational planning and execution, as well as any related products and services
3.	Novelty / Uniqueness	Aforable. Easy to use. Services: Advance analysis, Easy to use and maintain, Actionnable report.
4.	Social Impact / Customer Satisfaction	The results of data analysis show that, in overall, full service airline customers are more satisfied than that of the low cost airline customers. Further, regression analysis on low cost airline data shows that the promptness and accuracy of service, employee attitudes, and price significantly influence customer satisfaction. While in full service airline

		physical evidence, the attitude of employees, and the price are significant predictors of customer satisfaction. This study underlines that the service quality especially the service employees' attitudes and price are factors that should be given more attention for developing customer satisfaction in both types of airlines, although their competitive strategy and target market are different.
5.	Business Model (Revenue Model)	There are two main business models in the airline industry: traditional Full-Service Carriers (FSCs) and Low-Cost Carriers (LCCs). The LCC business model was first pioneered by US-based Southwest Airlines. In a nutshell, low-cost airlines minimize operations costs to offer the cheapest tickets possible.
6.	Scalability of the Solution	This study illustrates how airlines successfully adopt big data technology. The paper also explores the opportunities and challenges of big data in the airline industry.

# **4.2 Problem Solution Fit**

Stakeholders	Arrival/ parking	Check- in	Checking bag	Security	Airport activities	Boarding aircraft	Depar- ture	Arrival	Customs	Baggage claim	Departure parking
Airlines		Bag	tracking						Bag trackir	ng .	
	71	MCO Inc	door naviga	ition app:	\$				MCO inc	loor navig	ation apps
Concess- ionaires					Instore beacons						
Other commer- cial tenants	MCO										MCO app
Off-airport transport- ation	Geo- fencing										Geo- fencing
Vendors to airport					As	set trackin	g				
Security, customs & aviation authorities			Qu	eue analy	rzer				ueue anal		
Airport		W	I-FI passen	ger track	ing				Wi-Fi to	acking	
staff					Building	managem	nent sys	tem		- 12	

# **4.3 Solution Architecture**



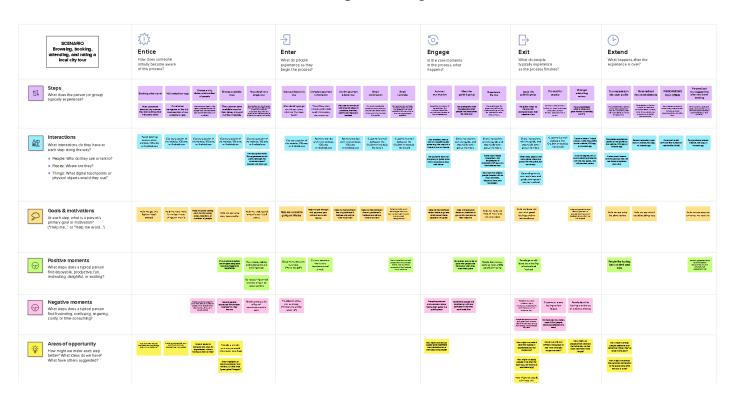
#### **5.PROJECT DESIGN PHASE 2**

### **5.1 Customer Journey Map**

Customer journey mapping is "the process of tracking and describing all the experiences customers have as they encounter a service or set of services, taking into account not only what happens to them, but also their responses to these experiences" (Dent, 2015). As stated in Eva Manrique's blog: (2016) "when airlines adopt the customer journey mapping strategy, it helps them to clarify what each individual customer expects at each touchpoint and which fitting service or product the airline could provide in order to fulfil these expectations."

The customer journey map is divided into various phases as the customer has various different options in regards to approaching the airline, these phases include the three communication arena's: physical, digital and social.

- Phase 1: attract: the customer recognizes the airline for (potentially) the first time, via tools such as social media but also through discount offers posted on the website or social media. So, this stage takes place in both the digital as the social arena.
- Phase 2: decide: the customer decides whether or not to purchase the flight ticket on digitally or via the phone with a call-centre employee / through a physical store that offers the airline's tickets. Furthermore, the customer is getting familiar with the airline by purchasing the ticket, which leads to an increase of flows to the social media accounts of the airline.
- Phase 3: use: happens when the customer will actually experience the flight that was purchased with the airline. As the customer receives the boarding pass through email communication, the flight information as well as maps will be provided to the customer in order to prepare for the trip, which can be saved in a digital wallet. Other additional resources, such as gate information and travel guides are provided to the customer either through email, social media or actual mail.
- Phase 4: support, as it is crucial for an airline to maintain contact with the customer before, during as well as after the flight. It is inevitable that things could go wrong during the flight which is why it is important that airlines show their support to the customer through either the digital and/or social arena.
- Phase 5: retain: airlines need to be able to retain customers, especially frequent flyers. Airlines can implement this by letting customers manage their own bookings on the website as well as offering discounts to returning customers as this makes the customer feel valued, which broadens the chances of customers repurchasing with the airline



# **5.2 Solution Requirements**

# **Functional Requirements:**

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	Can register through gmail/phone number
FR-2	User Confirmation	Receives Confirmation text via Email /OTP
FR-3	Visualization of data	Through IBM cognos analytics to know about delay of flights
FR-4	Generation of report	Users can know the timings and delay durations

# **Non-functional Requirements:**

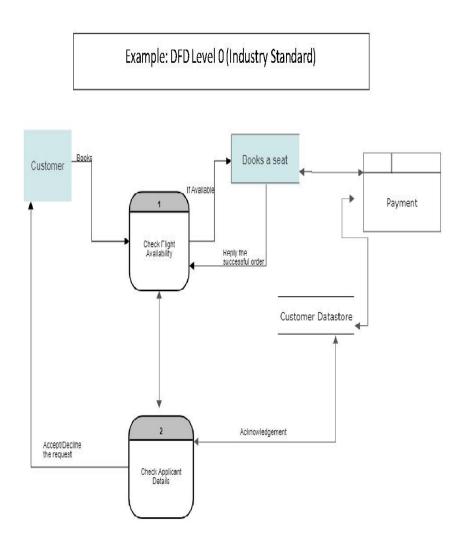
Following are the non-functional requirements of the proposed solution.

NFR-	Availability	The application must be available to	
5		access at anytime, anydays	
NFR-	Scalability	Better scalability that large number of	
6		users could able to access at a time	

FR	Non-Functional	Description	
No.	Requirement		
NFR-1	Usability	Users can access the application easily. Any functions can be performed with simple steps.	
NFR-2	Security	Proper digital privacy system should be implemented to protect the details of user. Efficient login system should be made.	
NFR-3	Reliability	When the system /server processing lowers,it should able to restain the saved datas of user	
NFR-4	Performance	The system should have a efficient speed for browsing details	

# **5.3 Data Flow Diagrams**

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.



User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Web user)	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	I can access my account / dashboard	High	Sprint-1
		USN-2	As a user, I will receive confirmation email once I have registered for the application	I can receive confirmation email & click confirm	High	Sprint-1
		USN-3	As a user, I can register for the application through Gmail.		Medium	Sprint-1
	Login	USN-4	As a user, I can log into the application by entering email & password.	I can get to access my web portal	High	Sprint-1
	Dashboard	USN-5	As a user, I can get to know what my dashboard consists of.	I can my details of my registration.	Low	Sprint-2
Customer Care Executive	Organization	USN-6	The organization which owns this airplane analysis system will enable the option to customers to reach out the organization if  they have any problem with the organization's system of customer interaction or  airplane issues- delay, landing in a different location	The customer care workers will help out the customers in trouble.	High	Sprint-1
Administrator	Administration	USN-7	The organization takes in-charge of the administrative policies of different departments like:  • registration • flight booking • delay visualization • generation of delay report	As an administrator, confirmation of user while registration is done.	High	Sprint-1

# **5.4 Technology Stack**

### **Technical Architecture:**

The Deliverable shall include the architectural diagram as below and the information as per the table 1 & table 2

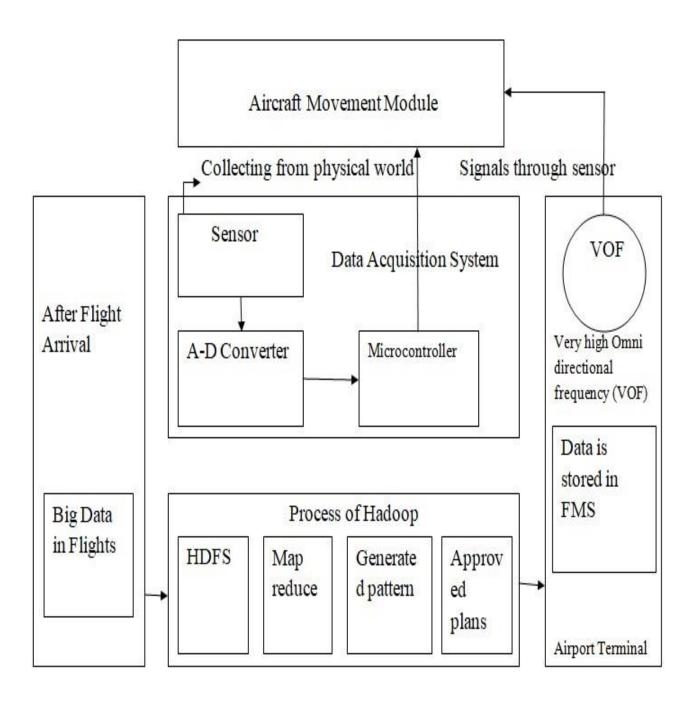


Table-1: Components & Technologies:

S.No	Characteristics	Description	Technology
1.	Open-Source Frameworks	List the open-source frameworks used	Technology of open- source framework
2.	Security Implementations	List all the security/access controls implemented, use of firewalls.	Example: SHA-256, Encryption, IAM Controls, OWASP

S.No	Components	Description	Technology	
1.	User Interface	How user interacts with application. Example: Mobile App	HTML, CSS, Java Script, Excel	
2.	Application Logic-1	Logic for a process in the application	IBM Watson STT service, Python	
3.	Application Logic-2	Logic for a process in the application	IBM Watson Assistant	
4.	Database	Data Type, Configurations	MySQL, NSQL	
5.	Cloud Database	Database service on cloud	IBM DB2, IBM Cloudant	
6.	File Storage	File Storage requirements	IBM Blocks Storage or other storage service or Local File system	
7.	External API-1	Purpose of External API used in the application	IBM Weather API	
8.	External API-1	Purpose of External API used in the application	Aadhar API	

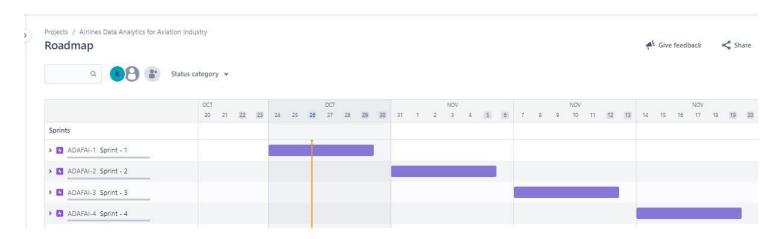
9.	Infrastructure (Server/Cloud)	Application Deployment on Local System/Cloud Local Server	Local, Cloud Foundry
		Configuration: Cloud	
		Server Configuration	

# **Table-2: Application Characteristics:**

3.	Scalable Architecture	Justify the scalability of architecture	Cognos Used
4.	Availability	Justify the availability of application (e.g: use of load balancers, distributed servers)	AWS Used
5.	Performance	Design consideration for the performance of the application (number of requests per second, use of Cache, use of CDN's)	Dashboard, Reports, Stories

## **6.PROJECT PLANNING PHASE**

# **6.1 Prepare Milestone and Activity List**



# **6.2 Sprint Delivery Plan**

# **Product Backlog, Sprint Schedule, and Estimation (4 Marks)**

Use the below template to create product backlog and sprint schedule

Sprint	int Functional User User Story / Task Story Priority Tear					Team Members
ор	Requirement	Story	Cool Glory, ruen	Points		
	(Epic)	Number				
Sprint1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	4	High	DHIFANI STENIKSHA D
Sprint1	Login	USN-2	As a user, I adapt to logging into the system with credentia	2	Low	BARATH
Sprint1	Designation of Region	USN-3	As a user, I can collect the dataset and select the region of interest to be monitored and analysed	2	Low	GOKUL RAJ
Sprint2	Exploration Of The Data	USN-4	As a developer,I will explore the given dataset through cognos.	3	Medium	BHUVANESH
Sprint2	Visualization Of The Dataset	USN-5	As a developer,I will visualize the given dataset into a dashboard using cognos	2	Low	DHIFANI STENIKSHA D
Sprint3	Customization Of The Dashboard	USN-6	As a user,I can customize the visualized dashboard.	2	Low	BARATH
Sprint3	Ease of Access	USN-7	As a user,I can easily access and	2	Low	GOKULRAJ
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
			manipulate the dashboard.			
Sprint4	Report Generation	USN-8	As a user,I can view the detailed report of my visualization.	4	High	DHIFANI STENIKSHA D

Sprint4	Establishment of	USN-9	As a developer,I	3	Medium	BHUVANESH
	the		established the			
	Dashboard		dashboard intoa			
			website and submit			
			the website.			

### Project Tracker, Velocity & Burndown Chart: (4 Marks)

Sprint	Total Story	Duration	Sprint Start	Sprint End Date	Story Points Completed (as on	Sprint Release Date
	Points		Date	(Planned)	Planned End Date)	(Actual)
Sprint1	20	6 Days	24 Oct 2022	29 Oct 2022	12	29 Oct 2022
Sprint2	20	6 Days	31 Oct 2022	05 Nov 2022	12	05 Nov 2022
Sprint3	20	6 Days	07 Nov 2022	12 Nov 2022	12	12 Nov 2022
Sprint4	20	6 Days	14 Nov 2022	19 Nov 2022	12	19 Nov 2022

#### **Velocity:**

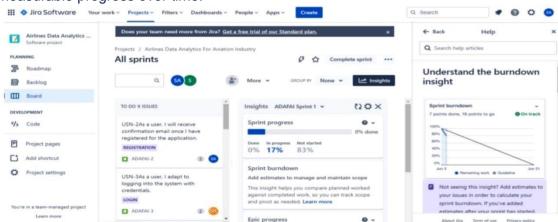
Imagine we have a 10-day sprint duration, and the velocity of the team is 20 (points per sprint). Let's calculate the team's average velocity (AV) per iteration unit (story points per day)

$$AV = \frac{sprint\ duration}{velocity} = \frac{20}{10} = 2$$

Average velocity=Sprint duration / velocity=12/6=2

### **Burndown Chart:**

A burn down chart is a graphical representation of work left to do versus time. It is often used in agile\_software development methodologies such as Scrum. However, burn down charts can be applied to any project containing measurable progress over time.



### 7. PROJECT DEVELOPMENT PHASE

# 7.1 Project Development - Delivery of Sprint - 1

### Flight count from Top 5 Airlines at Top 5 Airports

```
WITH top_5_airports AS (
       SELECT ORIGIN, COUNT(ORIGIN) AS count
       FROM
             airline-delay-canc.airlines_data.delay_canc_data
       GROUP BY
              1
       HAVING
             count > 100000
       ORDER BY
             2 DESC
       LIMIT 5
top_5_airlines AS (
       SELECT
              OP_CARRIER,
             COUNT(OP_CARRIER) AS count
       FROM
              airline-delay-canc.airlines_data.delay_canc_data main,
              top_5_airports top5
       WHERE
             top5.ORIGIN = main.ORIGIN
       GROUP BY
              1
       ORDER BY
              2 DESC
       LIMIT 5
airportwise_carrier_cnt AS (
       SELECT
              main.ORIGIN AS Airport,
             main.OP_CARRIER AS Carrier,
             COUNT(*) AS count
       FROM
              airline-delay-canc.airlines_data.delay_canc_data main,
             top 5 airports top5 ap,
             top_5_airlines top_al
       WHERE
             top5 ap.ORIGIN = main.ORIGIN
             AND top_al.OP_CARRIER = main.OP_CARRIER
       GROUP BY
              1,
```

```
2
resut_cte AS (
       SELECT
              Airport,
              Carrier,
              count,
              RANK() OVER (PARTITION BY Airport ORDER BY count) AS rank
       FROM
              airportwise_carrier_cnt
SELECT
       Airport,
       Carrier,
       count
FROM
       resut_cte
WHERE
       rank < 6
```

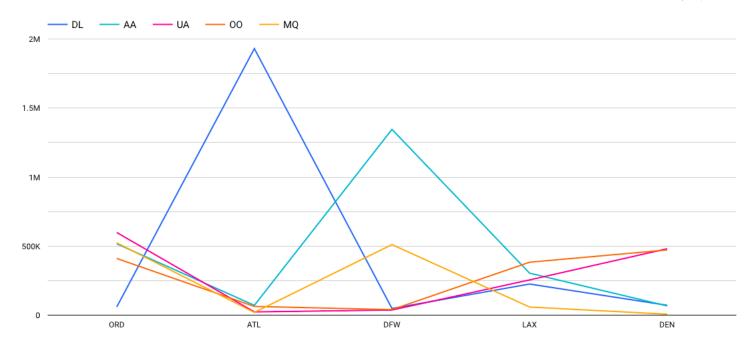
#### **Results**

*Top 5 Airports with maximum flight count:* 

- 1. **ORD** (O'Hare International Airport)
- 2. ATL (Hartsfield-Jackson Atlanta International Airport)
- 3. **DFW** (Dallas/Fort Worth International Airport)
- 4. **LAX** (Los Angeles International Airport)
- 5. **DEN** (Denver International Airport)

*Top 5 Airlines with maximum flight count:* 

- 1. **DL** (Delta Air Lines)
- 2. AA (American Airlines)
- 3. **UA** (United Airlines)
- 4. **OO** (SkyWest Airlines)
- 5. **MQ** (American Eagle Airlines)



• From the above, it is realized that on **Delta Airlines** has the highest flight frequence on the **Atlanta** airport.

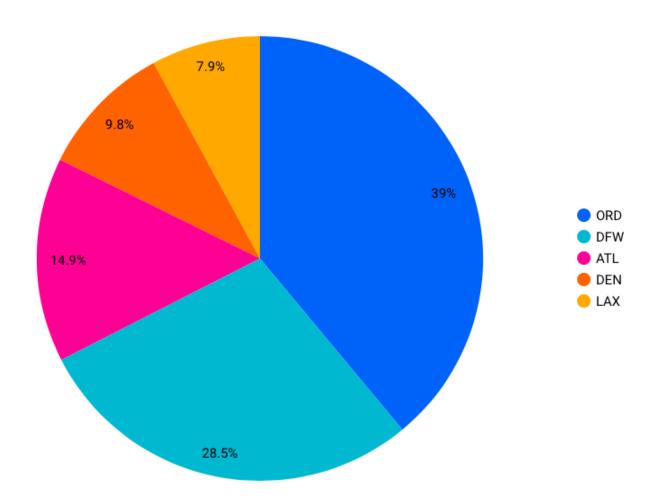
# **Top 5 Airports with Maximum Cancellations (decreasing order)**

```
WITH
top_5_airports AS (
SELECT
 ORIGIN,
 COUNT(ORIGIN) AS count
FROM
  `airline-delay-canc.airlines_data.delay_canc_data`
GROUP BY
 1
ORDER BY
 2 DESC
LIMIT
 5),
top_5_airlines AS (
SELECT
 OP_CARRIER,
 COUNT(OP_CARRIER) AS count
```

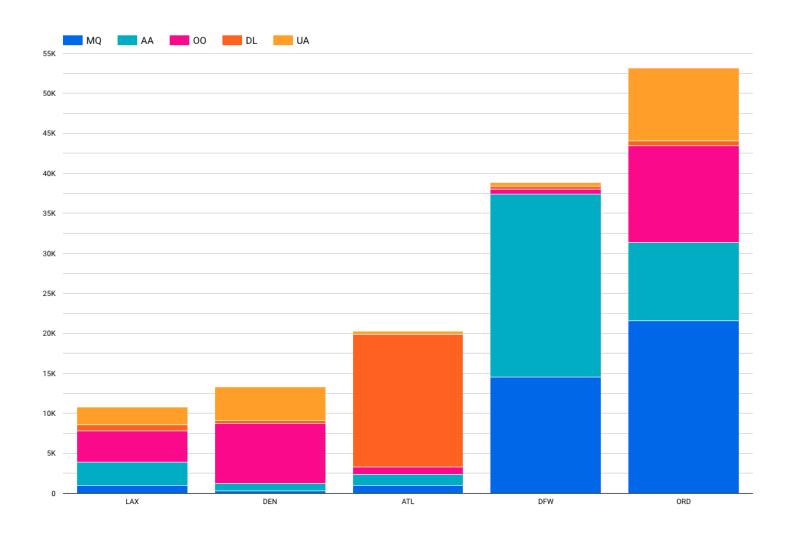
```
FROM
  `airline-delay-canc.airlines data.delay canc data` main,
  top_5_airports top5
 WHERE
  top5.ORIGIN = main.ORIGIN
 GROUP BY
  1
 ORDER BY
  2 DESC
 LIMIT
  5),
 all_flights AS (
 SELECT
  main.ORIGIN AS Airport,
  main.OP_CARRIER AS Carrier,
  COUNT(*) AS all cnt
 FROM
  `airline-delay-canc.airlines_data.delay_canc_data` main,
  top_5_airports top5_ap,
  top_5_airlines top_al
 WHERE
  top5_ap.ORIGIN = main.ORIGIN
  AND top_al.OP_CARRIER = main.OP_CARRIER
 GROUP BY
  1,
  2),
 cancelled_flights AS (
 SELECT
  main.ORIGIN AS Airport,
  main.OP_CARRIER AS Carrier,
  COUNT(*) AS cancelled_cnt
 FROM
  `airline-delay-canc.airlines_data.delay_canc_data` main,
  top_5_airports top5_ap,
  top 5 airlines top al
 WHERE
  top5_ap.ORIGIN = main.ORIGIN
  AND top al.OP CARRIER = main.OP CARRIER
  AND cancelled = 1
 GROUP BY
  1,
  2)
SELECT
 af.Airport,
 af.Carrier,
 af.all_cnt - cf.cancelled_cnt AS all_cnt,
 cf.cancelled cnt
FROM
 all_flights af,
 cancelled flights cf
WHERE
```

#### **Results**

| S No.| Airport Code | Airport Name | Cancellation (in %) | | - | - | - | - | - | 1. | **ORD** | (O'Hare International Airport) | 39 | 2. | **DFW** | (Dallas/Fort Worth International Airport) | 28.5 | 3. | **ATL** | (Hartsfield-Jackson Atlanta International Airport) | 14.9 | 4. | **DEN** | (Denver International Airport) | 9.8 | 5. | **LAX** | (Los Angeles International Airport) | 7.9 |



### Airline-wise Cancellation Bifurcation



# 7.2 Project Development - Delivery of Sprint - 2

# **Top Cancellation Reasons for Top 5 Busiest Airports**

**Query - JS UDF Function** 

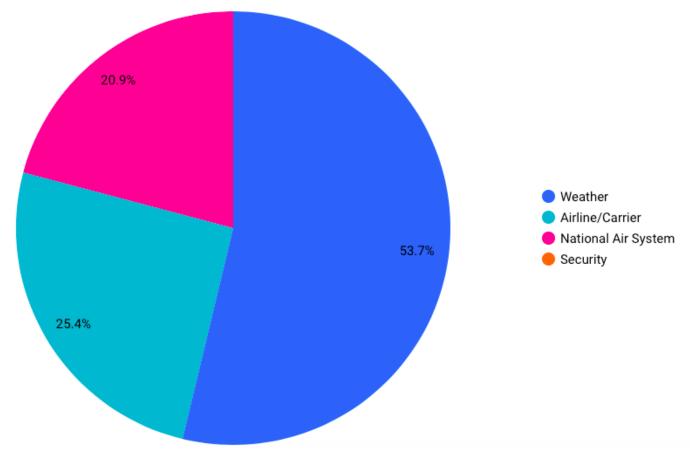
CREATE TEMP FUNCTION cancellation\_reason(code string) RETURNS string LANGUAGE js AS ''''' switch(code) {

```
case "A":
     return "Airline/Carrier";
    break:
    case "B":
     return "Weather";
    break:
    case "C":
     return "National Air System";
    break:
    case "D":
     return "Security";
    break;
    default:
     return "Others";
   break:
 }
""";
WITH
top_5_airports AS (
 SELECT
  ORIGIN,
  COUNT(ORIGIN) AS count
FROM
  `airline-delay-canc.airlines data.delay canc data`
 GROUP BY
  1
HAVING
  count > 100000
 ORDER BY
  2 DESC
LIMIT
  5)
SELECT
top5.ORIGIN,
cancellation_reason(main.CANCELLATION_CODE) AS reason,
 COUNT(main.CANCELLATION_CODE) AS count
FROM
 `airline-delay-canc.airlines_data.delay_canc_data` main,
top_5_airports top5
WHERE
```

```
CANCELLED = 1
AND EXTRACT(year
FROM
FL_DATE) = 2018
AND top5.ORIGIN = main.ORIGIN
GROUP BY
1,
2
ORDER BY
1,
2
```

#### Result

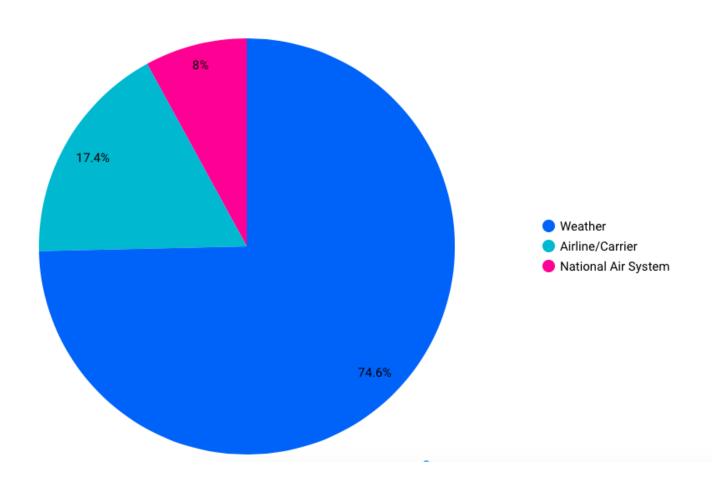
| S No.| Reason | Cancellation (in %) | | - | - | - | - | 1. | **Weather** | 53.7| | 2. | **Airline/Carrier Delays** | 25.4| | 3. | **National Air System** | 20.9| | 4. | **Airport Secutiy** | 0.01 (~ 0)|



# Top Cancellation Reasons at the Most Busiest Airport in practice (Atlanta)

• Atlanta is one of the largest inter-connect point (airport) for domestic and international flights in USA.

|S No.| Reason | Cancellation (in %) | | - | - | - | 1. | **Weather** | 74.6| | 2. | **Airline/Carrier Delays** | 17.4| | 3. | **National Air System** | 8|



# 7.3 Project Development - Delivery of Sprint -3

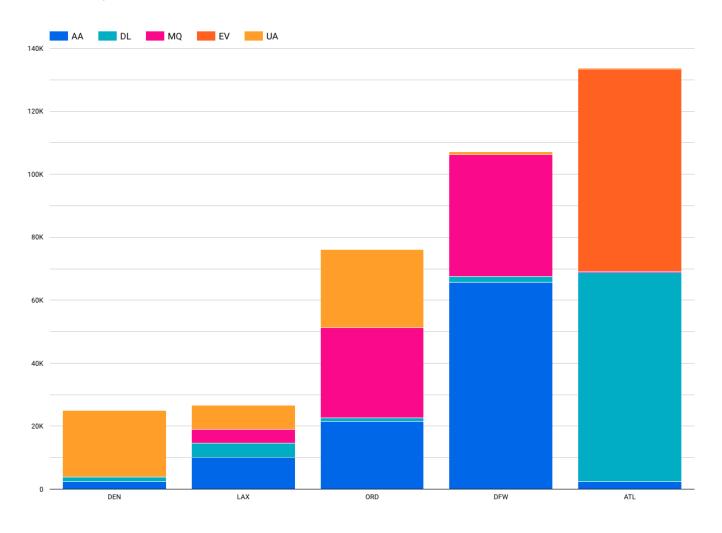
### Overall Delays at Top 5 Airports for top 5 airlines

```
Query
WITH
  top_5_airports AS (
```

```
SELECT
  ORIGIN,
  COUNT(ORIGIN) AS count
  `airline-delay-canc.airlines_data.delay_canc_data`
GROUP BY
  1
ORDER BY
  2 DESC
LIMIT
  5),
top_5_airlines AS (
SELECT
  OP CARRIER,
  COUNT(OP_CARRIER) AS count
  `airline-delay-canc.airlines_data.delay_canc_data` main,
  top_5_airports top5
WHERE
  top5.ORIGIN = main.ORIGIN
GROUP BY
ORDER BY
  2 DESC
LIMIT
  5),
all_flights AS (
SELECT
  main.ORIGIN AS Airport,
  main.OP CARRIER AS Carrier,
  COUNT(*) AS all_cnt
  `airline-delay-canc.airlines_data.delay_canc_data` main,
  top 5 airports top5 ap,
  top_5_airlines top_al
WHERE
  top5 ap.ORIGIN = main.ORIGIN
  AND top_al.OP_CARRIER = main.OP_CARRIER
GROUP BY
  1,
  2),
delayed_flights AS (
SELECT
  main.ORIGIN AS Airport,
  main.OP_CARRIER AS Carrier,
  COUNT(*) AS delayed_cnt
  `airline-delay-canc.airlines_data.delay_canc_data` main,
  top_5_airports top5_ap,
  top_5_airlines top_al
WHERE
  top5 ap.ORIGIN = main.ORIGIN
  AND top al.OP CARRIER = main.OP CARRIER
  AND (CARRIER_DELAY IS NOT NULL
    AND CARRIER DELAY > 0
    OR ARR_DELAY IS NOT NULL
```

```
AND ARR_DELAY > 0)
  GROUP BY
    1,
    2)
SELECT
  af.Airport,
  af.Carrier,
  af.all_cnt all_with_del,
  df.delayed_cnt,
  af.all_cnt - df.delayed_cnt AS all_without_del
FROM
  all_flights af,
  delayed_flights df
WHERE
  af.Airport = df.Airport
  AND af.Carrier = df.Carrier
```

### Overall Delays at Top 5 Airports with top 5 airlines



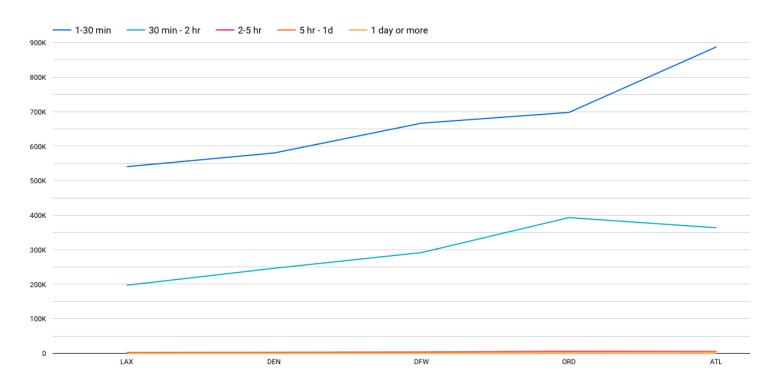
### **Overall Delay Time Frequency with Top 5 Airports**

#### Query

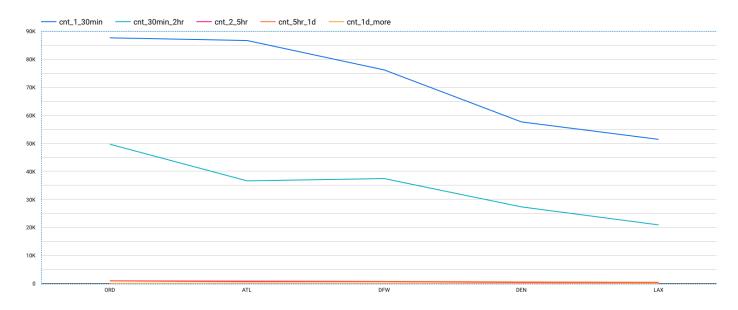
```
CREATE TEMP FUNCTION delay bifurcation(slot cnt ARRAY<STRUCT<slot int64,count int64>>)
   RETURNS STRUCT<cnt_1_30 float64, cnt_30_2 float64, cnt_2_5 float64, cnt_5_24 float64, cnt_24
float64>
  LANGUAGE js AS """
  let response = {"cnt_1_30": 0.0, "cnt_30_2": 0.0, "cnt_2_5": 0.0, "cnt_5_24": 0.0, "cnt_24":
0.0}
  for(let i = 0; i < slot_cnt.length; i++){</pre>
      let slotCntObj = slot_cnt[i];
      let result = slotCntObj.count;
      switch(parseInt(slotCntObj.slot)){
        case 1:
          response["cnt_1_30"] = result;
          break;
        case 2:
          response["cnt_30_2"] = result;
          break;
        case 3:
          response["cnt_2_5"] = result;
          break;
        case 4:
          response["cnt_5_24"] = result;
          break;
        case 5:
          response["cnt_24"] = result;
          break;
        default:
          response["cnt 1 30"] = 0.0;
          response["cnt_30_2"] = 0.0;
          response["cnt_2_5"] = 0.0;
          response["cnt_5_24"] = 0.0;
          response["cnt 24"] = 0.0;
          break;
      }
    }
    return response
WITH top_5_airports as (
      SELECT ORIGIN, count(ORIGIN) as count
      FROM `airline-delay-canc.airlines data.delay canc data`
      Group by 1
      having count > 100000
      order by 2 desc
      limit 5
      ),
    delay_bifurcation as (
      select ORIGIN,
          (case when ARR DELAY > 1440 then 5
             when ARR DELAY > 300 then 4
             when ARR DELAY > 240 then 3
             when ARR DELAY > 30 then 2
        else 1 end) as slot
```

```
from `airline-delay-canc.airlines data.delay canc data`
where ARR DELAY is not null and ARR DELAY > 0
   and EXTRACT(year FROM FL_DATE) = 2018
),
airport timeslots as(
select db.ORIGIN, db.slot, count(db.slot) as count
from delay_bifurcation db,top_5_airports top5
where top5.ORIGIN = db.ORIGIN
group by 1,2),
airport_struct as(
    select origin, struct(slot,count) as slot cnt from airport timeslots
),
udf_result as (select origin, delay_bifurcation(ARRAY_AGG(slot_cnt)) as slot_struct
from airport struct
group by 1
select origin, slot_struct.cnt_1_30 as cnt_1_30min,
    slot_struct.cnt_30_2 as cnt_30min_2hr,
    slot_struct.cnt_2_5 as cnt_2_5hr,
    slot_struct.cnt_5_24 as cnt_5hr_1d,
    slot_struct.cnt_24 as cnt_1d_more
from udf_result
```

### Overall Delay Time Frequency with Top 5 Airports (UDF Function)



Overall Delay Frequency (Year with max delays and cancellations)



### 7.4. Project Development - Delivery of Sprint – 4

### **Delay Percentage for top 5 airports**

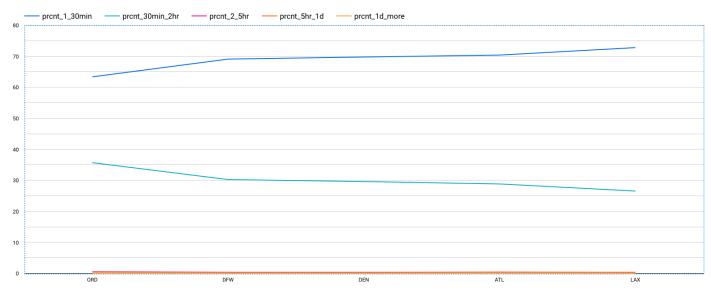
#### Query

```
CREATE TEMP FUNCTION delay_bifurcation(slot_cnt ARRAY<STRUCT<slot int64,count int64>>)
   RETURNS STRUCT<cnt_1_30 float64, cnt_30_2 float64, cnt_2_5 float64, cnt_5_24 float64, cnt_24
float64>
  LANGUAGE js AS """
  let response = {"cnt_1_30": 0.0, "cnt_30_2": 0.0, "cnt_2_5": 0.0, "cnt_5_24": 0.0, "cnt_24":
0.0}
  let total_delayed_flights = 0;
  for(let i = 0 ; i < slot_cnt.length; i++){</pre>
    total_delayed_flights += parseInt(slot_cnt[i].count);
  for(let i = 0; i < slot_cnt.length; i++){</pre>
      let slotCntObj = slot cnt[i];
                    parseFloat(parseInt(slotCntObj.count) / total_delayed_flights *
      let result =
100).toFixed(2);
      switch(parseInt(slotCntObj.slot)){
          response["cnt_1_30"] = result;
          break;
```

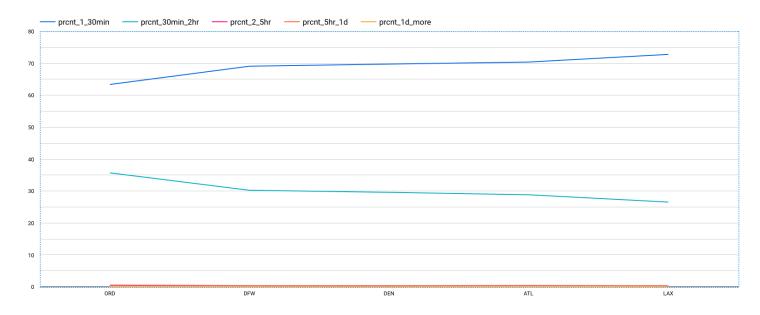
```
case 2:
          response["cnt_30_2"] = result;
          break;
        case 3:
          response["cnt_2_5"] = result;
          break;
        case 4:
          response["cnt_5_24"] = result;
          break;
        case 5:
          response["cnt_24"] = result;
          break;
        default:
          response["cnt_1_30"] = 0.0;
          response["cnt_30_2"] = 0.0;
          response["cnt 25"] = 0.0;
          response["cnt_5_24"] = 0.0;
          response["cnt_24"] = 0.0;
          break;
      }
    }
    return response
WITH top 5 airports as (
      SELECT ORIGIN, count(ORIGIN) as count
      FROM `airline-delay-canc.airlines data.delay canc data`
      Group by 1
      having count > 100000
      order by 2 desc
      limit 5
      ),
    delay_bifurcation as (
      select ORIGIN,
          (case when ARR DELAY > 1440 then 5
             when ARR DELAY > 300 then 4
             when ARR_DELAY > 240 then 3
             when ARR DELAY > 30 then 2
        else 1 end) as slot
  from `airline-delay-canc.airlines_data.delay_canc_data`
  where ARR_DELAY is not null and ARR_DELAY > 0
     and EXTRACT(year FROM FL DATE) = 2018 -- used for filtering
  ),
  airport timeslots as(
  select db.ORIGIN, db.slot, count(db.slot) as count
  from delay_bifurcation db,top_5_airports top5
  where top5.ORIGIN = db.ORIGIN
  group by 1,2),
  airport struct as(
      select origin, struct(slot,count) as slot_cnt from airport_timeslots
  ),
  udf_result as (select origin, delay_bifurcation(ARRAY_AGG(slot_cnt)) as slot_struct
```

```
from airport_struct
group by 1
)
select origin, slot_struct.cnt_1_30 as prcnt_1_30min,
    slot_struct.cnt_30_2 as prcnt_30min_2hr,
    slot_struct.cnt_2_5 as prcnt_2_5hr,
    slot_struct.cnt_5_24 as prcnt_5hr_1d,
    slot_struct.cnt_24 as prcnt_1d_more
from udf_result
```

## Delay Percentage for top 5 airports



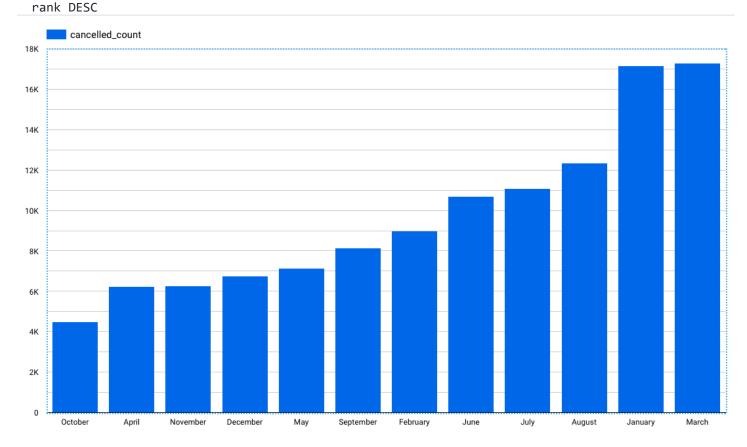
# Delay Percentage for top 5 airports



# Most unreliable month (Cancellations in ascending order)

#### Query

```
WITH
  cancelled_count_cte AS (
  SELECT
    ROW_NUMBER() OVER (ORDER BY cancelled_count) AS RANK
  FROM (
    SELECT
      FORMAT_DATE('%B', FL_DATE) AS month,
      SUM(CANCELLED) AS cancelled count
    FROM
       `airline-delay-canc.airlines_data.delay_canc_data`
    WHERE
      EXTRACT(year
      FROM
        FL_DATE) = 2018
    GROUP BY
      1))
SELECT
  month,
  cancelled_count
FROM
  cancelled_count_cte
ORDER BY
```



### 8.CONCLUSION

It can be used to predict future glitches, prevent them from happening, and make the maintenance procedures more accurate and thorough.

After analyzing the data, a lot of insights have been generated. Most of the delays and cancellations are due to three major reasons:

- Weather
- Airline/Carrier Issues
- National Air System

### 9.REFERENCES

- https://www.iata.org/en/publications/store/world-air-transport-statistics/
- https://www.google.com/search?lei=cl9oY5byKqSvmgesiqwDQ&q=data%20analytics%20in%20aviation%20industry&ved=2ahUKEwiW86\_T9Zr7AhWkl-YKHSzFC9YQsKwBKAB6BAhDEAE
- https://www.google.com/search?lei=cl9oY5byKqSvmgesiqwDQ&q=impact%20of%20covid-19%20on%20aviation%20industry%20research%20paper&ved=2ahUKEwiW86\_T9Zr7AhWkl-YKHSzFC9YQsKwBKAJ6BAhDEAM
- https://dl.acm.org/doi/abs/10.1145/3469028
- https://www.nap.edu/read/21909/chapter/5