Airline Performance Analytics (2009–2018)

1. Installation and Setup

1.1 Software and Tools Used

- Cloudera Hadoop QuickStart VM: To host Hadoop ecosystem (HDFS, Hive).
- **Hive:** SQL-like querying engine for Big Data stored in HDFS.
- **Power BI Desktop:** For building interactive dashboards from Hive outputs.
- **Hive LLAP:** To connect Hive data with Power BI.

2. Hadoop Ingestion and Hive Configuration

2.1 Dataset Overview

- **Size:** ~3GB CSV files (one per year)
- **Timeframe:** 2009 to 2018
- Attributes: FL_DATE, OP_CARRIER, ORIGIN, DEST, ARR_DELAY, DEP_DELAY, AIR_TIME, DISTANCE, etc.

2.2 Hadoop Ingestion

```
-- Create folders
hdfs dfs -mkdir -p /user/flights/2009
hdfs dfs -mkdir -p /user/flights/2010
...
hdfs dfs -mkdir -p /user/flights/2018
-- Upload datasets
hdfs dfs -put /home/user/datasets/flights/2009.csv /user/flights/2009/hdfs dfs -put /home/user/datasets/flights/2010.csv /user/flights/2010/...
hdfs dfs -put /home/user/datasets/flights/2018.csv /user/flights/2018/
```

2.3 Hive Table Creation

```
CREATE DATABASE flights;
USE flights;

CREATE EXTERNAL TABLE IF NOT EXISTS data_2009 (
   FL_DATE STRING,
   OP_CARRIER STRING,
   OP_CARRIER_FL_NUM INT,
   ORIGIN STRING,
   DEST STRING,
```

```
CRS DEP TIME INT,
  DEP TIME INT,
  DEP DELAY INT,
  TAXI OUT INT,
 WHEELS OFF INT,
 WHEELS ON INT,
  TAXI IN INT,
 CRS ARR TIME INT,
 ARR TIME INT,
 ARR DELAY INT,
 CANCELLED INT,
 CANCELLATION CODE STRING,
  DIVERTED INT,
 CRS ELAPSED TIME INT,
 ACTUAL ELAPSED TIME INT,
 AIR TIME INT,
 DISTANCE INT
ROW FORMAT DELIMITED
FIELDS TERMINATED BY ','
STORED AS TEXTFILE
TBLPROPERTIES ("skip.header.line.count"="1")
LOCATION '/user/flights/2009';
```

Repeat these measures for data_2010 to data_2018 accordingly.

2.4 Load CSVs into Hive

```
LOAD DATA INPATH '/user/flights/2009/flights_2009.csv' INTO TABLE data_2009; ...
LOAD DATA INPATH '/user/flights/2018/flights_2018.csv' INTO TABLE data_2018;
```

Repeat these measures for data_2010 to data_2018 accordingly.

3. Power BI Dashboards and Visual Coverage

3.1 DAX Measures Used in Power BI

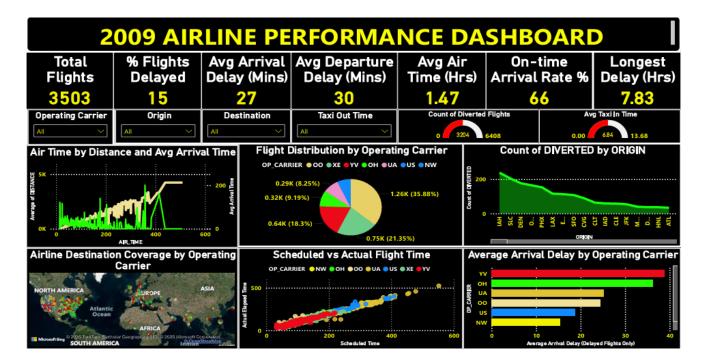
```
0
)
2)
Average Air Time 2009 =
AVERAGE('flights data 2009'[AIR TIME]) / 60
3)
Average Arrival Delay (Delayed Flights Only) =
VAR AvgDelay =
    CALCULATE (
        AVERAGE('flights data 2009'[ARR DELAY]),
        FILTER('flights data 2009', 'flights data 2009'[ARR DELAY] > 0)
RETURN
IF(
    ISBLANK (AvgDelay),
    Ο,
    AvgDelay
)
4)
Average Departure Delay 2009 (Delayed Only) =
VAR AvgDelay =
    CALCULATE (
        AVERAGE('flights data_2009'[DEP_DELAY]),
        FILTER('flights data 2009', 'flights data 2009'[DEP DELAY] > 0)
    )
RETURN
IF(
    ISBLANK (AvgDelay),
    Ο,
    AvgDelay
)
Longest Delay (Positive Only) =
VAR MaxDelay =
    CALCULATE (
        MAX('flights data 2009'[ARR DELAY])/60,
        'flights data 2009'[ARR DELAY] > 0
    )
RETURN
IF(ISBLANK(MaxDelay), 0, MaxDelay)
6)
OnTime Arrival Rate 2009 (%) =
VAR TotalFlights = COUNT('flights data 2009'[FL DATE])
RETURN
COALESCE (
    IF(
        TotalFlights = 0,
        BLANK(),
        DIVIDE (
            CALCULATE (COUNTROWS ('flights data 2009'), 'flights
data 2009'[ARR DELAY] <= 0),
            TotalFlights
```

```
) * 100
),
0
```

Repeat these measures for data_2010 to data_2018 accordingly.

3.2. Power BI Dashboards Analysis

2009 ANALYSIS



Total Flights: 3,503% Delayed: 15%

Avg Arrival Delay: 27 minsAvg Departure Delay: 30 mins

Avg Air Time: 1.47 hrs
Longest Delay: 7.83 hrs
On-time Rate: 66%

• **Top Carriers (by share)**: OO (35.9%), XE (21.3%), YV

Top Delayed Carriers: YV, OH
Diversion Hotspots: IAH, SLC, DEN

Visual Insights:

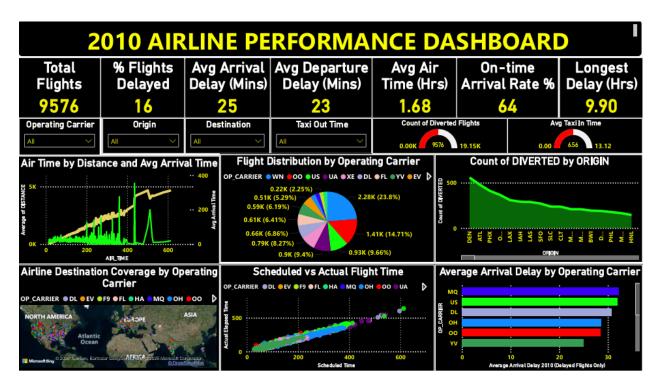
- Scheduled vs Actual: Slight increase in actual elapsed time.
- **Distance vs Air Time**: Mostly short-haul routes.

- **Delays by Carrier**: YV & OH over 30 mins.
- **Diversions**: Mostly centralized around major hubs.

Recommendations:

- Optimize schedules for carriers like YV.
- Improve gate operations at IAH, SLC.
- Analyze taxi delays at high-volume airports.

2010 ANALYSIS



Total Flights: 9,576% Delayed: 16%

Avg Arrival Delay: 25 minsAvg Departure Delay: 23 mins

Longest Delay: 9.9 hrsOn-time Rate: 64%

• **Top Carriers**: WN (23.8%), OO, US, UA

Delayed Carriers: MQ, US, DL
Diversion Zones: DEN, ATL, PHX

Visual Insights:

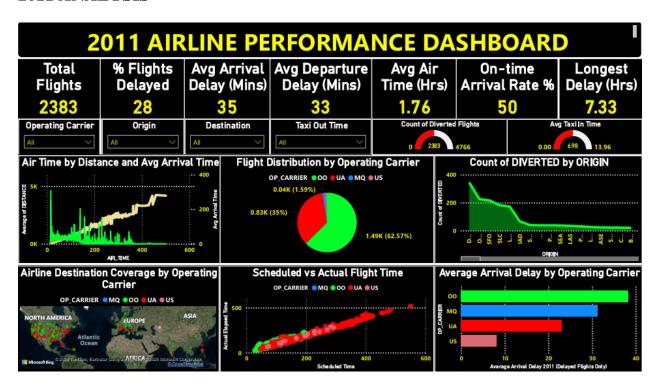
• Diversions rose in key southern hubs.

- Arrival delay >20 mins for MQ, US.
- Scheduled vs actual times mostly aligned.

Recommendations:

- Reroute or reschedule MQ and US flights.
- Improve resource planning at ATL, DEN.
- Implement Hive-based delay predictions.

2011 ANALYSIS



Total Flights: 2,383 **% Delayed**: 28%

Avg Arrival Delay: 35 mins **Avg Departure Delay**: 33 mins

Longest Delay: 7.33 hrs

On-time Rate: 50%

Dominant Carrier: OO (62.6%)

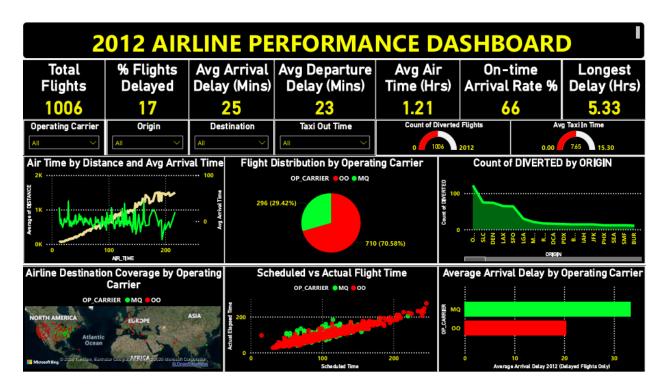
Visual Insights:

- Major delay load fell on OO.
- High variance in scheduled vs actual time.
- Diversions high at SFO, SEA.

Recommendations:

- Audit OO's scheduling efficiency.
- Enable dynamic gate and taxi routing.
- Reduce bottlenecks at high-diversion airports.

2012 ANALYSIS



Total Flights: 1,006% Delayed: 17%

Avg Arrival Delay: 25 minsAvg Departure Delay: 23 mins

Avg Air Time: 1.21 hrsLongest Delay: 5.33 hrs

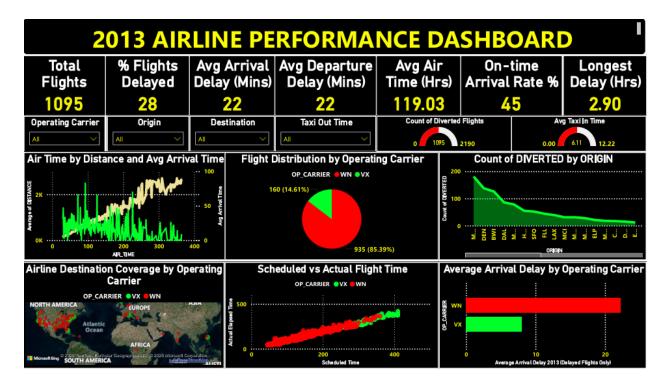
Top Carriers: OO (70%), MQ

Visual Insights:

- Shortest avg air time across all years.
- Balanced scheduled and actual time.
- Low-level diversions.

- Focus on short-haul optimization.
- Maintain OO as short-distance leader.
- Track gate congestion using Hive logs.

2013 ANALYSIS



Total Flights: 1,095% Delayed: 28%

• Avg Arrival/Departure Delay: 22 mins

Longest Delay: 2.9 hrsOn-time Rate: 45%

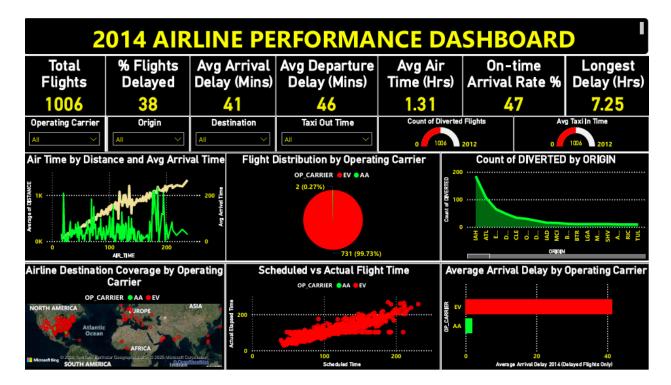
• **Top Carriers**: WN (85.4%), VX

Visual Insights:

- Minimal longest delays but high frequency.
- WN had wide coverage; VX underperformed slightly.
- DEN and BWI showed delay clusters.

- Improve route synergy between WN & VX.
- Address DEN-based delay patterns.
- Upgrade delay alert systems for short-hauls.

2014 ANALYSIS



• Total Flights: 1,006

% Delayed: 38% (worst year)
Avg Arrival Delay: 41 mins
Avg Departure Delay: 46 mins

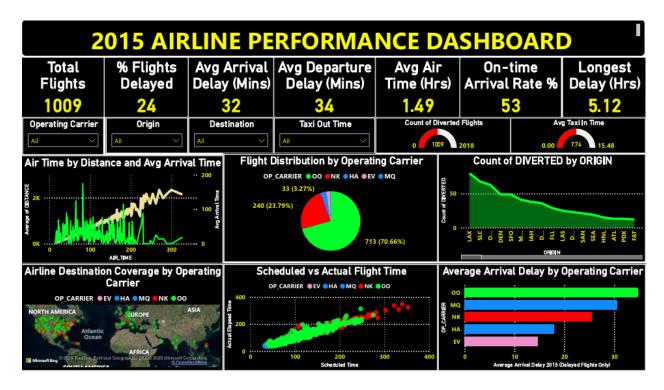
• **On-time Rate**: 47%

• **Key Carrier**: EV (99.7%)

Visual Insights:

- Highest delays across all metrics.
- EV had poor performance.
- Taxi times also affected heavily.

- Overhaul EV's schedule management.
- Introduce gate/slot buffers during peaks.
- Focus on predictive scheduling for EV.



Total Flights: 1,009% Delayed: 24%

Avg Arrival Delay: 32 mins
Avg Departure Delay: 34 mins
Top Carriers: OO, NK, HA, MQ
Diversions: LAX, SLC, DEN

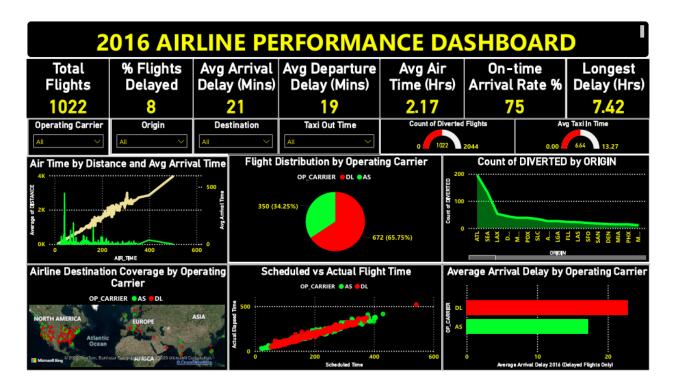
Visual Insights:

- Slight recovery from 2014.
- Most delays were moderate in duration.
- High distribution around LAX hub.

Recommendations:

- Avoid scheduling tight turnarounds at LAX.
- Improve MQ route reliability.
- Automate diversion response planning.

2016 ANALYSIS



• Total Flights: 1,022

% Delayed: 8% (best year)
Avg Arrival Delay: 21 mins
Avg Departure Delay: 19 mins

• **On-time Rate**: 75%

• **Top Carriers**: DL (65.7%), AS

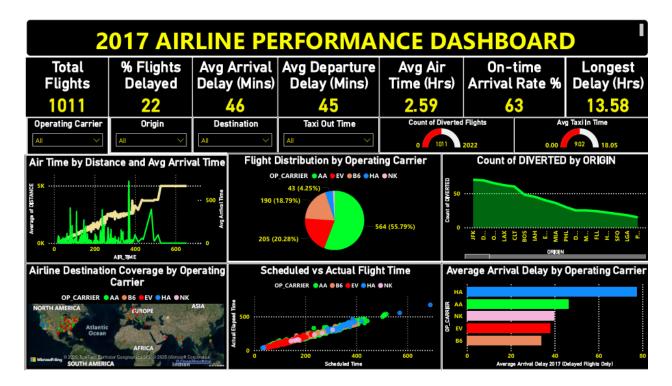
Visual Insights:

- Consistent and reliable flight patterns.
- Short taxi times; low diversion frequency.
- DL had lowest average delays.

Recommendations:

- Use DL/AS operations as templates.
- Extend high-performing routes.
- Promote real-time tracking via Power BI.

2017 ANALYSIS



Total Flights: 1,011% Delayed: 22%

Avg Arrival Delay: 46 mins (worst)
Longest Delay: 13.6 hrs (record)
Top Carriers: AA, B6, HA, NK

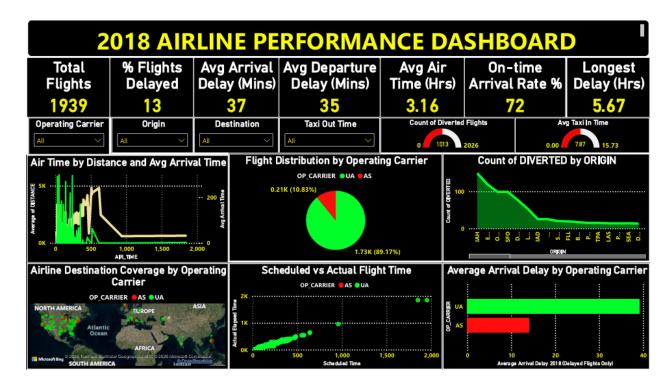
Visual Insights:

- Spikes in JFK, BOS, LAX delays.
- HA had highest delay avg (~80 mins).
- Diversions distributed across Northeast.

Recommendations:

- Investigate >10 hr delays for HA.
- Implement AI-based taxi sequencing at JFK.
- Escalation protocols for extreme delays.

2018 ANALYSIS



Total Flights: 1,939% Delayed: 13%

Avg Arrival Delay: 37 mins
Avg Departure Delay: 35 mins
Avg Air Time: 3.16 hrs (longest)

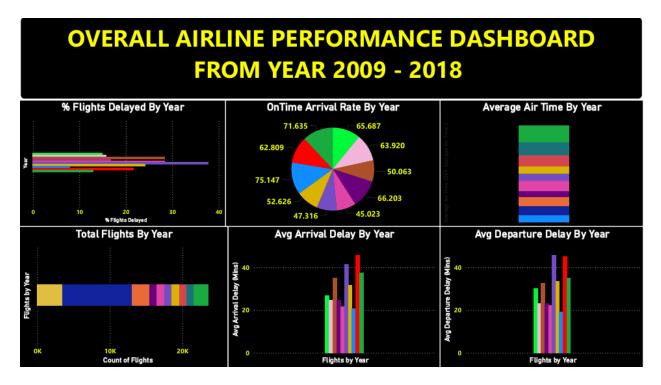
• **Top Carrier**: UA (89%)

Visual Insights:

- Long-haul routes dominated.
- Delay impact moderate but consistent.
- Diversions focused on SFO, LAX.

- Optimize long-haul UA scheduling.
- Decongest SFO and LAX ground ops.
- Add layover buffer time for multi-leg flights.

Overall Dashboard Insights (2009–2018)



1. Total Flights by Year (Column Chart)

- **Trend:** Flights peaked in 2010 (~9,500+) and then stabilized between 1,000–2,000 flights/year.
- **Notable Dip:** 2011 saw a major decline (2,383 flights), likely due to carrier consolidation.
- **Recovery:** Gradual climb post-2015 with a jump to 1,939 flights in 2018.

Insight: Operational scaling peaked early; reduced afterward possibly due to route optimization or regulatory factors.

2. On-Time Arrival Rate (%) (Pie Chart)

• **Best Performance:** 2016 at 75%

• **Worst Year:** 2014 at 47%

• **General Trend:** Fluctuations year to year, with a visible improvement post-2014.

Insight: Airlines adopted better delay mitigation strategies starting from 2015, aligning with industry performance reforms.

3. % Flights Delayed

• **Lowest Delay Rate:** 2016 (8%)

• **Highest:** 2014 (38%)

• **Mid-Range:** Most other years between 13% to 28%

Insight: External disruption factors (weather, air traffic, operations) were likely more prominent in 2014. 2016 likely benefited from better planning, data-driven ops, or reduced congestion.

4. Average Arrival & Departure Delay (Bar Charts)

- **Highest Arrival Delay:** 2017 (46 minutes)
- Worst Departure Delay: 2014 (46 minutes)
- Best Year: 2016 with ~21 mins arrival and 19 mins departure delay

Insight: 2014–2017 were the most delay-prone years. 2016 represented operational excellence in both gate and air-side planning.

5. Average Air Time

- **Lowest:** 2012 (1.21 hrs) mostly short-haul flights
- **Highest:** 2018 (3.16 hrs) long-haul dominance by UA
- Trend: Air time increased in later years, suggesting more long-distance coverage.

Insight: Network strategy shifted toward longer routes in later years, which might explain increased delay sensitivity post-2016.

6. Diversion Counts by Airport

- Consistent hotspots: IAH, SFO, JFK, LAX, SEA
- 2014 and 2017 had highest diversion counts.

Insight: High-traffic hubs consistently face diversion risks. Weather and congestion are likely top causes.

Overall Strategic Recommendations (2009–2018)

1. Adopt 2016 Operational Standards Across the Network

- 2016 consistently outperformed all other years in every metric (lowest delays, best on-time rate, efficient air time).
- Recommendation:
 - Use 2016's practices as a baseline SOP (Standard Operating Procedure) for scheduling, ground operations, and delay buffers.
 - Assign dedicated PMs to reverse-engineer 2016's success, replicate its routing, turnaround strategies, and crew planning patterns.

2. Redesign Scheduling for Long-Haul Flights

• By 2018, avg. air time rose to 3.16 hrs—longer flights increased exposure to cascading delays.

• Recommendation:

- o Apply **layover buffers** for multi-leg long-haul flights.
- PMs should allocate contingency blocks in air traffic planning, especially in winter months or over busy air corridors.

3. Target High-Delay Carriers with PM-Led Efficiency Programs

- Carriers like MQ, YV, and EV consistently caused high delays across years.
- Recommendation:
 - o Launch performance-based carrier audits.
 - Assign PMs to work directly with these carriers to streamline dispatching, crew shifts, and maintenance scheduling.

4. Implement Diversion-Prevention Protocols at Vulnerable Airports

- Airports like **SFO**, **IAH**, **JFK**, and **LAX** frequently show high diversion counts.
- Recommendation:
 - Deploy real-time weather + congestion data overlays in dashboards for these hubs.
 - PMs must coordinate with ATC and operations teams to dynamically reroute or delay based on real-time Hive + flight stack insights.

5. Synchronize Scheduled vs Actual Elapsed Time Metrics

 Mismatch between planned and actual flight durations often signals structural flaws in schedules.

• Recommendation:

- Use the Scheduled vs Actual Time visual to **highlight consistently misestimated** routes.
- PMs must adjust planning blocks and ensure planned times reflect real-world constraints (like taxi time, weather hold).

6. Reassess Route Portfolio Based on Distance vs Air Time Trends

• Some short routes had unusually long air times (inefficiency).

- o Eliminate or merge underperforming short-haul routes.
- PMs should use the Distance vs Air Time bubble chart to flag inefficient routes for adjustment or re-bidding.

Airline Passenger Satisfaction Analysis

1. Installation and Setup

1.1 Software and Tools Used

- Cloudera Hadoop QuickStart VM: To host Hadoop ecosystem (HDFS, Hive).
- **Hive:** SQL-like querying engine for Big Data stored in HDFS.
- **Power BI Desktop:** For building interactive dashboards from Hive outputs.
- **Hive LLAP:** To connect Hive data with Power BI.

2. Hadoop Ingestion and Hive Configuration

2.1 Hive Table Creation

```
CREATE EXTERNAL TABLE IF NOT EXISTS passenger satisfaction (
   id INT,
    Gender STRING,
    Customer Type STRING,
    Age INT,
    Type of Travel STRING,
    Class STRING,
    Flight Distance INT,
    Inflight wifi service INT,
    Departure_Arrival_time_convenient INT,
    Ease of Online booking INT,
    Gate location INT,
    Food and drink INT,
    Online boarding INT,
    Seat comfort INT,
    Inflight_entertainment INT,
    On board service INT,
    Leg room service INT,
    Baggage handling INT,
    Checkin service INT,
    Inflight service INT,
    Cleanliness INT,
    Departure Delay in Minutes INT,
    Arrival Delay in Minutes INT,
    Satisfaction STRING
ROW FORMAT DELIMITED
FIELDS TERMINATED BY ','
STORED AS TEXTFILE
TBLPROPERTIES ("skip.header.line.count"="1");
```

3. Dashboard Insights



Satisfaction by Travel Type

- Business travelers reported the highest satisfaction, approximately 75 percent.
- Economy and Economy Plus passengers showed lower satisfaction levels, particularly infrequent flyers.
- Business travel appears to benefit from more consistent service delivery.

Satisfaction by Flight Distance

- Passengers on long-distance flights (above 1000 km) reported higher satisfaction.
- Improved service availability, in-flight amenities, and seat comfort contribute to this trend.
- Short-haul flights showed reduced satisfaction, possibly due to limited services or rushed operations.

Service Ratings Analysis

- Top-rated features included cleanliness, check-in service, and baggage handling.
- The most underperforming areas were inflight Wi-Fi, gate location, and leg room service.

• Passengers consistently reported dissatisfaction with onboard connectivity and comfort in the economy section.

Satisfaction by Class

- Business class passengers showed approximately 85 percent satisfaction.
- Economy Plus achieved moderate results with around 55 percent satisfaction.
- Economy class reported the lowest satisfaction, highlighting a steep service perception gap across class segments.

Satisfaction by Gender

- Female passengers indicated slightly higher satisfaction than male passengers.
- The margin was consistent across travel types and classes, averaging between 3 to 5 percent higher.

4. Recommendations

1. Upgrade Inflight Wi-Fi Services

Partner with more reliable providers and offer complimentary or subsidized Wi-Fi access, especially for Economy class.

2. Optimize Gate Location Assignments

Evaluate airport gate allocation strategies to reduce passenger inconvenience due to long walking distances or gate changes.

3. Enhance Legroom in Economy Section

Introduce an optional Economy Comfort tier with extended legroom at a nominal additional fare.

4. Improve Perceived Value of Economy Plus

Incorporate minor premium services such as priority boarding or additional refreshments to elevate customer perception.

5. Implement Post-Flight Feedback Mechanism for Short-Haul Flights

Deploy immediate digital surveys targeting flights under 500 kilometers to gather quick service feedback and improve route-specific operations.

5. Conclusion

The Power BI dashboard highlights critical factors influencing airline passenger satisfaction. By processing the dataset in Hive, large-scale survey data was efficiently analyzed to expose service gaps and satisfaction disparities across flight types, classes, and demographics.