

The background of the slide is a photograph taken from an airplane window. It shows the wing of the aircraft on the left side, extending towards the center. The sky is a mix of orange, yellow, and blue, indicating a sunset or sunrise. The clouds are wispy and scattered across the sky.

Optimizing Air Travel: A Data-Driven Approach to Flight Delay Analysis and Prediction

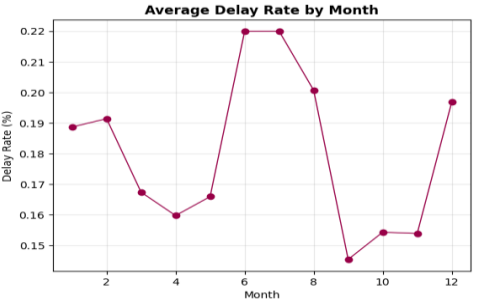
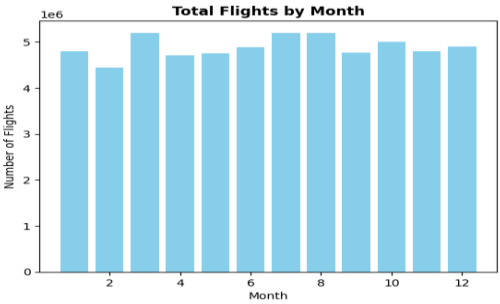
Meenakshi
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ECE(4y)

EXPLORATORY DATA ANALYSIS



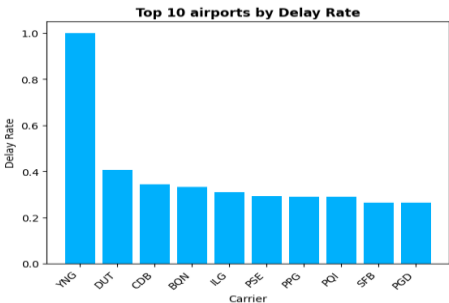
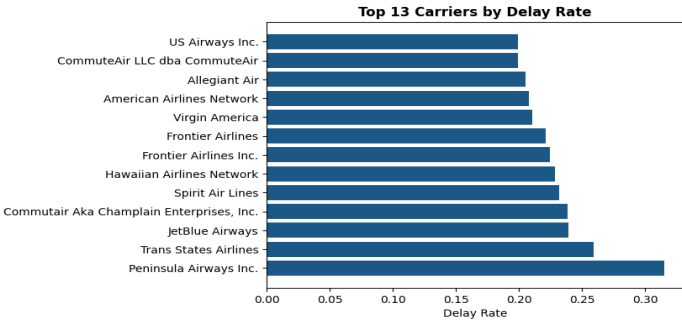
Track Monthly Delay Trends

Visualize how average delays change over time.



Identify Top Airlines & Airports

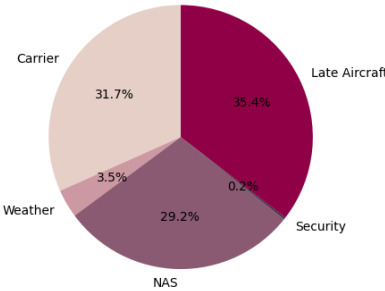
Pinpoint carriers and airports most affected by delays.



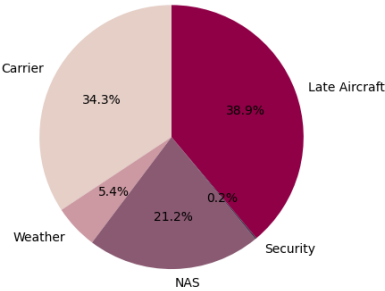
Understand Seasonal Impact

Assess how seasonal changes influence delay frequency.

Distribution of Delay Causes

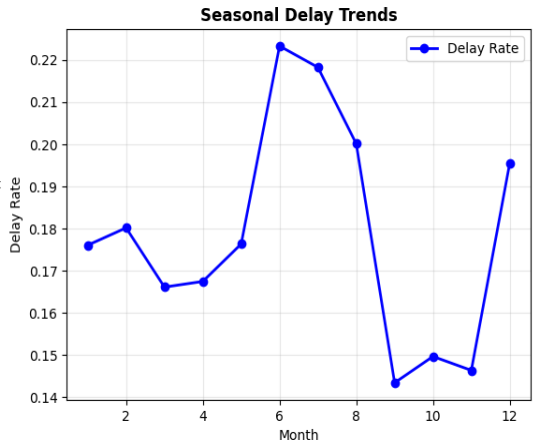


Distribution of Delay Duration



Conduct Root Cause Analysis

Examine specific delay types: carrier, weather, NAS, security, and late aircraft.



Advanced Delay Insights



Average Delay by Cause:

Bar chart showing mean delay contributions

from:



Correlation Heatmap:

Reveals interrelationships among different delay factors.



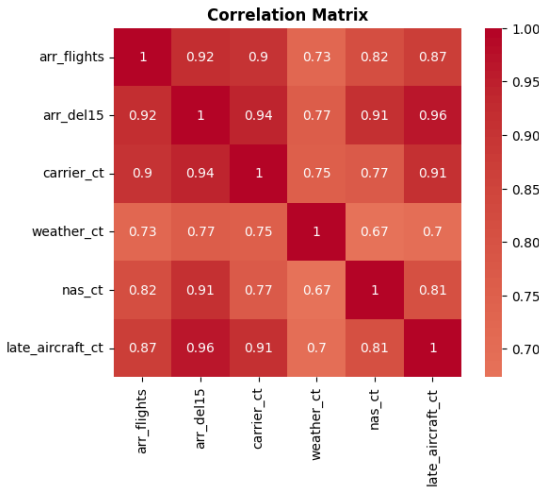
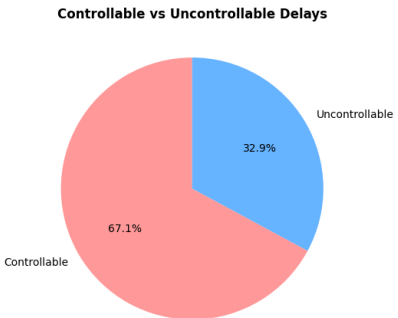
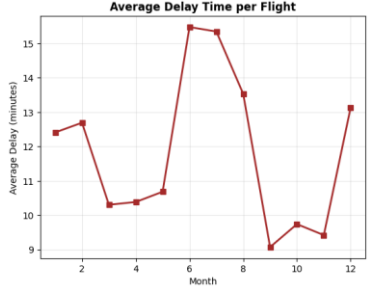
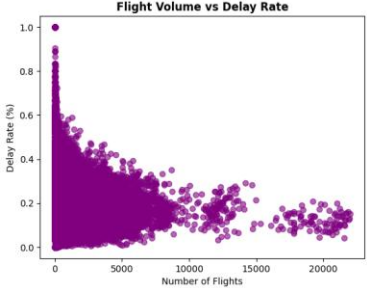
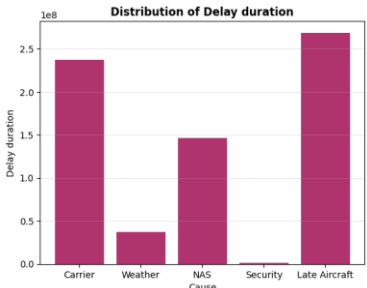
Delay Rate Distribution:

Histogram showing how delay rates vary across all flights.

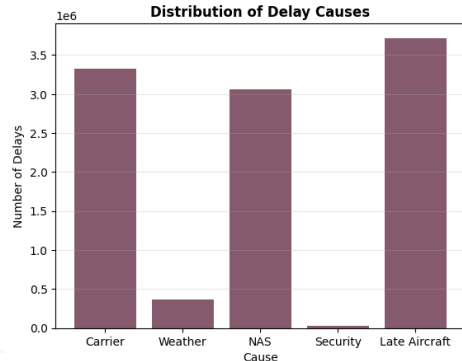


Summary Statistics:

Displayed key metrics like mean, median, min, max, and standard deviation of delay rates.



- SEASONAL PATTERNS:
- Month with highest average delay: July (0.22%)
 - Month with lowest average delay: September (0.15%)
- OVERALL STATS:
- Total flights analyzed: 58,676,070.0
 - Delayed flights: 10,494,073.0
 - Overall delay rate: 17.88%
- CARRIER PERFORMANCE:
- Most delay-prone carrier: Peninsula Airways Inc. (0.32%)
 - Most reliable carrier: Cape Air (0.08%)



Feature Engineering

1. Core Performance Metrics

- `delay_rate` : Proportion of delayed flights
- `avg_delay` : Average delay per flight (in minutes)

2. Delay Cause Proportions

- `carrier_prop` : Delay share due to airline
- `weather_prop` : Delay share due to weather

3. Target Variables

- `is_delayed` : Binary label for delay presence
- `target_delay` : Average delay for regression

4. Historical Reliability Features

- `carrier_his` : Historical delay rate per airline
- `airport_his` : Historical delay rate per airport

5. Encoded Categorical Variables

- `carrier_encoded` : Encoded airline names
- `airport_encoded` : Encoded airport names

Cleaning & Filtering:



Removed rows with missing values in key delay-related columns.



Excluded records where arr_flights = 0 to avoid invalid calculations.



Outliers in continuous variables such as avg_delay and delay_rate were treated using the **IQR (Interquartile Range) method**, ensuring robust model performance without removing valid extreme delay instances.

Operational Adjustability Index (OAI)



What is OAI?

A custom metric to measure how much of the delay is operationally controllable.



Formula

$$\text{OAI} = (\text{carrier_delay} + \text{late_aircraft_delay}) / \text{total_delay}$$

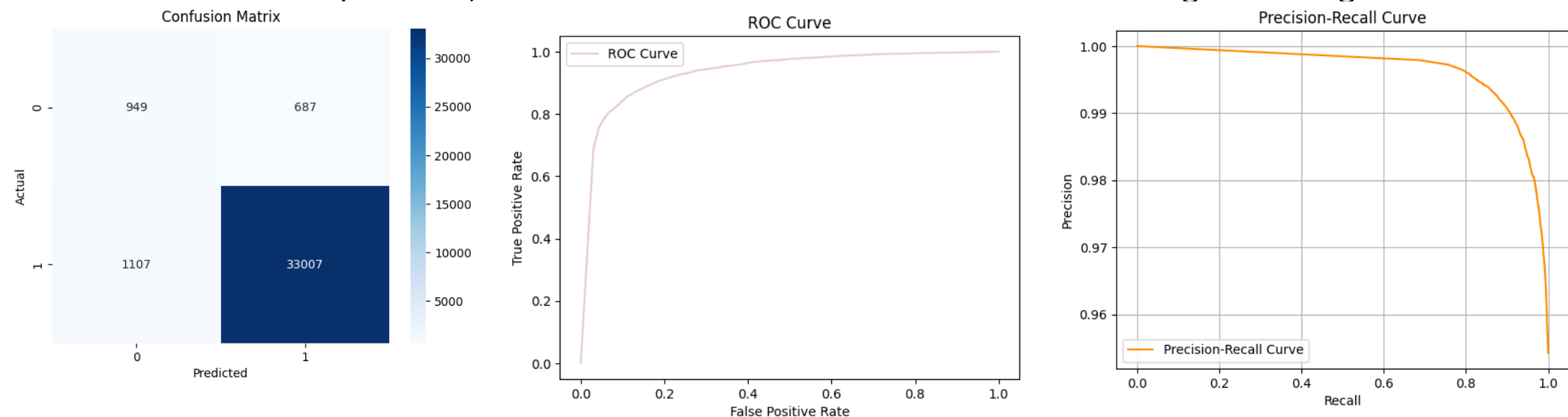

It helps airline ops teams prioritize improvements on controllable delays.

High OAI → Focus on scheduling, fleet readiness, crew management

Low OAI → Indicates external issues like weather/NAS delays

Classification model for Flight Delays

- For the classification problem , **Random Forest classifier** is used after balancing data using **SMOTE**

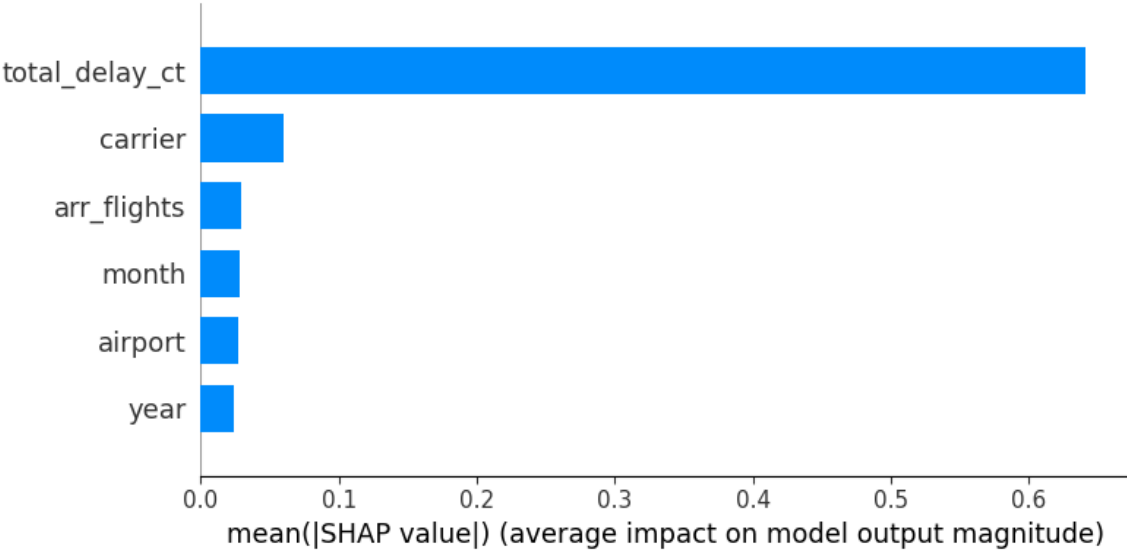


Evaluation Metrics:

Accuracy: 0.9498181818181818
Precision: 0.9796106131655488
Recall: 0.9675499794805652
F1 Score: 0.973542944785276
ROC AUC: 0.9346907349197204
Confusion Matrix:
[[949 687]
 [1107 33007]]

Classification Report:

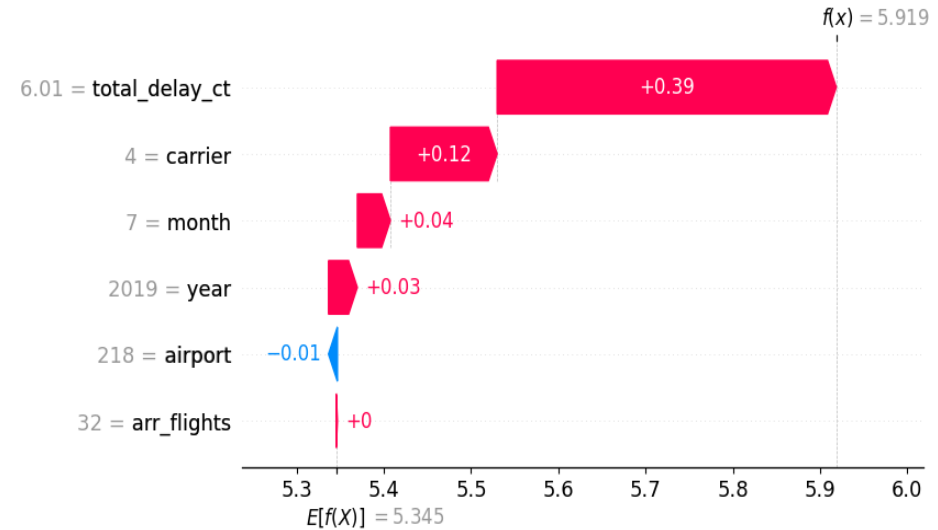
	precision	recall	f1-score	support
0	0.46	0.58	0.51	1636
1	0.98	0.97	0.97	34114
accuracy			0.95	35750
macro avg	0.72	0.77	0.74	35750
weighted avg	0.96	0.95	0.95	35750



Predicting Arrival Delays with different Models

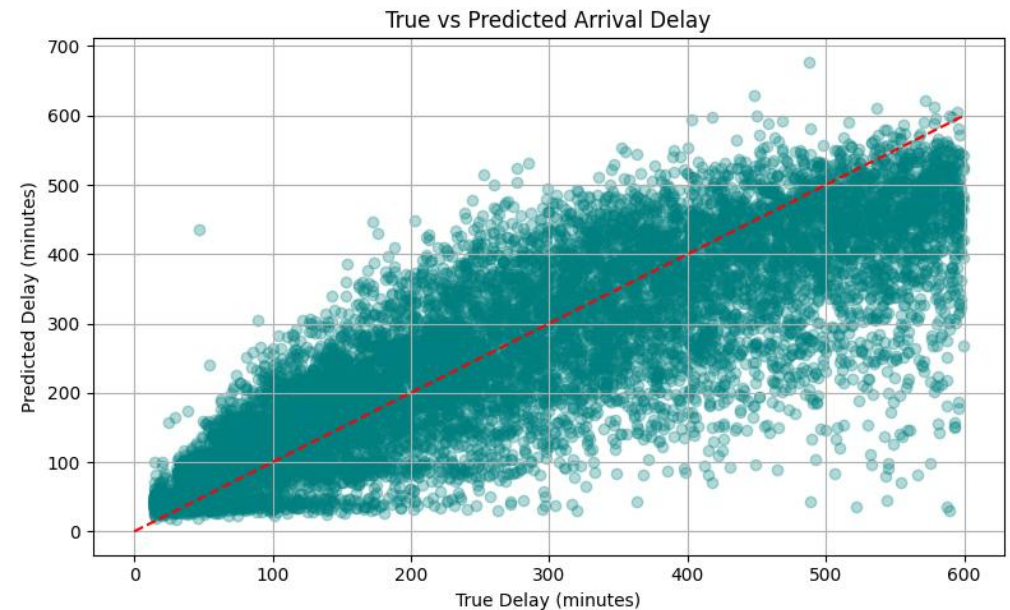
Linear regression model

- Mean Absolute Error (MAE): 7.20
- Mean Squared Error (MSE): 236.01



XG BOOST Regression model

Mean Absolute Error (MAE): 67.65 minutes
Root Mean Squared Error (RMSE): 91.80 minutes
R2 Score: 0.71



Actionable Recommendations & Consulting Insights

1

Improve Ground Operations for Controllable Delays

- Focus on reducing **carrier-related** and **late aircraft delays**, as these are the most controllable and contribute significantly to total delay time.
- Implement tighter turnaround management and buffer times in scheduling.

2

Schedule Adjustments and Network Planning

- Reschedule flights during peak congestion hours or airports with **historically high delays**.
- Use insights from seasonal features (e.g., more delays during summer) to plan better staffing and resource deployment.

3

Prioritize Airports and Carriers with High Delay Rates

- Use the **carrier_historical_performance** and **airport_historical_performance** metrics to identify consistently underperforming segments.
- Initiate targeted interventions like performance audits or collaborative improvement initiatives.

4

Enhance Real-Time Communication & Customer Engagement

- Improve **proactive communication** with passengers during expected NAS/weather-related delays.
- Deploy digital tools (SMS/Email/Apps) for early alerts and rebooking options.

5

Data-Driven Resource Allocation

- Allocate more operational resources (staffing, gates, crew) during **holiday/winter peaks** or **high OAI score periods**.
- Monitor **OAI-weighted duration** to dynamically adjust schedules.