



Airline



SC1015

Mini-Project

Predicting Flight Delays



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BOARDING PASS

● GROUP

FCS2

● VENUE

HWLAB 3

● DATASET

KAGGLE





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01 Problem Formulation



Flight Delay:

>15 MINS



On-time:

~80%



Average length:

~50 MINS



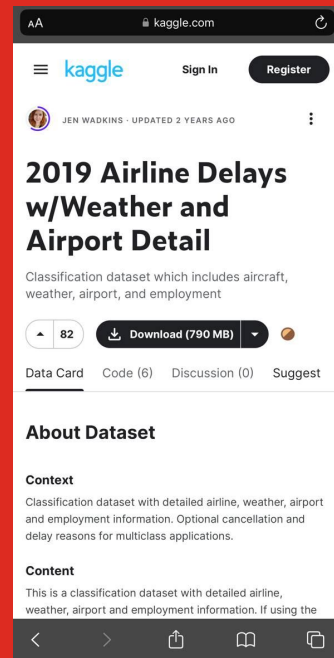
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Dataset

- 8 data files
- Overlapping variables
- All data files connected
- Extensive data preparation and cleansing techniques needed

Source





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Data Preparation

**Handling NaN
values**

**Feature
Engineering**

**Recheck for
NaN values**

**Encode categorical
variables**

**Drop irrelevant
columns & duplicates**

**Changing Data
Types**

Merging into one



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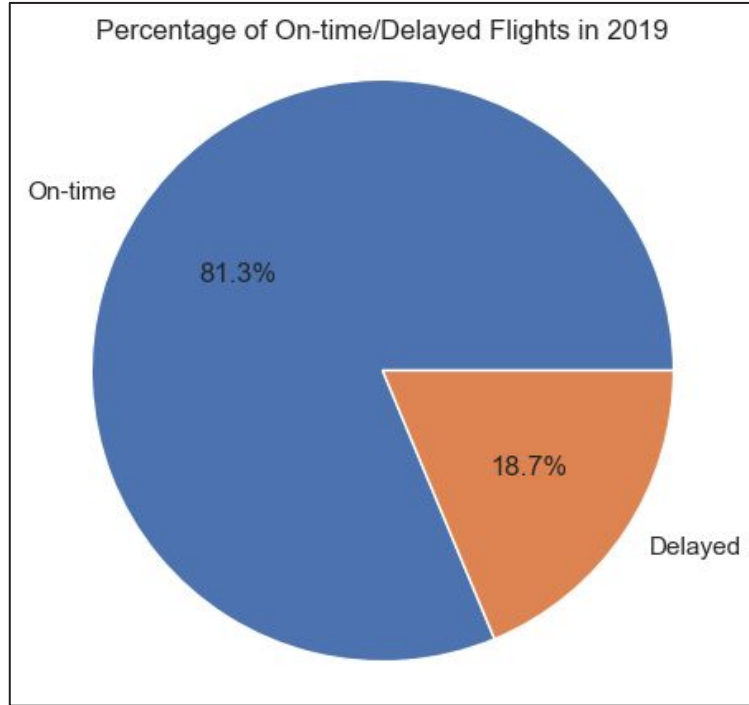
EDA

|02



Exploratory Data Analysis



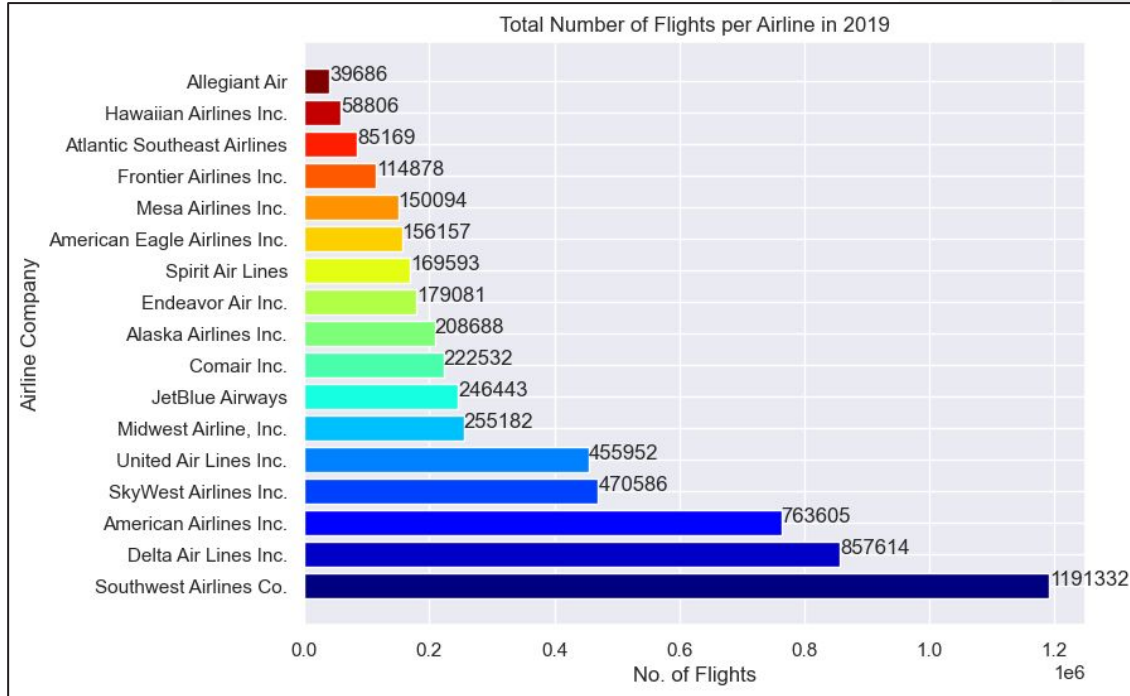


On-time vs Delayed

| ON-TIME: 4.5 million flights

| DELAYED: 1 million flights

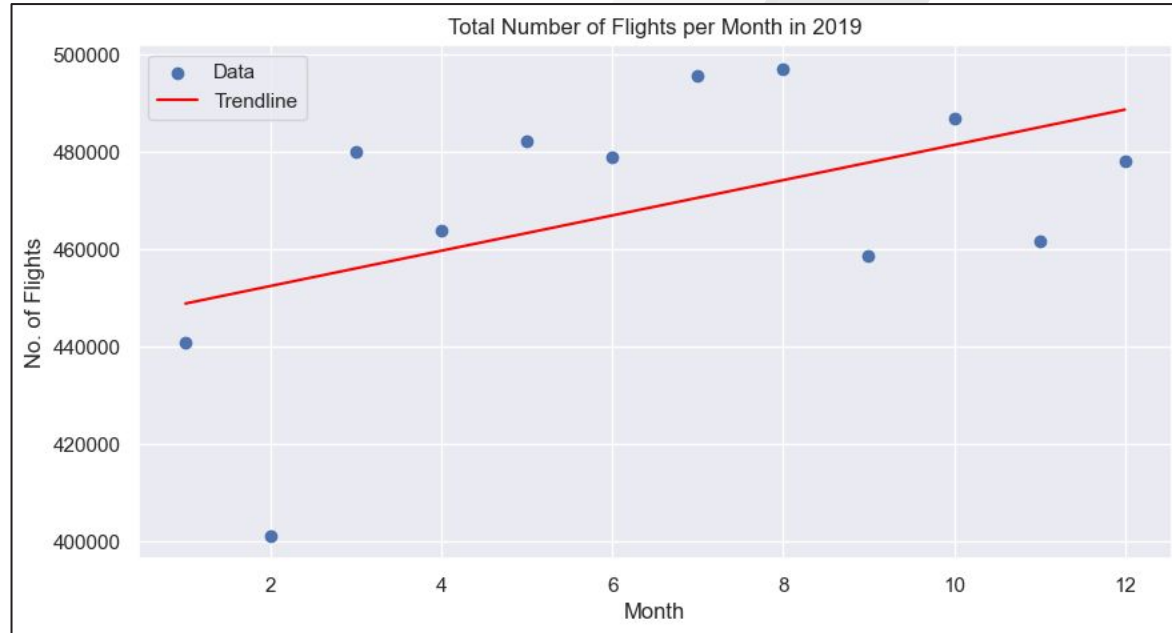
EXPLORATORY DATA ANALYSIS



Airlines w/ Most Flights

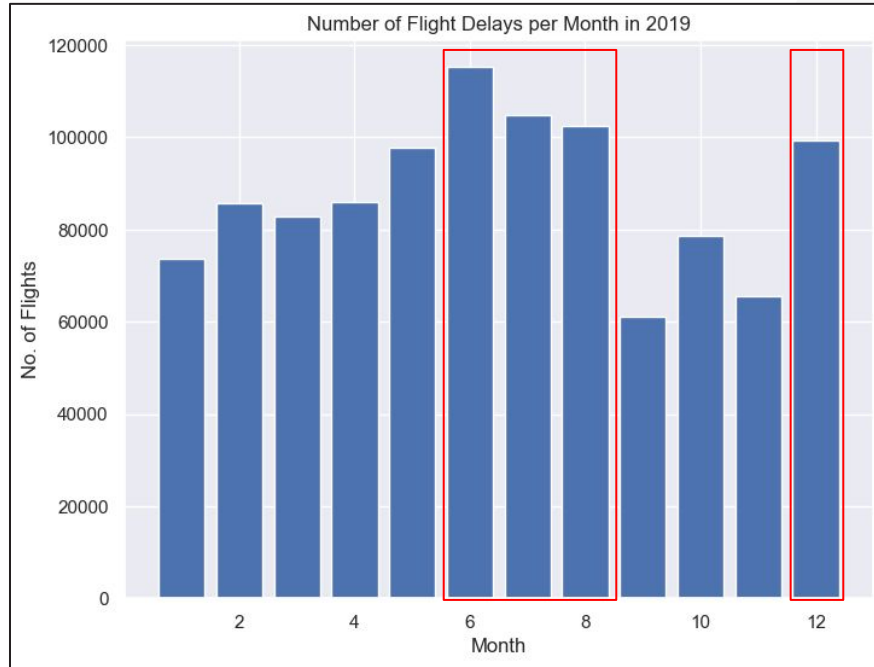
1. Southwest Airline Co.
 - 1.1M / 21.2%
2. Delta Air Lines Inc.
 - 850k / 15.2%
3. American Airlines Inc.
 - 760k / 13.6%

EXPLORATORY DATA ANALYSIS



Increasing trend in the number of flights per month

Which month has the most flight delays?

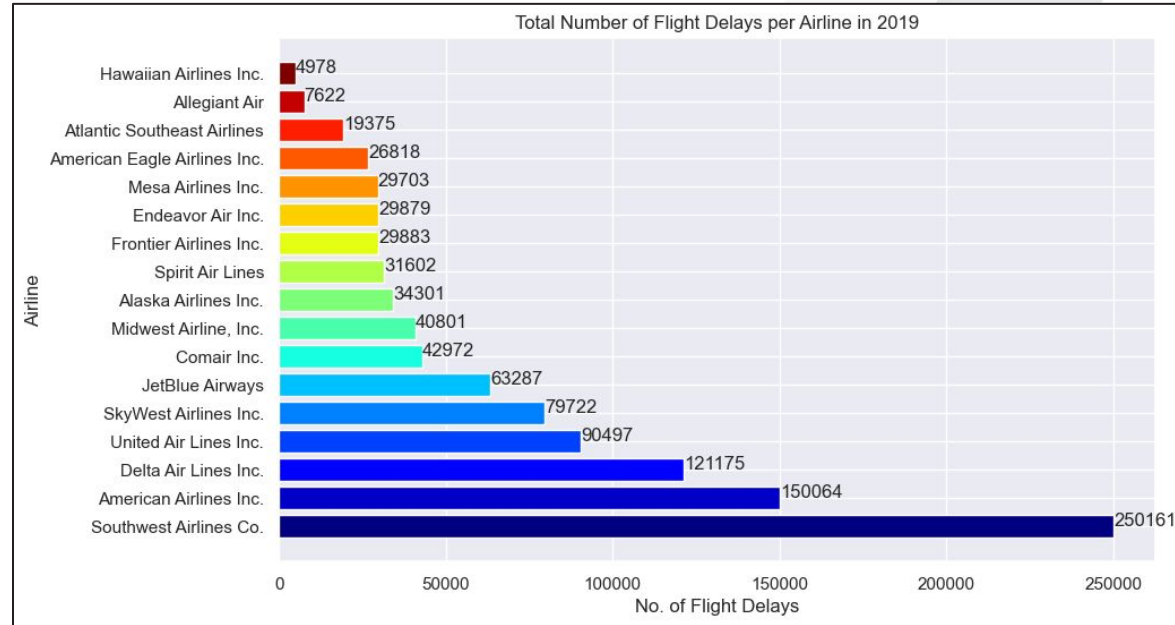


1. June, July & August (Summer)
2. December (Winter)

Possible reasons:

- Adverse weather conditions
- Vacation period
- High number of concurrent flights

Which airline has the most flight delays?



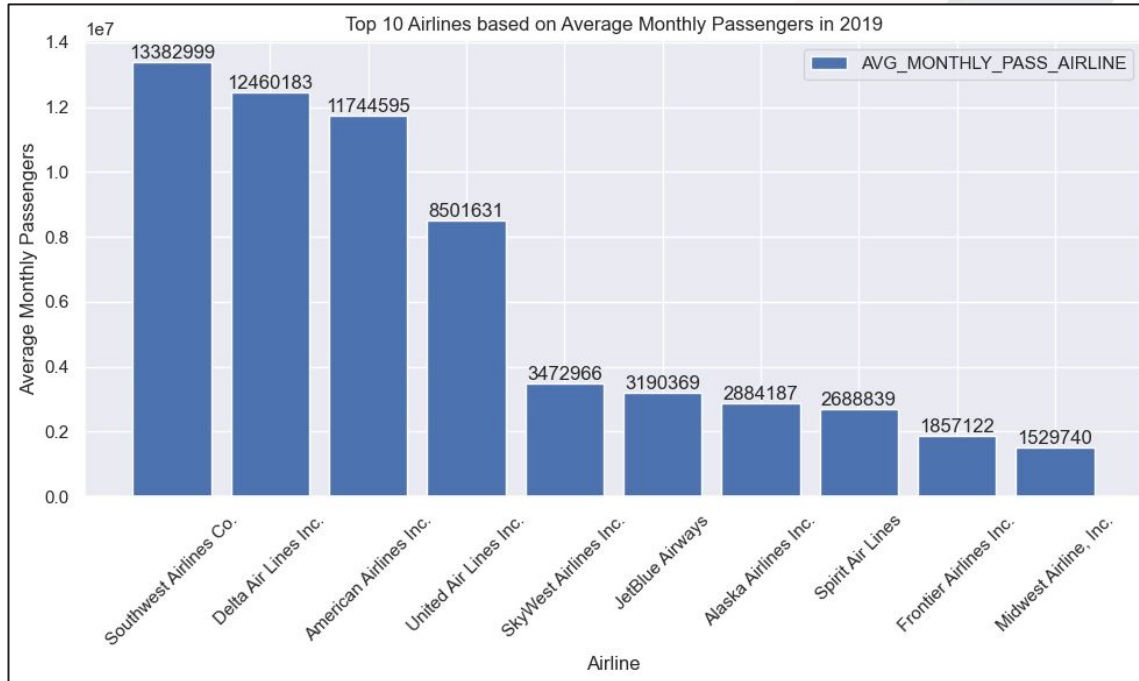
Most number of flight delays:

Southwest Airlines Co.

Possible reasons:

- Budget airline
- Popular

Most Popular Airlines



1. Southwest Airlines Co.
 2. Delta Air Lines Inc.
 3. American Airlines Inc.
- Major American airlines



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| 03 Models



1st

| Decision Tree

2nd

| Random Forest

3rd

| AdaBoost



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4th

| XGBoost

5th

| Naive Bayes

**Comparison
Summary**



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Metrics used for models



| **Accuracy**

% of correct classifications



| **Precision**

% of true positive / (true pos + false **pos**)



| **Recall**

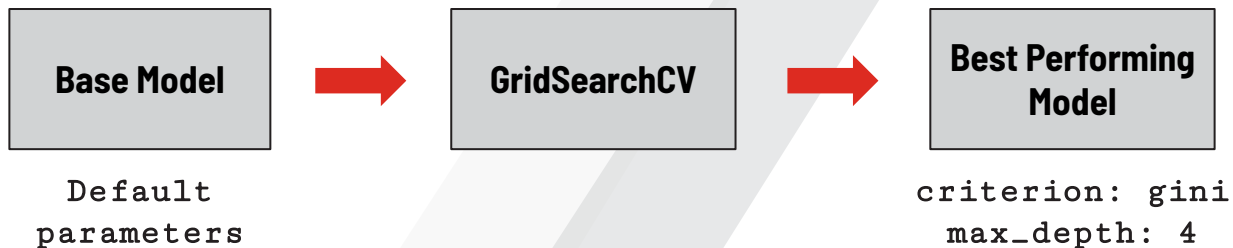
% of true positive / (true pos + false **neg**)



| **F1 Score**

Mean of precision and recall

Decision Tree Classifier



Hyperparameter Tuning using
GridSearchCV:

- To obtain the best parameters to generate the model with the highest accuracy

Accuracy	81.3%
Precision	67.0%
Recall	0%
F1-Score	0%

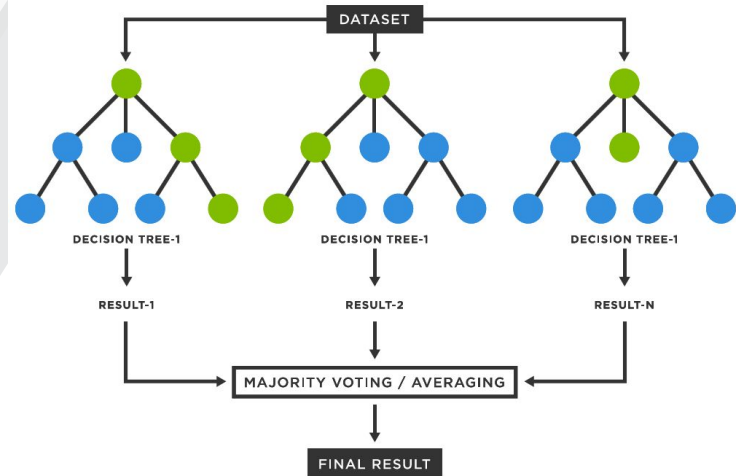
Random Forest Classifier

Base Model

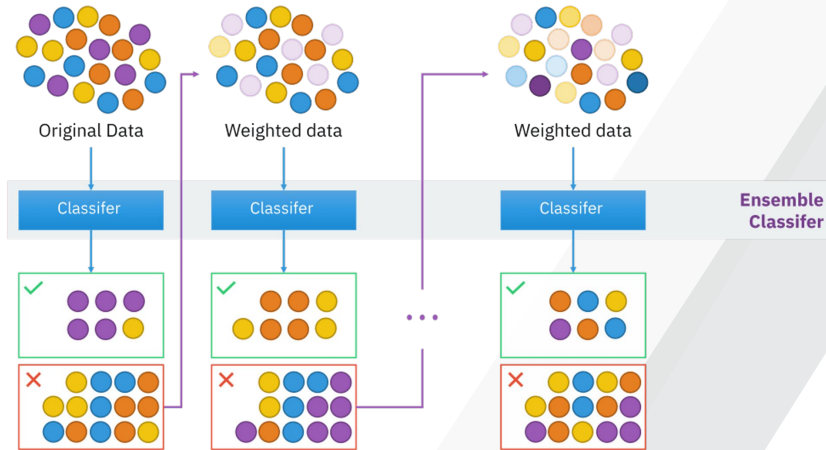


K-folds Cross Validation

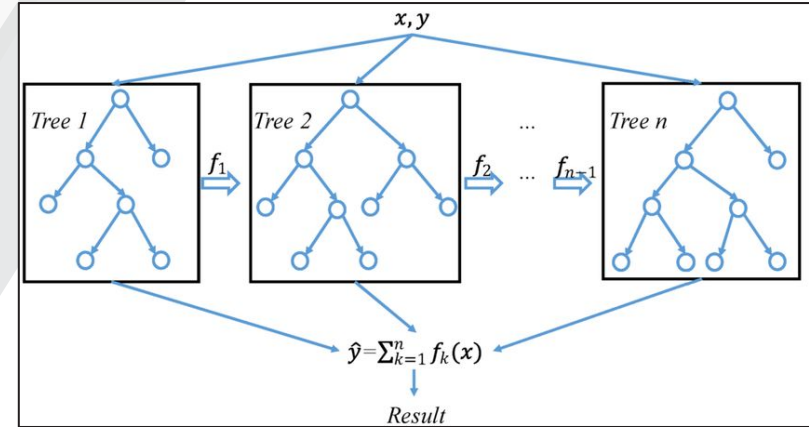
- Combination of decision trees
- Prevent overfitting
- Result = avg no. of models, evaluated k times



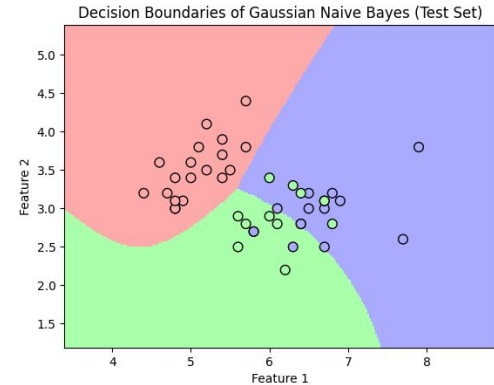
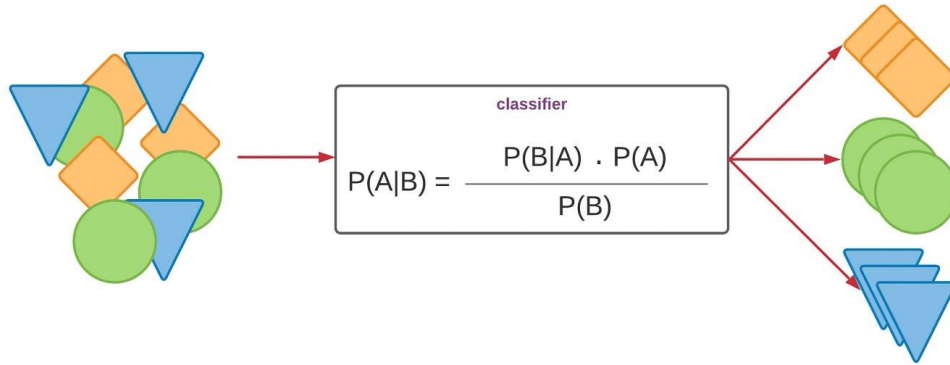
AdaBoost Classifier



XGBoost Classifier



Naive Bayes Classifier



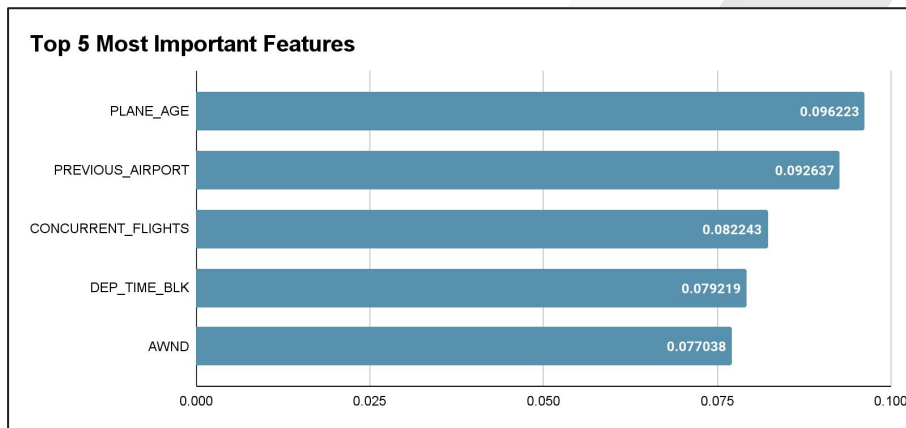
- Supervised learning
- Probabilistic approach + Gaussian distribution

Summary of Classification Models

Model	Train Accuracy	Test Accuracy	Precision	Recall	F1-Score
Decision Tree	81.3%	81.3%	67.0%	0.0	0.0
Random Forest	99.3%	82.0%	55.5%	20.1%	29.5%
AdaBoost	81.3%	81.3%	54.4%	0.0%	1.0%
XGBoost	82.4%	82.3%	66.1%	10.7%	18.4%
Naive Bayes	81.3%	81.3%	0.0%	0.0%	0.0%

Feature Importances on the Best Model

Model	Train Accuracy	Test Accuracy	Precision	Recall	F1-Score
Random Forest	99.3%	82.0%	55.5%	20.1%	29.5%



1. Age of Plane
2. Previous Airport
3. Concurrent Flights
4. Departure Time Block
5. Max Wind Speed



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Data Driven Insights & Recommendations



| Travellers

Choose less
popular seasons



| Airlines

Prioritize newer
aircraft models



| Airport

Interact with each
other



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Thank You !



References:

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