

PROJECT REPORT

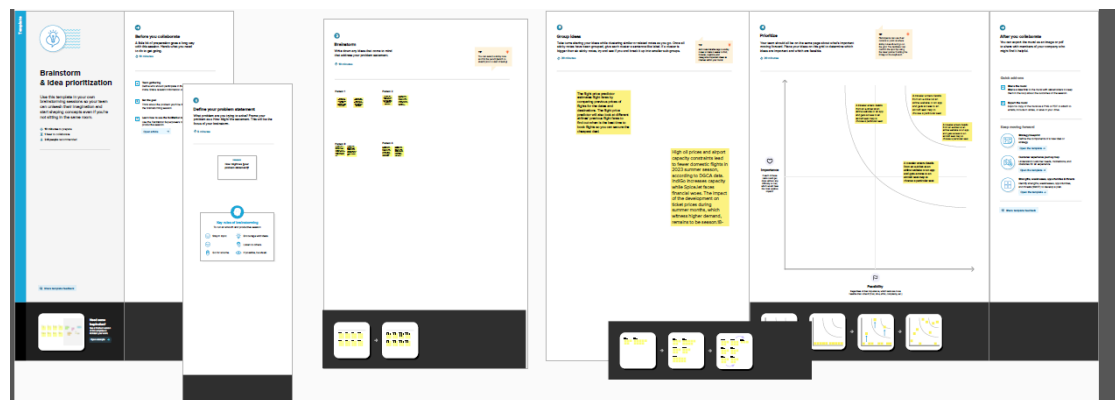
1. Introduction

1.1 Overview

There has been a remarkable expansion in commercial aviation in the past decade, with more people preferring air travel for a fast and comfortable journey. However, flight delays have become quite common across the world with the growth of the aviation sector. Besides inconvenience to the travellers, flight delays have a negative impact on the economy. The airline companies incur substantial monetary losses and observe a fall in their reputation if their flights are delayed often. The unforeseen delays also have a cascading effect on various other sectors. According to a report by the Joint Economic Committee of United States Congress, the total cost of flight delays to the US economy was over \$40 billion with \$19 billion to the airlines, \$12 billion to the passengers and around \$10 billion to other industries. The delayed flights also pose certain environmental concerns. Delayed flights consumed an additional 740 million units of jet fuel and released over 7 million metric tonnes of additional Carbon Dioxide[1].

1.2 Purpose

- Flight delays are gradually increasing and bring more financial difficulties and customer dissatisfaction to airline companies. To resolve this situation, supervised machine learning models were implemented to predict flight delays.
- Therefore, predicting flight delays can improve airline operations and passenger satisfaction, which will result in a positive impact on the economy. In this study, the main goal is to compare the performance of machine learning classification algorithms when predicting flight delays.
- Flight delays not only irritate air passengers and disrupt their schedules but also cause a decrease in efficiency, an increase in capital costs, reallocation of flight crews and aircraft, and additional crew expenses



3.Result

3.1 Date Model



Prediction of Flight Delay

Enter the Flight Number :

Month :

Day of Month :

Day of Week :

origin

destination

Scheduled Departure Time :

Scheduled Arrival Time :

Actual Departure Time :



Prediction of Flight Delay

Enter the Flight Number :

Month :

Day of Month :

Day of Week :

origin

destination

Scheduled Departure Time :

Scheduled Arrival Time :

Actual Departure Time :



The screenshot shows a web browser window with the address bar displaying 'localhost:5000/prediction'. The page has a dark teal background with a large, stylized yellow airplane flying over a clock face. The title 'Prediction of Flight Delay' is centered at the top in white. Below the title, there is a form with the following fields: 'Enter the Flight Number :', 'Month :', 'Day of Month :', 'Day of Week :', 'origin' (a dropdown menu showing 'MSP'), 'destination' (a dropdown menu showing 'MSP'), 'Scheduled Departure Time :', 'Scheduled Arrival Time :', and 'Actual Departure Time :'. A yellow 'SUBMIT' button is located below the form fields. At the bottom left, a red-bordered box contains the text 'The Flight will be on time'.

2. Advantages & Disadvantages

* Effective prediction of flight delays will not only reduce the economic losses caused by airline delays, but also reduce the stay time and improve the traveling satisfaction of passengers.

* Carriers attribute flight delays to several causes such as bad weather conditions, airport congestion, airspace congestion, and use of smaller aircraft by airlines. These delays and cancellations tarnish the airlines' reputation, often resulting in loss of demand by passengers.

6 . Appendix

```
#!/usr/bin/env python
```

```
# coding: utf-8
```

```
# In[1]:
```

```
from flask import Flask, request, jsonify, render_template,  
url_for , request  
  
import pickle  
  
from sklearn.preprocessing import LabelEncoder  
  
from sklearn.model_selection import train_test_split  
  
import pandas as pd  
  
# Import dataset  
  
df = pd.read_csv('Data/Processed_data15.csv')  
  
# Label Encoding  
  
le_carrier = LabelEncoder()  
  
df['carrier'] = le_carrier.fit_transform(df['carrier'])  
  
le_dest = LabelEncoder()  
  
df['dest'] = le_dest.fit_transform(df['dest'])  
  
le_origin = LabelEncoder()  
  
df['origin'] = le_origin.fit_transform(df['origin'])
```

```

# Converting Pandas DataFrame into a Numpy array

X = df.iloc[:, 0:6].values # from column(years) to
column(distance)

y = df['delayed']


X_train, X_test, y_train, y_test = train_test_split(X,y,
test_size=0.25,random_state=61) # 75% training and 25% test


app = Flask(__name__)

model = pickle.load(open('model.pkl', 'rb'))


@app.route('/')

def home():

    return render_template('index.html')


@app.route('/predict',methods=['POST'])

def predict():

    year = request.form['year']

    month = request.form['month']

    day = request.form['day']

```

```
carrier = request.form['carrier']
```

```
origin = request.form['origin']
```

```
dest = request.form['dest']
```

```
year = int(year)
```

```
month = int(month)
```

```
day = int(day)
```

```
carrier = str(carrier)
```

```
origin = str(origin)
```

```
dest = str(dest)
```

```
if year >= 2013:
```

```
    x1 = [year,month,day]
```

```
    x2 = [carrier, origin, dest]
```

```
    x1.extend(x2)
```

```
    df1 = pd.DataFrame(data = [x1], columns = ['year',  
'month', 'date', 'carrier', 'origin', 'dest'])
```

```
df1['carrier'] = le_carrier.transform(df1['carrier'])
```

```
df1['origin'] = le_origin.transform(df1['origin'])
```

```
df1['dest'] = le_dest.transform(df1['dest'])
```



```
x = df1.iloc[:, :6].values  
ans = model.predict(x)  
output = ans  
  
return render_template('index.html',  
prediction_text=output)  
  
if __name__ == '__main__':  
    app.run(debug=False)  
  
# For mac, make 'app.run(debug=True)'
```

TEAM PROFILE

- **Team leader: YOGESHWARI.R**
- **Member 1: PAVITHRA.P**
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- **Member 3: NAVEEN.P**