# Final Project Info 7390 - Advanced Data Science and architecture



# **Predicting Airline Delays**

Presented by:

Dhanisha Damodar Phadate 001859234 phadate.d@husky.neu.edu

Palak Sharma
001828562
<a href="mailto:sharma.pala@husky.neu.edu">sharma.pala@husky.neu.edu</a>

Dharani Thirumalaisamy 001887922 thirumalaisamy.d@husky.neu.edu

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# PART 1: MODEL DEVELOPMENT

#### Abstract:

The project's aim is to predict the airline delay based on the historic datasets. In this part of the project, the data is taken from the cloud and pre-processing is carried out to generate the pickle files which when loaded will provide the desired output. The pickle files at the end is pushed to the Amazon's cloud.

#### a) Data Ingestion:

```
url ='https://storage.googleapis.com/team3/final_flight_delay.csv'
df = urllib.request.urlopen(url)

df = pd.read_csv(df)|
```

The dataset that is used in this project is of 2008's and 2007's. These two datasets and concatenated and pushed to the Google Cloud Platform. The url link provided is the public link of the dataset that is on Google Cloud Platform.

The dataset is ingested into the notebook from the cloud and the pre-processing techniques are carried out.

#### b) Data Cleaning:

When calculated for the number of missing values, there were a plenty of it.

So, in order to deal with it, the missing values are filled with a method called 'bfill' which will fill the missing values or the 'NaN' values with the previous records.

```
#def cleaning(data):
data = df.fillna(method='bfill',axis=0).fillna('0')
# return data
#cleaning(df)
```

#### c) Feature Selection:

In the dataset that we have there are string values for 'Unique Carrier', 'Dest' and 'Origin'. These values should be encoded before they are used in the prediction model.

There are different methods to do this . For example : Onehotencoding , Labelencoder , but the problem with these two is that they can't be converted back to string format . So , if we display the output in the numeric format , it is not possible for the user to understand that. So , to avoid that :

```
dest = data['Dest']
                                   origin = data['Origin']
uc = data.UniqueCarrier
                                                                        dest = dest.replace('ABE','1')
uc = uc.replace('9E','1')
                                   origin = origin.replace('ABE','1')
                                                                        dest = dest.replace('ABI','2')
uc = uc.replace('AA','2')
                                   origin = origin.replace('ABI','2')
                                                                        dest = dest.replace('ABQ','3')
uc = uc.replace('AQ','3')
                                   origin = origin.replace('ABQ','3')
                                                                        dest = dest.replace('ABY','4')
uc = uc.replace('AS','4')
                                   origin = origin.replace('ABY','4')
                                                                        dest = dest.replace('ACT','5')
uc = uc.replace('B6','5')
                                                                        dest = dest.replace('ACV','6')
                                   origin = origin.replace('ACT','5')
uc = uc.replace('CO','6')
                                                                        dest = dest.replace('ACY','7')
                                   origin = origin.replace('ACV','6')
uc = uc.replace('DL','7')
                                                                        dest = dest.replace('ADK','8')
                                   origin = origin.replace('ACY','7')
uc = uc.replace('EV','8')
                                                                        dest = dest.replace('ADQ','9')
uc = uc.replace('F9','9')
                                   origin = origin.replace('ADK','8')
                                                                        dest = dest.replace('AEX','10')
uc = uc.replace('FE','10')
                                   origin = origin.replace('ADQ','9')
                                                                        dest = dest.replace('AGS','11')
uc = uc.replace('HA','11')
                                                                        dest = dest.replace('AKN','12')
                                   origin = origin.replace('AEX','10')
uc = uc.replace('MQ','12')
                                                                        dest = dest.replace('ALB','13')
                                   origin = origin.replace('AGS','11')
uc = uc.replace('NW','13')
                                                                        dest = dest.replace('ALO','14')
                                   origin = origin.replace('AKN','12')
uc = uc.replace('OH','14')
                                                                        dest = dest.replace('AMA','15')
                                   origin = origin.replace('ALB','13')
uc = uc.replace('00','15')
                                                                        dest = dest.replace('ANC','16')
                                   origin = origin.replace('ALO','14')
uc = uc.replace('UA','16')
                                                                        dest = dest.replace('APF','17')
uc = uc.replace('US','17')
                                   origin = origin.replace('AMA','15')
                                                                        dest = dest.replace('ASE','18')
uc = uc.replace('WN','18')
                                                                        dest = dest.replace('ATL','19')
                                   origin = origin.replace('ANC','16')
                                                                        dest = dest.replace('ATW','20')
uc = uc.replace('XE','19')
                                   origin = origin.replace('APF','17')
                                                                        dest = dest.replace('AUS','21')
uc = uc.replace('YV','20')
                                   origin = origin.replace('ASE','18')
                                                                        dest = dest.replace('AVL','22')
uc = uc.replace('FL','21')
                                   origin = origin.replace('ATL','19')
                                                                        dest = dest.replace('AVP','23')
```

In this problem, it is not necessary to use any automated feature selection tool to select the features that affect the output because we know the answer our self.

Therefore in the final dataset that we are using for the model will be 'data' which has the following labels:

'Month', 'DayofMonth', 'DayOfWeek', 'uniquecarrier\_int', 'Cancelled', 'Diverted', 'CarrierDelay', 'WeatherDelay', 'NASDelay', 'SecurityDelay', 'LateAircraftDelay', 'FlightNum', 'ArrDelay', 'DepDelay', 'origin\_int', 'dest\_int'

The input that we will be taking from the user are:

- 1. Month which month he is travelling (1-12)
- 2. DayofMonth which date he is travelling (1-31)
- 3.Dayodweek which day it is in that particular week (1-7)
- 4. uniquecarrier\_int which flight he is taking
- 5. origin\_int origin airport
- 6. dest\_int destination airport

The first output that the user will be given is: If there is a delay or not? If 1 - not delayed; If 0 - delayed.

The second output which the user will be given is: By how much is the arrival delayed and departure delayed?

The third output for the user will be: What type of delay it is? There are 5 types of delay: Carrier delay, weather delay, NAS delay, security delay, Late aircraft delay

#### d)Models Used:

i) For the first output given to the user, the model used is **Naive Bayes classifier** which gives a accuracy of 0.97. Irrespective of the type of algorithm we use, the accuracy is very high. So any algorithm can be used for this prediction.

```
from sklearn.naive_bayes import GaussianNB
gnb = GaussianNB()

# Train our classifier
gnb.fit(df_inputs_train, df_target_train)

pickle.dump(gnb,open('delay.pkl', "wb" ))
#from sklearn.ensemble import RandomForestClassifier
```

```
g = gnb.predict(df_inputs_test)
from sklearn.metrics import accuracy_score
r = accuracy_score(df_target_test,g)
r
0.97853725205181741
```

accuracy\_score simply implies

since it is a single label output, the confusion matrix doesn't make much sense in this case.

ii) For the second output, the model used is Random Forest Regressor.

```
rfr = RandomForestRegressor(n_estimators =10, random_state = 1,n_jobs=-1)
rfr.fit(inputs_train,target_train)
#pickle.dump(rfr,open('delay_value.pkl', "wb" ))
```

```
r = rfr.predict(inputs_test)
rl = r2_score(target_test , r)
rl |
0.0021386247940772818
```

For logistic regression:

```
lr = LinearRegression()
lr.fit(inputs_train,target_train)
LinearRegression(copy_X=True, fit_in)

l = lr.predict(inputs_test)
l = r2_score(target_test , l)
l
-0.066348491988103442
```

R2\_score value simply implies how well the dataset fits the regression line. Any variability in the dataset will be seen in the output. The greater the r2\_score, better the model.

iii) For the third output, the model used is **Random Forest Regressor.**The reason behind the usage of this algorithm is, this is a multi-label regression problem, so other algorithms can't be used since they don't support multi-class regression problem.

```
x_train , x_test , y_train , y_test = train_test_split(df_inputs, df_target_reg_2, test_size=0.30,random_state=1)
rfr_2 = RandomForestRegressor(n_estimators =5, random_state = 1,n_jobs=-1)
rfr_2.fit(x_train,y_train)
pickle.dump(rfr_2,open('delay_type_value.pkl', "wb" ))
```

The pickle files of each model is created and sent to cloud.

#### e) Cloud upload:

By dockerizing the whole code, access\_key, secret\_key can be given which will push the files to the Amazon cloud.

#### f) Dockerization:

- 1. TO BUILD AND RUN THE DOCKER IMAGE LOCALLY vim dockerfile
- 2. TO BUILD THE IMAGE docker build -f dockerfile -t finalproject
- 3. TO RUN THE IMAGE TO CREATE CONTAINER docker run -e Access\_key=Access\_key -e Secret\_key=Secret\_key -ti finalproject
- 4. TO TAG THE IMAGE docker tag <image id> dhanisha/finalproject
- 5. TO PUSH THE DOCKER IMAGE TO DOCKER HUB docker push dhanisha/finalproject
- 6. TO PULL THE DOCKER IMAGE FROM THE DOCKER HUB docker pull dhanisha/finalproject
- 7. TO RUN THE IMAGE PULLED FROM DOCKER HUB docker run -e Access\_key=Access\_key -e Secret\_key=Secret\_key -ti finalproject

# PART 2: WeB App Deployment

#### Abstract:

The project's aim is to predict the airline delay based on the historic datasets. In this part of the project, all the prediction values obtained from the pipeline are used as a backend code for our web app. The web App is created in python using Flask and jinja templates. Features like css, REST APIs, javascript and Bootstrap are used for the development of a fully deployed App. The App is deployed on EC2 instance which allows users to get the predicted value of whether the Flight is delay and which type of delay it is. And also the average arrival and departure time.

#### A) FIRING the EC2 instance:

The following things are required:

- 1) Aws login account.
- 2) Putty Key Generator (to use the pem and ppk files)
- 3) Putty shell.
- 4) The following commands to make the environment ready for the App to be deployed:
  - Changing the version of python to python3 as in ubuntu the default version is python 2.7 using the below commands:

GO TO THE ROOT and check the version using: python –version Execute nano ~/ .bashrc and add the lines:

alias python=python3 alias python='/usr/bin/python3'

Execute source ~/.bashrc and check the new version of python

- Installing all the required packages: pandas, skipy, numpy, sklearn, boto3, sqlalchemy, bokeh, flask, pip3.
- sudo apt-get install apache2 mysql-client mysql-server
- sudo apt-get install libapache2-mod-wsgi
- sudo a2enmod wsgi
- cd /var/www
- sudo mkdir FlaskApp
- cd FlaskApp

- sudo mkdir FlaskApp
- cd FlaskApp
- sudo mkdir static
- sudo mkdir templates
- sudo nano \_\_init\_\_.py
- 5) Three main Files are created and added for the complete configuration
  - a) nano /var/www/FlaskApp/FlaskApp/\_\_init\_\_.py
  - b) sudo nano /etc/apache2/sites-available/FlaskApp.conf
  - c) sudo nano /var/www/FlaskApp/flaskapp.wsgi
- 6) Giving the following permissions:

sudo chown -R ubuntu:ubuntu /var/www sudo chown -R ubuntu:ubuntu /var/www/FlaskApp/FlaskApp sudo chown -R ubuntu:ubuntu/var/www/FlaskApp/FlaskApp/static

sudo chown -R ubuntu:ubuntu /var/www/FlaskApp/FlaskApp/templates

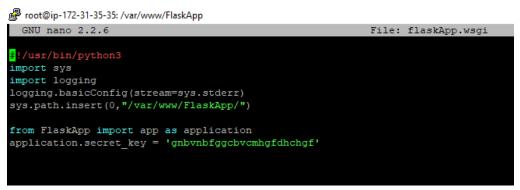
sudo chmod a+rwx /var/www/FlaskApp

sudo chmod a+rwx /var/www/FlaskApp/FlaskApp sudo chmod a+rwx /var/www/FlaskApp/FlaskApp/static

#### 7) CONFig File:

```
뤔 root@ip-172-31-35-35: /var/www
 GNU nano 2.2.6
                                                 File: /etc/apache2/sites-available/
VirtualHost *:80>
                ServerName ec2-54-191-117-224.us-west-2.compute.amazonaws.com
                ServerAdmin youremail@email.com
                WSGIScriptAlias / /var/www/FlaskApp/flaskApp.wsgi
                <Directory /var/www/FlaskApp/FlaskApp/>
                        Order allow, deny
                        Allow from all
                </Directory>
                Alias /static /var/www/FlaskApp/FlaskApp/static
                <Directory /var/www/FlaskApp/FlaskApp/static/>
                        Order allow, deny
                        Allow from all
                </Directory>
                ErrorLog ${APACHE LOG DIR}/error.log
                LogLevel warn
                CustomLog ${APACHE_LOG_DIR}/access.log combined
/VirtualHost>
```

8) flaskapp.wsgi file

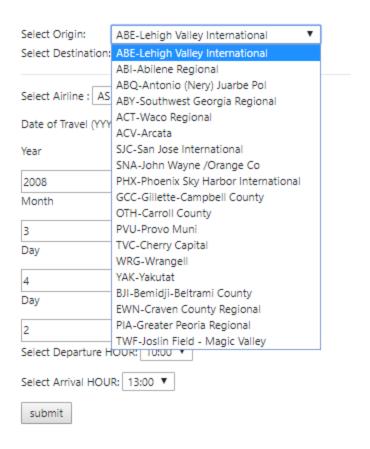


B) Deploying a Flask App: An App is up and running on : ec2-54-191-117-224.us-west-2.compute.amazonaws.com

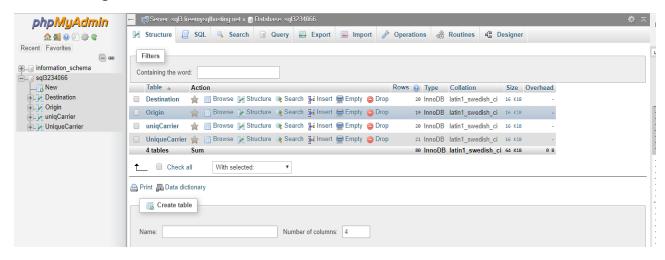


#### C) Using REST APIs:

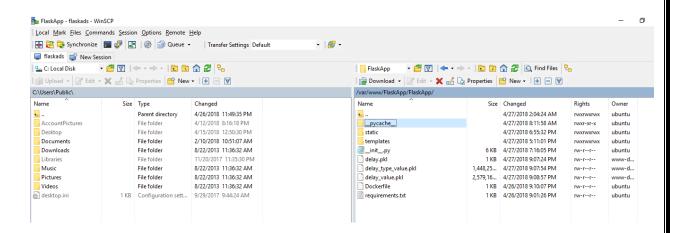
We have made the dynamic web pages which will auto fill the dropdown of the form using data from third party hosted SQL data base. Rest API calls will be sent to the server and a page will be rendered dynamically.



#### D) Creating an SQL database:



#### E) File Folder using WinSCP

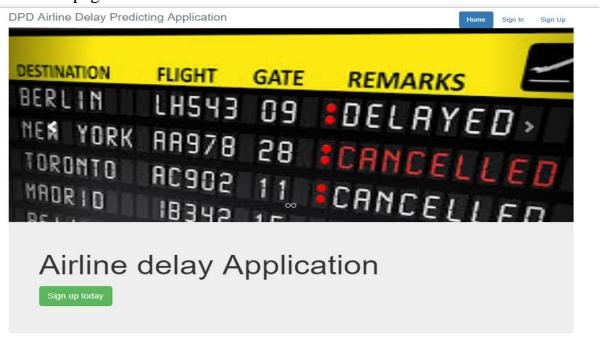


#### **File Structure:**

/var/www/FlaskApp/FlaskApp/\_\_init\_\_.py /var/www/FlaskApp/flaskApp.wsgi /var/www/FlaskApp/FlaskApp/static /var/www/FlaskApp/FlaskApp/templates

#### F) Fully Functional App Deployed

Domain name: http://ec2-54-191-117-224.us-west-2.compute.amazonaws.com Home page:



# Login page: DPD Airline Delay Predicting Application Sign In Username Password Login Login

## Register page:

DPD Airline Delay Predicting Application

# Registration Form



# Form page:

# PDP Aireline Delay Predicting Application

Select Origin:	ABE-Lehigh Valley International ▼
Select Destination:	SUX-Sioux Gateway ▼
Select Airline :	9E-Pinnacle Airlines Inc. ▼
Date of Travel (YYYY-MM-DD):	
Year	
Month	
Day	
Day of Week	
Select Departure HOUR:	10:00 ▼
Select Arrival HOUR:	13:00 ▼
submit	
Company 2018	

Predict page:



## G) Dockerization:

- 1. TO BUILD AND RUN THE DOCKER IMAGE LOCALLY vim dockerfile
- 2. TO BUILD THE IMAGE docker build -f dockerfile -t flask-image:latest
- 3. TO RUN THE IMAGE TO CREATE CONTAINER docker run -ti flask-image
- 4. TO TAG THE IMAGE docker tag <image id> dhanisha/finalprojectflask
- 5. TO PUSH THE DOCKER IMAGE TO DOCKER HUB docker push dhanisha/finalprojectflask
- 6. TO PULL THE DOCKER IMAGE FROM THE DOCKER HUB docker pull dhanisha/finalprojectflask
- 7. TO RUN THE IMAGE PULLED FROM DOCKER HUB

#### docker run -ti flask-image

Conclusion: We could successfully deploy an Airline delay prediction application for the user on AWS ec2 instances.

#### References:

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