

Detail Project Report Flight Fare Prediction

Revision Number – 1.2

Last Date of Revision: 30 – 09 -2022

Anishv Rawal

Kunal Bhandawat

Document Version Control

| Date | Version | Description | Author |
|----------------|---------|--|--------|
| 23 – 09 - 2022 | 1.0 | Abstract Introduction General Description | Anishv |
| 26 – 09 - 2022 | 1.1 | Technical Requirements Data Requirements Data Preprocessing Design Flow | Kunal |
| 29 – 09 - 2022 | 1.2 | Data from User and its validation Rendering the Results Deployment Conclusion | Anishv |

Contents

| | |
|--------------------------------|---|
| Document Version Control | 2 |
| Abstract | 4 |
| 1. Introduction | 5 |

| | |
|---|----|
| 1.1 Why this DPR Document ? | 5 |
| 2. General Description | 5 |
| 2.1 Problem Perspective | 5 |
| 2.2 Problem Statement | 5 |
| 2.3 Proposed Solution | 5 |
| 2.4 Further Improvements | 6 |
| 3. Technical Requirements | 6 |
| 3.1 Tools Used | 6 |
| 4. Data Requirements | 6 |
| 4.1 Data Collection | 6 |
| 4.2 Data Description | 7 |
| 4.3 Importing Data into Database | 7 |
| 4.4 Exporting Data from Database | 7 |
| 5. Data Preprocessing | 8 |
| 6. Design Flow | 8 |
| 3.1 Model Creation and Evaluation | 8 |
| 3.2 UI Integration | 8 |
| 3.2 Deployment Process | 9 |
| 3.3 Logging | 9 |
| 7. Data from User | 9 |
| 8. Data Validation | 9 |
| 9. Rendering the Results | 9 |
| 10. Deployment | 9 |
| 11. Conclusion | 10 |
| 12. FAQs | 10 |

Abstract

The recent changes in the international market had a large impact on the Aviation sector because of several reasons. These impact the two class folks, the first is Business perspective and second is Customer perspective. The major reason for such an impact is the governments around the world amended totally different rules to their various Airline

firms. Taking these factors into consideration, the value of the flight tickets has varied from one place to another. Booking a flight ticket has its price tag split into two, one is online bookings and other is offline bookings. Each of these have their various criteria for value of the price, one such example is the server load and therefore the range of booking requests. During this machine learning implementation, we are going to see numerous factors that impact the price of the flight ticket and predict the acceptable price of the ticket.

1. Introduction

1.1 Why this DPR Document ?

The main purpose of this DPR documentation is to add the necessary details of the project and provide the description of the machine learning model and the written

code. This also provides the detailed description on how the entire project has been designed end-to-end.

Key points :

- Describes the design flow
- Implementations
- Software requirements □ Architecture of the project □ Non-functional attributes like:
 - Reusability
 - Portability
 - Resource utilization

2. General Description

2.1 Problem Perspective

The flight fare prediction may be a machine learning model that helps users to predict the price of the flight tickets and help them to understand the price of their journey.

2.2 Problem Statement

After amendment of the new rules, there is changes in the flight fare price from one location to another. The main goal of the system is to create a model to predict the price of their flight fare on the basis of bound input provided by user like date of journey, Source, Destination and many more.

2.3 Proposed Solution

To solve the problem, we have created a User interface for taking the input from the user to predict the flight fare price using our trained ML model after processing the input and at last the output (predicted value) from the model is communicated to the User.

2.4 Further Improvements

We also analyze the data used for training the ML model by considering different occasions such as Weekday, Season or any Social reasons, considering different

angles of business. If we use such information and predict the discounted flight fare price, it will bring some loss to the airline companies but users can benefit from that. If we develop these using the Business perspective of Airline, this technique isn't thought - about.

3. Technical Requirements

As technical requirements, we don't need any specialized hardware for virtualization of the application. The user should have the device that has the access to the web and the fundamental understanding of providing the input.

3.1 Tools Used

- Python 3.9 is employed because of the programming language and frameworks like NumPy, Pandas, Scikit - learn and alternative modules for building the model.
- Jupyter - Notebook is employed as an IDE.
- For Data visualizations, seaborn and components of matplotlib are getting used.
- For information assortment prophetess info is getting used.
- Front end development is completed victimization HTML/CSS.
- Flask is employed for each information and backend readying.
- GitHub is employed for version management.
- Heroku is employed for deployment.

4. Data Requirements

The Data requirements totally supported the matter statement and also the dataset is accessible on the Kaggle within the file format of (.xlsx).

4.1 Data Collection

The data for these project is collected from the Kaggle Dataset, the URL for the dataset is [kaggle.com/datasets/nikhilmittal/flight-fare-prediction-mh](https://www.kaggle.com/datasets/nikhilmittal/flight-fare-prediction-mh)

4.2 Data Description

Flight Fare Prediction is 10K+ dataset publicly available on the Kaggle. The information in the dataset is present in two separated excel files named as train.xlsx and test.xlsx. Dataset contains 10683 rows which shows the information such Date of Journey, Source, Destination, Arrival Time, Departure Time, Total stops, Airlines, Additional Info and Price. The glance of the Dataset is :

| Airline | Date_of_Journey | Source | Destination | Route | Dep_Time | Arrival_Time | Duration | Total_Stops | Additional_Info | Price |
|-------------------|-----------------|----------|-------------|-----------------------|----------|--------------|----------|-------------|-----------------------------|-------|
| IndiGo | 24/03/2019 | Banglore | New Delhi | BLR → DEL | 22:20 | 01:10 22 Mar | 2h 50m | non-stop | No info | 3897 |
| Air India | 1/05/2019 | Kolkata | Banglore | CCU → IXR → BBI → BLR | 05:50 | 13:15 | 7h 25m | 2 stops | No info | 7662 |
| Jet Airways | 9/06/2019 | Delhi | Cochin | DEL → LKO → BOM → COK | 09:25 | 04:25 10 Jun | 19h | 2 stops | No info | 13882 |
| IndiGo | 12/05/2019 | Kolkata | Banglore | CCU → NAG → BLR | 18:05 | 23:30 | 5h 25m | 1 stop | No info | 6218 |
| IndiGo | 01/03/2019 | Banglore | New Delhi | BLR → NAG → DEL | 16:50 | 21:35 | 4h 45m | 1 stop | No info | 13302 |
| SpiceJet | 24/06/2019 | Kolkata | Banglore | CCU → BLR | 09:00 | 11:25 | 2h 25m | non-stop | No info | 3873 |
| Jet Airways | 12/03/2019 | Banglore | New Delhi | BLR → BOM → DEL | 18:55 | 10:25 13 Mar | 15h 30m | 1 stop | In-flight meal not included | 11087 |
| Jet Airways | 01/03/2019 | Banglore | New Delhi | BLR → BOM → DEL | 08:00 | 05:05 02 Mar | 21h 5m | 1 stop | No info | 22270 |
| Jet Airways | 12/03/2019 | Banglore | New Delhi | BLR → BOM → DEL | 08:55 | 10:25 13 Mar | 25h 30m | 1 stop | In-flight meal not included | 11087 |
| Multiple carriers | 27/05/2019 | Delhi | Cochin | DEL → BOM → COK | 11:25 | 19:15 | 7h 50m | 1 stop | No info | 8625 |
| Air India | 1/06/2019 | Delhi | Cochin | DEL → BLR → COK | 09:45 | 23:00 | 13h 15m | 1 stop | No info | 8907 |
| IndiGo | 18/04/2019 | Kolkata | Banglore | CCU → BLR | 20:20 | 22:55 | 2h 35m | non-stop | No info | 4174 |
| Air India | 24/06/2019 | Chennai | Kolkata | MAA → CCU | 11:40 | 13:55 | 2h 15m | non-stop | No info | 4667 |
| Jet Airways | 9/05/2019 | Kolkata | Banglore | CCU → BOM → BLR | 21:10 | 09:20 10 May | 12h 10m | 1 stop | In-flight meal not included | 9663 |

4.3 Importing data into Database

Created associate API for the transfer of the info into the Cassandra info, steps performed are :

- Connection is created with the info.
- Created a info with name FlightInfo.
- cqlsh command is written for making the info table with needed parameters.
- And finally, a cqlsh command is written for uploading the Knowledge Set into data table by bulk insertion.

4.4 Exporting Data from Database

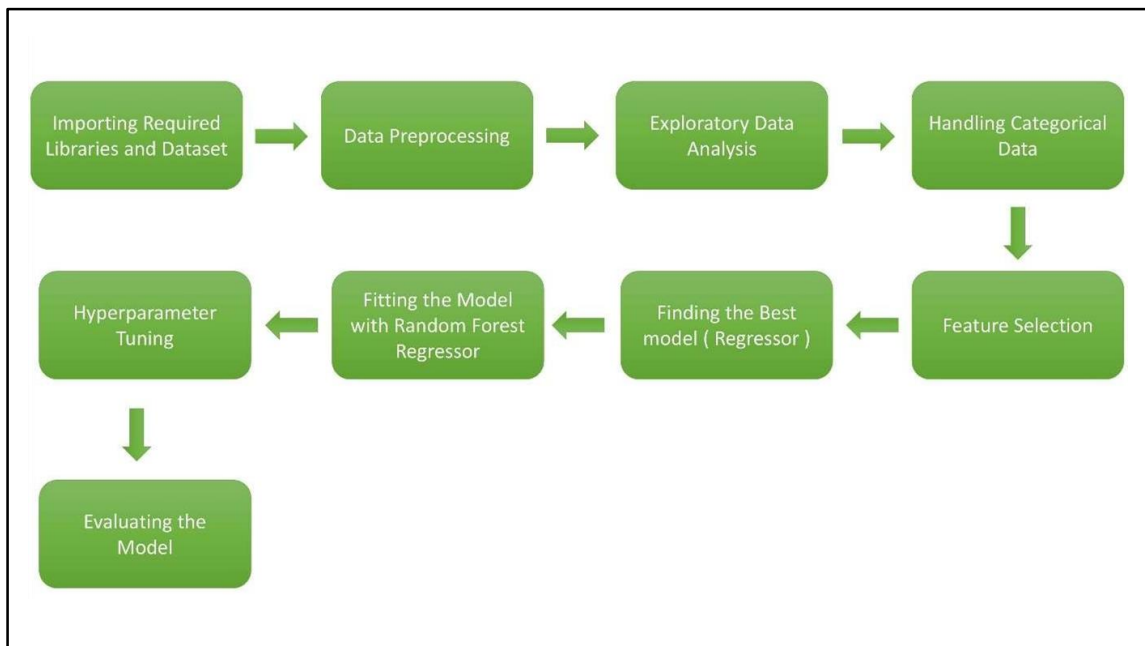
In the above created API, the download URL is also being created, which downloads the data into a csv file format.

5. Data Preprocessing

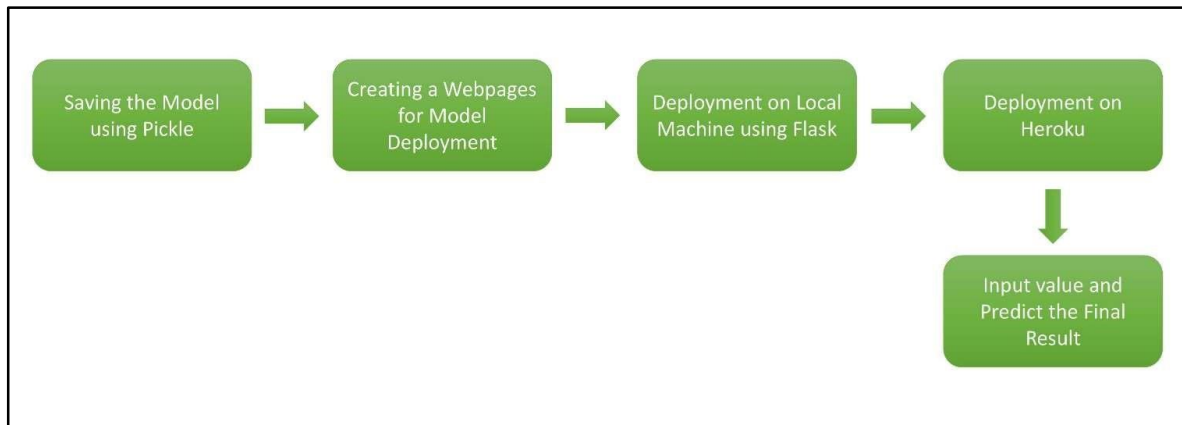
- Checked for info of the Dataset, to verify the correct datatype of the Columns.
- Checked for Null values, because the null values can affect the accuracy of the model.
- Converted all the desired columns into Datetime format.
- Performed One – Hot encoding on the desired columns.
- Checking the distribution of the columns to interpret its importance.
- Now, the info is prepared to train a Machine Learning Model.

6. Design Flow

6.1 Modelling Creation and Evaluation



6.3 Deployment Process



6.4 Logging

In logging, at each time an error or an exception occurs, the event is logged into the system log file with reason and timestamp. This helps the developer to debug the system bugs and rectify the error.

7. Data from User

The data from the user is retrieved from the created HTML web page.

8. Data Validation

The data provided by the user is then being processed by app.py file and validated. The validated data is then sent to the prepared model for the prediction.

9. Rendering the Results

The data sent for the prediction is then rendered to the web page.

10. Deployment

The tested model is then deployed to Heroku. So, users can access the project from any internet device.

11. Conclusion

The Flight Fare Prediction system will predict the price for helping the customers with the trained knowledge with a set of rules. The user can use this system to recognize the approximate value of its flight fare for his or her journey.

12. Frequently Asked Questions (FAQs)

Q1) What's the source of data ?

The data for training is provided by the client in multiple batches and each batch contain multiple files.

Q2) What was the type of data ?

The data was the combination of numerical and Categorical values.

Q3) What's the complete flow you followed in this Project ?

Refer Page no 6 for better Understanding.

Q4) After the File validation what you do with incompatible file or files which didn't pass the validation ?

Files like these are moved to the Achieve Folder and a list of these files has been shared with the client and we removed the bad data folder.

Q5) How logs are managed ?

We are using different logs as per the steps that we follow in validation and modeling like File validation log, Data Insertion, Model Training log, prediction log etc.

Q6) What techniques were you using for data pre-processing ?

- Removing unwanted attributes.
- Visualizing relation of independent variables with each other and output variables.
- Checking and changing Distribution of continuous values.

- Removing outliers
- Cleaning data and imputing if null values are present.
- Converting categorical data into numeric values.

Q7) How training was done or what models were used ?

- Before dividing the data in training and validation set, we performed pre-processing over the data set and made the final dataset.
- As per the dataset training and validation data were divided.
- Algorithms like Linear regression, SVM, Decision Tree, Random Forest, XGBoost were used based on the recall, final model was used on the dataset and we saved that model.

Q8) How Prediction was done ?

The testing files are shared by the client. We Performed the same life cycle on the provided dataset. Then, on the basis of dataset, model is loaded and prediction is performed. In the end we get the accumulated data of predictions.

Q9) What are the different stages of deployment?

- First, the scripts are stored on GitHub as a storage interface.
- The model is first tested in the local environment.
- After successful testing, it is deployed on Heroku.