

NAME OF THE PROJECT

Flight Price Prediction

Submitted by:

Ankita Srivastava

ACKNOWLEDGMENT

We have scraped the data for over 1500 cars using Selenium script from 4 different websites from different locations around the country. The websites are as followed:

- 1...yatra.com,
- 2 .. skyscanner.com
3. Makemytrip.com
4. goibibo.com

I use this link to make my dataset. Makemytrip.com

<https://www.makemytrip.com/flights>

I use sklearn to find data coding description

INTRODUCTION

1))Business Problem Framing

Flight ticket prices can be something hard to guess. we have been provided with prices of flight tickets for various airlines between the months of March and June ,and between various cities, using which we aim to build a model which predicts the prices of the flights using various input features.

2))Conceptual Background of the Domain Problem

1. Time of purchase patterns (making sure last-minute purchases are expensive)
2. Keeping the flight as full as they want it (raising prices on a flight which is filling up in order to reduce sales and hold back inventory for those expensive last-minute expensive purchases)

- Review of Literature

The first thing that we can do when tackling a data science problem is getting an understanding of the dataset that you are working with. Key observations and trends in the data were noted down. All correlations within the variables and the output '*price*' were monitored. For this you can use `df.info`,`df.head()`etc..

Analytical Problem Framing

- **Mathematical/ Analytical Modeling of the Problem**

Here mathematical function use calculating the data loss. Data describe the statistical function. Some regression model use to build the regression model.

- **Data Sources and their formats**

We use some flight booking website to build the dataset. like I use makemytrip.com to make my project. I used only flight booking detail. firstly I scrap all data using BeautifulSoup method and after that I make a prediction model. where price is target.

```
In [7]: page = requests.get(url2)
        print("Legality Response number from makemytrip URL is:", page) # to show the response output from the webpage
        soup = BeautifulSoup(page.content)

        Legality Response number from makemytrip URL is: <Response [200]>

In [8]: # first connect to the browser
        driver = webdriver.Chrome(r"C:\Users\DELL\Downloads\chromedriver_win32.zip\chromedriver.exe" )

In [9]: url2 = "https://www.makemytrip.com/flights/"
        driver.get(url2)

In [11]: #Clicking on mumbai to delhi link in myflighttrip
        location_click = driver.find_element(by='xpath', value="//html/body/div[1]/div/div[2]/div/main/div[7]/div/div[5]/div/p[2]/p/a[1]/span")
        location_click.click()

In [12]: # Airlines name
        Airlines_name = []
        try :
            flight = driver.find_elements(by='xpath', value="//span[@class='boldFont blackText airlineName']")
            for i in flight:
                Airlines_name.append(i.text)
        except NoSuchElementException:
            Airlines_name.append('No details available')
        except StaleElementReferenceException:
            Airlines_name.append('No details available')
```

Scrap the data

- **Data Preprocessing Done**

1. firstly I check the null values in this data, no any null values are present.
2. how many unique values are present?
3. any white spaces are present or not?
4. which one is categorical data and which is continuous?
5. which type of this problem regression or classification? When

we check all these thing I clean all data and fill appropriate.
After that data visualization part proceed.

- **Data Inputs- Logic- Output Relationships**

Data model preparation depend on the large data . it has 228 rows and 7 column. every column affect the o/p. mostly depend on the duration, means how much it take time to one place from another.

- **State the set of assumptions (if any) related to the problem under consideration**

I assume that target depend on flight name but I was wrong..it's totally depend on Duration.

- **Hardware and Software Requirements and Tools Used**

Data modelling need 8 Gb RAM . I can make model on Google colab or jupyter notebook. Here I use 2 type libraries one for scaraping the data and other for data modelling.

```
In [5]: import warnings
warnings.simplefilter("ignore")
warnings.filterwarnings("ignore")

import time, sys
import tqdm.notebook as tqdm
import pandas as pd
import numpy as np

import requests
import selenium
from bs4 import BeautifulSoup
from selenium import webdriver
from urllib.parse import urljoin
from selenium.webdriver.common.by import By
from selenium.webdriver.support.ui import WebDriverWait
from selenium.webdriver.support import expected_conditions as ec
from selenium.common.exceptions import TimeoutException, StaleElementReferenceException
from selenium.common.exceptions import NoSuchElementException, ElementClickInterceptedException
```

This is for scaraping the data

```

import numpy as np

#plotting Libraries
import matplotlib.pyplot as plt
import seaborn as sns
import joblib

#feature engineering
from sklearn.preprocessing import StandardScaler, LabelEncoder

#train test split and #cross validation
from sklearn.model_selection import train_test_split, GridSearchCV, cross_val_score
from scipy.stats import zscore

#metrics
from sklearn.metrics import confusion_matrix, classification_report
from sklearn.metrics import accuracy_score
from sklearn.metrics import r2_score, mean_squared_error

#ML models
import sklearn
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.linear_model import SGDClassifier
from sklearn.linear_model import LinearRegression, Ridge, Lasso
from sklearn.ensemble import RandomForestRegressor
from sklearn.svm import SVR
from sklearn.tree import DecisionTreeRegressor
from sklearn.neighbors import KNeighborsRegressor
from sklearn.ensemble import AdaBoostRegressor
from sklearn.ensemble import ExtraTreesRegressor
from sklearn.ensemble import GradientBoostingRegressor

```

This is for model building

Model/s Development and Evaluation

- Identification of possible problem-solving approaches (methods)

```
In [13]: # Departure time is when a plane leaves the gate.  
  
# Extracting Hours  
df["Dep_hour"] = pd.to_datetime(df["Departure"]).dt.hour  
  
# Extracting Minutes  
df["Dep_min"] = pd.to_datetime(df["Departure"]).dt.minute  
  
# Now we can drop Dep_Time as it is of no use  
df.drop(["Departure"], axis = 1, inplace = True)
```

```
In [14]: df.head(3)
```

```
Out[14]:
```

	Unnamed: 0	Airlines name	Stops	Duration	Arrival	Price	Dep_hour	Dep_min
0	0	SpiceJet	1 stop via Jaipur	04 h 20 m	10:30	3680.0	6	10
1	1	AirAsia	1 stop via Jaipur	07 h 50 m	23:05	4546.0	15	15
2	2	AirAsia	1 stop via Bengaluru	05 h 45 m	01:25	4546.0	19	40

```
In [15]: # Arrival time is when the plane pulls up to the gate.  
  
# Extracting Hours  
df["Arrival_hour"] = pd.to_datetime(df.Arrival).dt.hour  
  
# Extracting Minutes  
df["Arrival_min"] = pd.to_datetime(df.Arrival).dt.minute  
  
# Now we can drop Arrival Time as it is of no use
```

- Run and Evaluate selected models

In [85]: #Randomized Search CV

```
# Number of trees in random forest
n_estimators = [int(x) for x in np.linspace(start = 100, stop = 1200, num = 12)]
# Number of features to consider at every split
max_features = ['auto', 'sqrt']
# Maximum number of levels in tree
max_depth = [int(x) for x in np.linspace(5, 30, num = 6)]
# Minimum number of samples required to split a node
min_samples_split = [2, 5, 10, 15, 100]
# Minimum number of samples required at each leaf node
min_samples_leaf = [1, 2, 5, 10]
```

In [86]: # Create the random grid

```
random_grid = {'n_estimators': n_estimators,
               'max_features': max_features,
               'max_depth': max_depth,
               'min_samples_split': min_samples_split,
               'min_samples_leaf': min_samples_leaf}
```

In [87]: # Random search of parameters, using 5 fold cross validation,
search across 100 different combinations

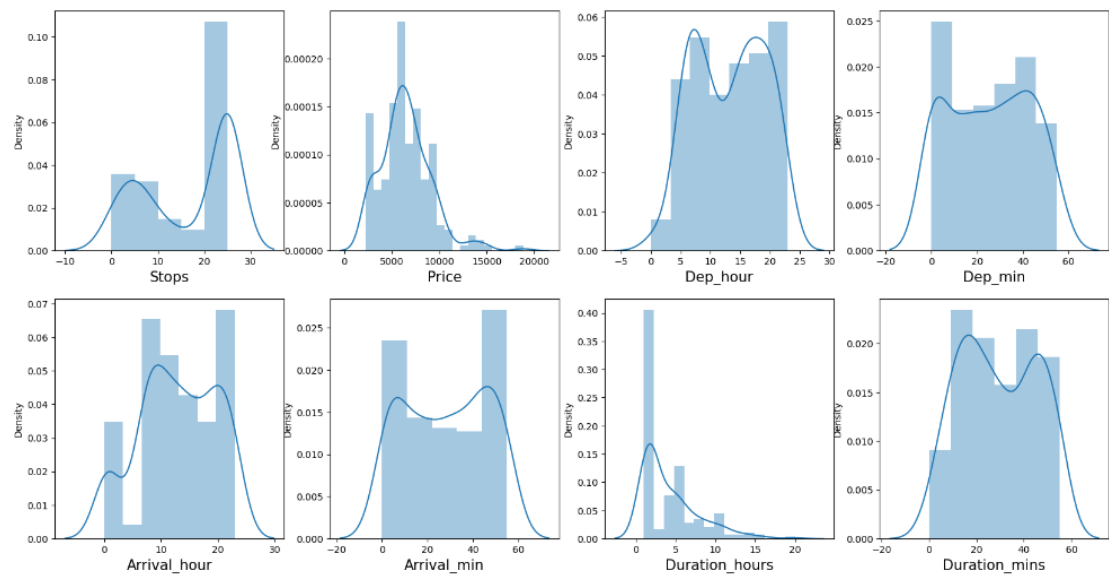
```
rf_random = RandomizedSearchCV(estimator = reg_rf, param_distributions = random_grid, scoring='neg_mean_squared_error', n_iter = 1
```

In [88]: rf_random.fit(X_train,Y_train)

Fitting 5 folds for each of 10 candidates, totalling 50 fits

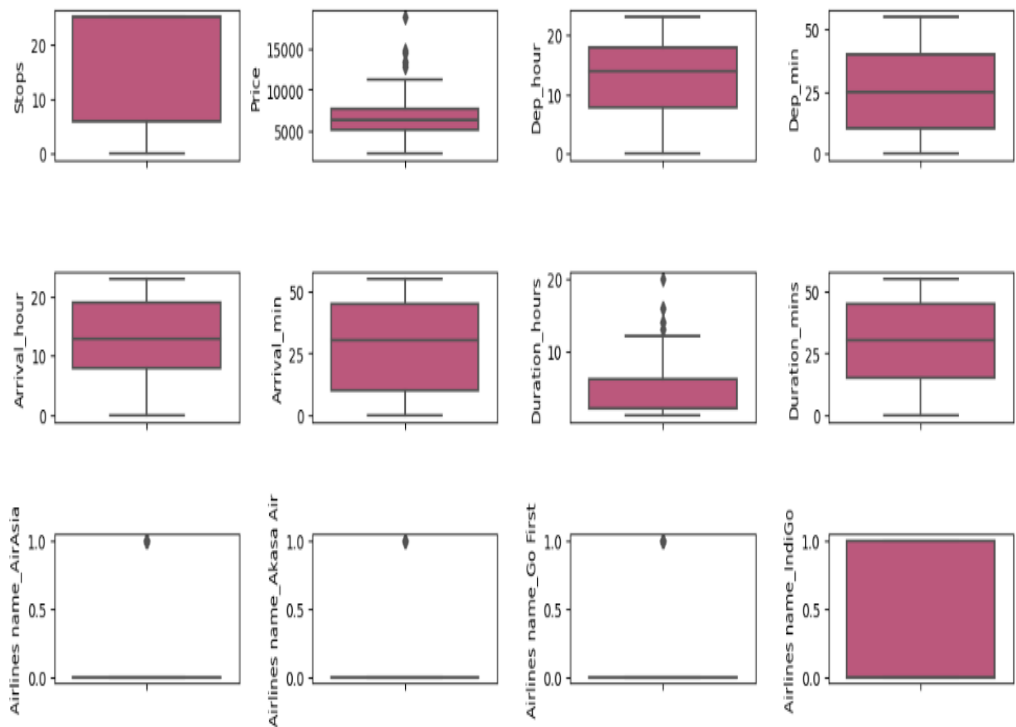
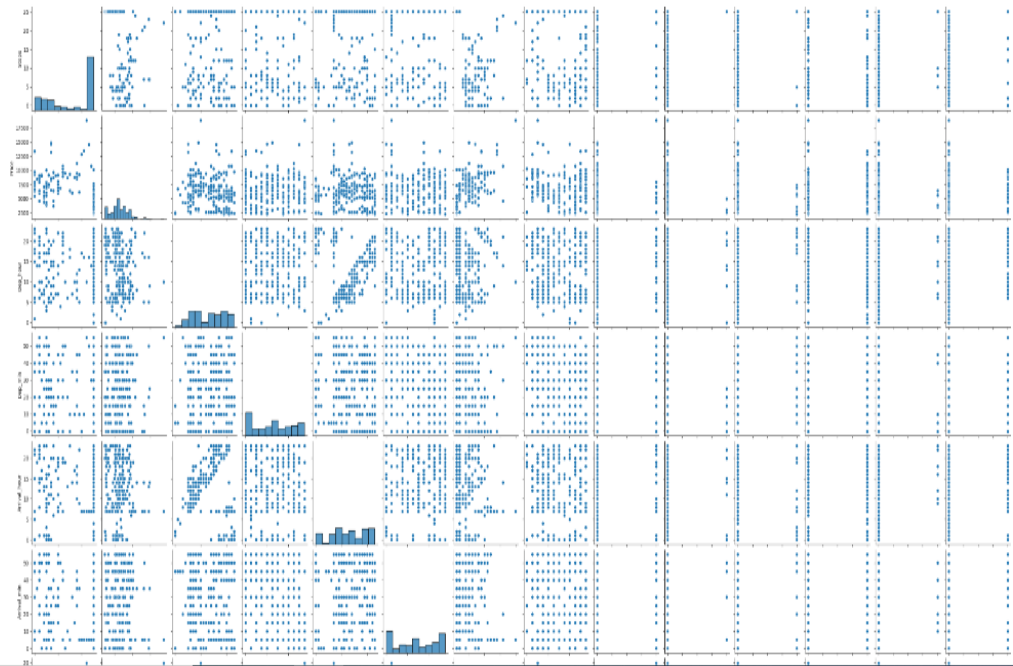
[CV] END max depth=10, max features=sqrt, min samples leaf=5, min samples split=5, n estimators=900; total time= 1.2s

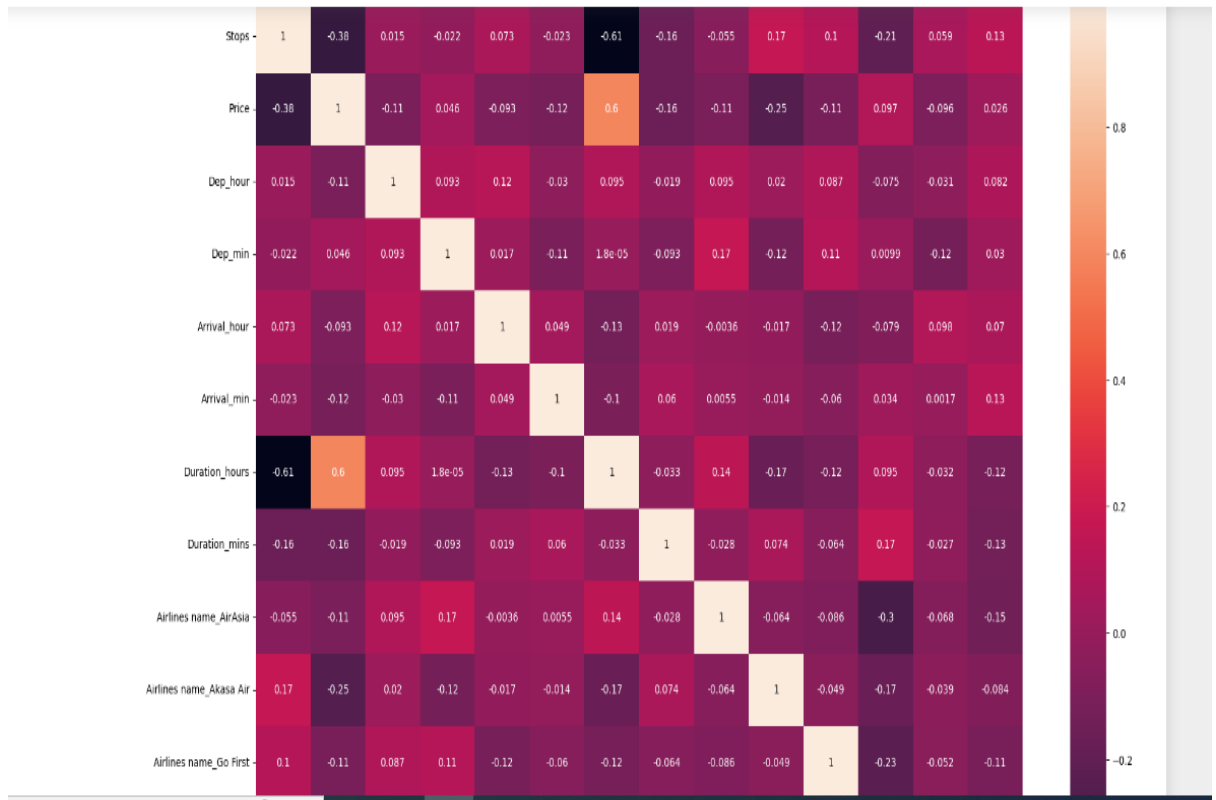
• Visualizations




```
In [61]: plt.figure(figsize=(15,10))
sns.pairplot(data)
plt.show()
```

<Figure size 1500x1000 with 0 Axes>





- Interpretation of the Results

I can see after data preprocessing step only duration hour column highly correlated with target column price. here minor outliers are present in continuous and categorical column. but i did not remove outliers from numerical column. but I have less data so I did not accept data after removing outliers because it can not build my train and test data..

CONCLUSION

- **Key Findings and Conclusions of the Study**

Here, we can see that all the predicted prices are either equal or nearly equal to the other airlines prices. Hence we conclude that our model 'price' is working very well. And we shall save it for further use..

- **Learning Outcomes of the Study in respect of Data Science**

- 1..use dist plot to show data distribution is normal.
- 2...use line plot to show features are linear with target.
- 3..using Heat map to show correlation b/w target and features.

Random forest regressor is good perform other than all algorithm.but it also not showing satisfied result. First challenge was to scrap all data from different website. 2 nd was to create dataframe and convert it to string to float

- **Limitations of this work and Scope for Future Work**

As a part of future work, we aim at the variable choices over the algorithms that were used in the project. We could only explore two algorithms whereas many other algorithms exist and might be more accurate. More specifications will be added to a system or provide more accuracy in terms of price in the system