Article

Flight Fare Prediction

Problem Statement

Flight ticket prices can be something hard to guess, today we might see a price, check out the price of the same flight tomorrow, it will be a different story. We might have often heard travelers saying that flight ticket prices are so unpredictable. Here you will be provided with prices of flight tickets for various airlines between the months of March and June of 2019 and between various cities.

Problem Definition

In this article, we are going to analyze and predict flight fare prediction using Machine Learning, and exploratory data analysis techniques. with the help of features, we are going to predict the flight price. So basically there are two datasets available

Train dataset Test dataset

About the Dataset

The dataset contains several features based on that features we are going to predict the price.

FEATURES:

Airline: The name of the airline.

Date_of_Journey: The date of the journey

Source: The source from which the service begins.

Destination: The destination where the service ends.

Route: The route was taken by the flight to reach the destination.

Dep_Time: The time when the journey starts from the source.

Arrival_Time: Time of arrival at the destination.

Duration: Total duration of the flight.

Total_Stops: Total stops between the source and destination.

Additional_Info: Additional information about the flight

Price: The price of the ticket

Importing libraries

import pandas as pd import numpy as np import matplotlib.pyplot as plt

import pickle

from pandas.plotting import scatter matrix

%matplotlib inline

import warnings

warnings.filterwarnings('ignore')

import seaborn as sns

from sklearn.metrics import r2 score

from sklearn import metrics

from sklearn.linear model import LinearRegression

from sklearn.model_selection import train_test_split

from sklearn.tree import DecisionTreeClassifier

from sklearn.preprocessing import StandardScaler

from sklearn.neighbors import KNeighborsClassifier

from sklearn.metrics import accuracy_score, confusion_matrix, classification_report

from sklearn.ensemble import RandomForestClassifier

DATA

Train Data

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	IndiGo	24/03/2019	Banglore	New Delhi	$BLR \to DEL$	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	Air India	1/05/2019	Kolkata	Banglore	$\begin{array}{c} CCU \to IXR \to BBI \to \\ & BLR \end{array}$	05:50	13:15	7h 25m	2 stops	No info	7662
2	Jet Airways	9/06/2019	Delhi	Cochin	$\begin{array}{c} DEL \to LKO \to BOM \to \\ COK \end{array}$	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	IndiGo	12/05/2019	Kolkata	Banglore	$CCU \to NAG \to BLR$	18:05	23:30	5h 25m	1 stop	No info	6218
4	IndiGo	01/03/2019	Banglore	New Delhi	$BLR \to NAG \to DEL$	16:50	21:35	4h 45m	1 stop	No info	13302
5	SpiceJet	24/06/2019	Kolkata	Banglore	$CCU \to BLR$	09:00	11:25	2h 25m	non-stop	No info	3873
6	Jet Airways	12/03/2019	Banglore	New Delhi	$BLR \to BOM \to DEL$	18:55	10:25 13 Mar	15h 30m	1 stop	In-flight meal not included	11087
7	Jet Airways	01/03/2019	Banglore	New Delhi	$BLR \to BOM \to DEL$	08:00	05:05 02 Mar	21h 5m	1 stop	No info	22270
8	Jet Airways	12/03/2019	Banglore	New Delhi	$BLR \to BOM \to DEL$	08:55	10:25 13 Mar	25h 30m	1 stop	In-flight meal not included	11087
9	Multiple carriers	27/05/2019	Delhi	Cochin	$DEL \to BOM \to COK$	11:25	19:15	7h 50m	1 stop	No info	8625

Test Data

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
0	Jet Airways	6/06/2019	Delhi	Cochin	$DEL \to BOM \to COK$	17:30	04:25 07 Jun	10h 55m	1 stop	No info
1	IndiGo	12/05/2019	Kolkata	Banglore	$CCU \to MAA \to BLR$	06:20	10:20	4h	1 stop	No info
2	Jet Airways	21/05/2019	Delhi	Cochin	$DEL \to BOM \to COK$	19:15	19:00 22 May	23h 45m	1 stop	In-flight meal not included
3	Multiple carriers	21/05/2019	Delhi	Cochin	$DEL \to BOM \to COK$	08:00	21:00	13h	1 stop	No info
4	Air Asia	24/06/2019	Banglore	Delhi	$BLR \to DEL$	23:55	02:45 25 Jun	2h 50m	non-stop	No info
5	Jet Airways	12/06/2019	Delhi	Cochin	$DEL \to BOM \to COK$	18:15	12:35 13 Jun	18h 20m	1 stop	In-flight meal not included
6	Air India	12/03/2019	Banglore	New Delhi	$BLR \to TRV \to DEL$	07:30	22:35	15h 5m	1 stop	No info
7	IndiGo	1/05/2019	Kolkata	Banglore	$CCU \to HYD \to BLR$	15:15	20:30	5h 15m	1 stop	No info
8	IndiGo	15/03/2019	Kolkata	Banglore	CCU → BLR	10:10	12:55	2h 45m	non-stop	No info
9	Jet Airways	18/05/2019	Kolkata	Banglore	$CCU \rightarrow BOM \rightarrow BLR$	16:30	22:35	6h 5m	1 stop	No info

First I am train data

Checking null value in train data

Here we can see that in Total_stops column there is a Null value

data_train.isnull().sum()

Airline 0
Date_of_Journey 0
Source 0
Destination 0
Route 1
Dep_Time 0
Arrival_Time 0
Duration 0
Total_Stops 1
Additional_Info 0
Price 0
dtype: int64

knowing some information regarding dataset

data_train.info()

Dropping null values

```
data_train = data_train.dropna(axis = 0)
data_train.isnull().sum()
```

```
Date of Journey
Source
Destination
Route
Dep Time
                 0
                 0
Arrival Time
Duration
Total Stops
Additional Info
                 0
Price
dtype: int64
```

Using value_counts function in some of the columns

data_train["Additional_Info"].value_counts()

```
No info
                                 8344
                                 1982
In-flight meal not included
                                  320
No check-in baggage included
1 Long layover
                                   19
                                    7
Change airports
Business class
No Info
                                    3
                                    1
1 Short layover
                                    1
Red-eye flight
2 Long layover
Name: Additional Info, dtype: int64
```

data train['Duration'].value counts()

```
2h 50m
           550
1h 30m
           386
2h 45m
           337
2h 55m
           337
           329
2h 35m
31h 30m 1
30h 25m 1
42h 5m
4h 10m
47h 40m
Name: Duration, Length: 368, dtype: int64
```

```
data train['Airline'].value counts()
Jet Airways
                                     3849
IndiGo
                                     2053
Air India
                                    1751
Multiple carriers
                                    1196
```

```
SpiceJet
                                       818
Vistara
                                       479
Air Asia
                                       319
GoAir
                                       194
Multiple carriers Premium economy
                                       13
Jet Airways Business
                                         6
                                         3
Vistara Premium economy
Trujet
                                         1
```

Name: Airline, dtype: int64

data_train["Source"].value_counts()

Delhi 4536 Kolkata 2871 Banglore 2197 Mumbai 697 Chennai 381

Name: Source, dtype: int64

data_train["Destination"].value_counts()

Cochin 4536
Banglore 2871
Delhi 1265
New Delhi 932
Hyderabad 697
Kolkata 381

Name: Destination, dtype: int64

data train["Route"].value counts()

```
DEL → BOM → COK
                                       2376
                                       1552
BLR → DEL
CCU → BOM → BLR
                                        979
CCU → BLR
                                         724
BOM → HYD
                                        621
CCU \rightarrow VTZ \rightarrow BLR
CCU \rightarrow IXZ \rightarrow MAA \rightarrow BLR
                                          1
BOM → COK → MAA → HYD
\texttt{BOM} \ \rightarrow \ \texttt{CCU} \ \rightarrow \ \texttt{HYD}
BOM \rightarrow BBI \rightarrow HYD
Name: Route, Length: 128, dtype: int64
```

```
data_train["Total_Stops"].value_counts()
1 stop 5625
```

non-stop 3491

2 stops 1520
3 stops 45
4 stops 1
Name: Total Stops dtype:

Name: Total_Stops, dtype: int64

Dropping some columns

data_train =data_train.drop(['Additional_Info'],axis =1)
data_train =data_train.drop(['Route'],axis =1)

data_train.head()

	Airlin e	Date_of_Jou ey	Source	Destinati on	Dep_Ti me	Arrival_Ti me	Duratio n	Total_Sto ps	Pric e
0	IndiGo	24/03/2019	Banglor e	New Delhi	22:20	01:10 22 Mar	2h 50m	non-stop	3897
1	Air India	1/05/2019	Kolkata	Banglore	05:50	13:15	7h 25m	2 stops	7662
2	Jet Airway s	9/06/2019	Delhi	Cochin	09:25	04:25 10 Jun	19h	2 stops	1388 2
3	IndiGo	12/05/2019	Kolkata	Banglore	18:05	23:30	5h 25m	1 stop	6218
4	IndiGo	01/03/2019	Banglor e	New Delhi	16:50	21:35	4h 45m	1 stop	1330 2

Importing Date time function Further we are going to clean the data

import datetime

 $\label{lem:data_train['Day_of_journey'] = pd.to_datetime(data_train['Date_of_Journey'], format = "%d/%m/%Y").dt.day$

data_train['Month_of_journey'] =pd.to_datetime(data_train['Date_of_Journey'], format ="%d/%m/%Y").dt.month data_train.head()

	Airli ne	Date_of_ Journey	Sour ce	Destin ation	Dep_ Time	Arrival _Time	Dura tion	Total_ Stops	Pri ce	Day_of_j ourney	Month_of_ journey
0	Indi Go	24/03/201 9	Bang lore	New Delhi	22:20	01:10 22 Mar	2h 50m	non- stop	38 97	24	3
1	Air Indi a	1/05/2019	Kolk ata	Banglo re	05:50	13:15	7h 25m	2 stops	76 62	1	5
2	Jet Airw ays	9/06/2019	Delhi	Cochin	09:25	04:25 10 Jun	19h	2 stops	13 88 2	9	6
3	Indi Go	12/05/201 9	Kolk ata	Banglo re	18:05	23:30	5h 25m	1 stop	62 18	12	5
4	Indi Go	01/03/201 9	Bang lore	New Delhi	16:50	21:35	4h 45m	1 stop	13 30 2	1	3

So with the help of the DateTime function here I have separated the day and month from the Date of Journey column and made a new column for both Day of journey, Month of Journey

As I have extracted the data from the Date of journey column I am going to drop that column with the help of drop function

```
data_train = data_train.drop(['Date_of_Journey'],axis = 1)
data train.head()
```

Same extracted both minute and hour from the Dep_time column and made two new column containg

Dep min and Dep hour

```
data_train['Dep_min'] =pd.to_datetime(data_train['Dep_Time']).dt.minute data_train['Dep_hour'] =pd.to_datetime(data_train['Dep_Time']).dt.hour
```

Drop the column Dep_time

data_train =data_train.drop(['Dep_Time'],axis =1)
data_train.head()

	Airli ne	Sour ce	Destin ation	Arrival_ Time	Dura tion	Total_ Stops	Pri ce	Day_of_j ourney	Month_of_j ourney	Dep_ min	Dep_ hour
0	Indi Go	Bang lore	New Delhi	01:10 22 Mar	2h 50m	non- stop	389 7	24	3	20	22
1	Air India	Kolka ta	Banglo re	13:15	7h 25m	2 stops	766 2	1	5	50	5
2	Jet Airw ays	Delhi	Cochin	04:25 10 Jun	19h	2 stops	138 82	9	6	25	9
3	Indi Go	Kolka ta	Banglo re	23:30	5h 25m	1 stop	621 8	12	5	5	18
4	Indi Go	Bang lore	New Delhi	21:35	4h 45m	1 stop	133 02	1	3	50	16

Separate both arrival min and arrival hour from the Arrival-time column

data_train['Arrival_min'] =pd.to_datetime(data_train['Arrival_Time']).dt.minute data_train['Arrival_hour'] =pd.to_datetime(data_train['Arrival_Time']).dt.hour

data_train =data_train.drop(['Arrival_Time'],axis =1)
data_train.head()

	Airl ine	Sou rce	Desti nation	Dur atio n	Total_ Stops	Pri ce	Day_of_ journey	Month_of _journey	Dep _min	Dep_ hour	Arriva I_min	Arrival _hour
0	Indi Go	Ban glor e	New Delhi	2h 50m	non- stop	38 97	24	3	20	22	10	1
1	Air Indi a	Kolk ata	Bangl ore	7h 25m	2 stops	76 62	1	5	50	5	15	13
2	Jet Air way s	Delh i	Cochi n	19h	2 stops	13 88 2	9	6	25	9	25	4

	Airl ine	Sou rce	Desti nation	Dur atio n	Total_ Stops	Pri ce	Day_of_ journey	Month_of _journey	Dep _min	Dep_ hour	Arriva I_min	Arrival _hour
3	Indi Go	Kolk ata	Bangl ore	5h 25m	1 stop	62 18	12	5	5	18	30	23
4	Indi Go	Ban glor e	New Delhi	4h 45m	1 stop	13 30 2	1	3	50	16	35	21

For the duration I have defined a function

data_train.head()

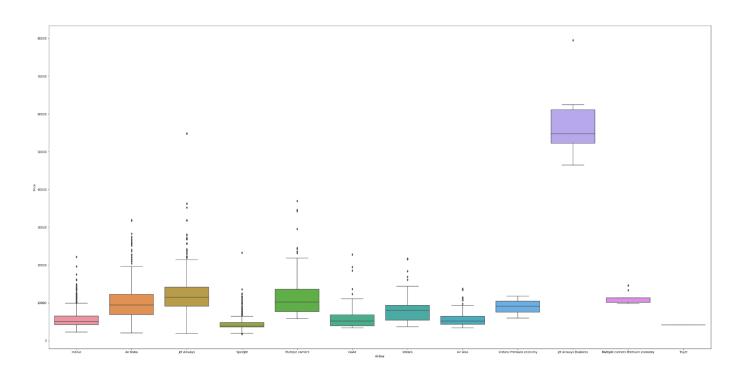
```
duration = list(data_train["Duration"])
for i in range(len(duration)):
  if len(duration[i].split()) != 2:
    if "h" in duration[i]:
      duration[i] = duration[i].strip() + " 0m" # 0m = 0 minute
    else:
      duration[i] = "0h " + duration[i] # 0h = 0 hour
duration_hours = []
duration_mins = []
for i in range(len(duration)):
  duration_hours.append(int(duration[i].split(sep = "h")[0]))
                                                                    # Extract hours from duration
  duration_mins.append(int(duration[i].split(sep = "m")[0].split()[-1])) # Extracts only minutes from
duration
data_train["Duration_hours"] = duration_hours
data_train["Duration_mins"] = duration_mins
data_train =data_train.drop(['Duration'],axis =1)
```

	Air Iin e	So urc e	Desti natio n	Total _Sto ps	Pr ic e	Day_of _journ ey	Month_ of_jour ney	Dep _mi n	Dep _ho ur	Arriv al_m in	Arriv al_ho ur	Durati on_ho urs	Durati on_mi ns
0	Ind iGo	Ban glor e	New Delhi	non- stop	38 97	24	3	20	22	10	1	2	50
1	Air Ind ia	Kol kat a	Bang lore	2 stops	76 62	1	5	50	5	15	13	7	25
2	Jet Air	Del hi	Coch in	2 stops	13 88 2	9	6	25	9	25	4	19	0

Air Iin e	urc	Desti natio n	Total _Sto ps	Pr ic e	Day_of _journ ey	Month_ of_jour ney	Dep _mi n	Dep _ho ur	Arriv al_m in	Arriv al_ho ur	Durati on_ho urs	Durati on_mi ns
wa ys												
3 Ind iGo		Bang lore	1 stop	62 18	12	5	5	18	30	23	5	25
4 Ind iGo	Ban glor e	New Delhi	1 stop	13 30 2	1	3	50	16	35	21	4	45

Using boxplot method for better visualization

```
plt.figure(figsize=(40,20))
sns.boxplot( x ="Airline" , y="Price" ,data =data_train)
plt.show()
```

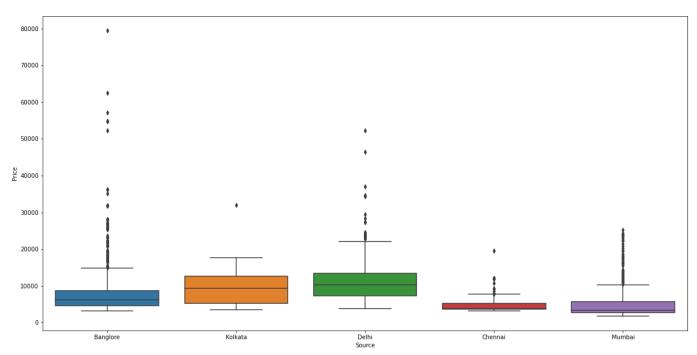


Making dummies for further model making

Airline = data_train[["Airline"]]

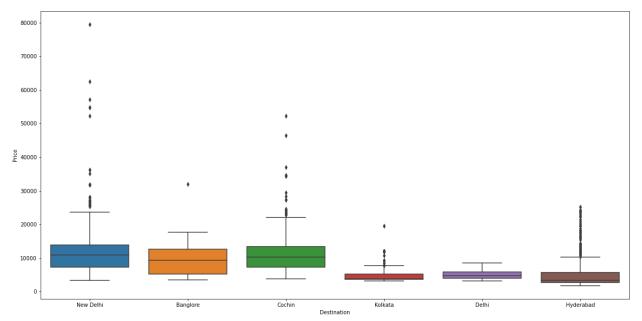
Airline = pd.get_dummies(Airline, drop_first= True)

```
plt.figure(figsize=(20,10))
sns.boxplot( x ="Source" , y="Price" ,data =data_train)
plt.show()
```



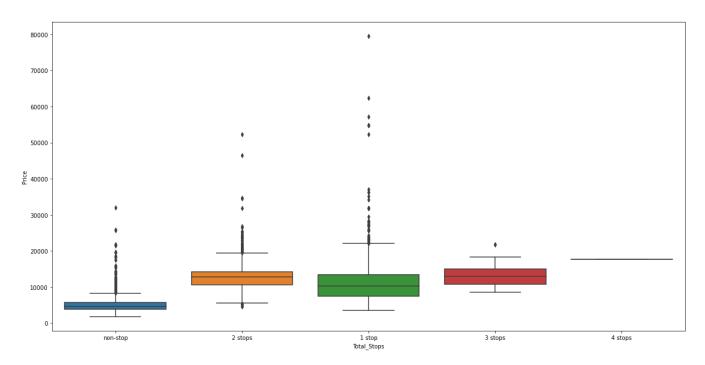
Source = data_train[["Source"]]
Source = pd.get_dummies(Source, drop_first= True)

```
plt.figure(figsize=(20,10))
sns.boxplot( x ="Destination" , y="Price" ,data =data_train)
plt.show()
```



Destination = data_train[["Destination"]]
Destination = pd.get_dummies(Destination, drop_first= True)

plt.figure(figsize=(20,10))
sns.boxplot(x ="Total_Stops" , y="Price" ,data =data_train)
plt.show()



Using replace method for model building

data_train.replace({"non-stop":0, "1 stop":1, "2 stops":2, "3 stops":3, "4 stops":4},inplace =True)

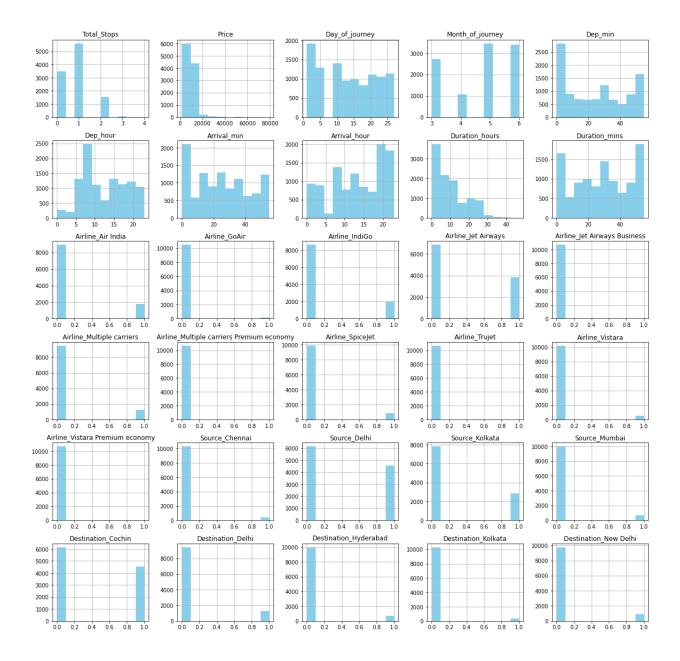
using Concat function to add up all the column

data_train1 =pd.concat([data_train,Airline,Source,Destination], axis=1)

we have these columns' data already so drooping the columns

data_train1.drop(["Airline","Source","Destination"] ,axis =1 ,inplace =True)

Using Histogram as we have cleaned all the data and for better visualization



As we have clean and visualize train data

Now further we are working on Test data I have followed same functions, methods for the cleaning of data as I have used above in train data

Dataset

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
0	Jet Airways	6/06/2019	Delhi	Cochin	DEL → BOM → COK	17:30	04:25 07 Jun	10h 55m	1 stop	No info
1	IndiGo	12/05/2019	Kolkata	Banglore	$CCU \to MAA \to BLR$	06:20	10:20	4h	1 stop	No info
2	Jet Airways	21/05/2019	Delhi	Cochin	$DEL \to BOM \to COK$	19:15	19:00 22 May	23h 45m	1 stop	In-flight meal not included
3	Multiple carriers	21/05/2019	Delhi	Cochin	$DEL \to BOM \to COK$	08:00	21:00	13h	1 stop	No info
4	Air Asia	24/06/2019	Banglore	Delhi	$BLR \to DEL$	23:55	02:45 25 Jun	2h 50m	non-stop	No info
5	Jet Airways	12/06/2019	Delhi	Cochin	$DEL \to BOM \to COK$	18:15	12:35 13 Jun	18h 20m	1 stop	In-flight meal not included
6	Air India	12/03/2019	Banglore	New Delhi	$BLR \to TRV \to DEL$	07:30	22:35	15h 5m	1 stop	No info
7	IndiGo	1/05/2019	Kolkata	Banglore	$CCU \to HYD \to BLR$	15:15	20:30	5h 15m	1 stop	No info
8	IndiGo	15/03/2019	Kolkata	Banglore	$CCU \to BLR$	10:10	12:55	2h 45m	non-stop	No info
9	Jet Airways	18/05/2019	Kolkata	Banglore	$CCU \rightarrow BOM \rightarrow BLR$	16:30	22:35	6h 5m	1 stop	No info

Checking the shape of dataset using .shape function

data_test.shape
(2671, 10)

This dataset contains 2671 rows and 10 columns

Checking information in the dataset

data_test.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2671 entries, 0 to 2670
Data columns (total 10 columns):

Daca	COTAMIND (COCAT I	o coramino,.	
#	Column	Non-Null Count	Dtype
0	Airline	2671 non-null	object
1	Date_of_Journey	2671 non-null	object
2	Source	2671 non-null	object
3	Destination	2671 non-null	object
4	Route	2671 non-null	object
5	Dep_Time	2671 non-null	object
6	Arrival_Time	2671 non-null	object
7	Duration	2671 non-null	object
8	Total_Stops	2671 non-null	object
9	Additional Info	2671 non-null	object

dtypes: object(10) memory usage: 208.8+ KB

Dropping Null values

Drooped irrelevant columns from the dataset

```
data_test =data_test.drop(['Additional_Info'],axis =1)
data_test =data_test.drop(['Route'],axis =1)
data_test.head()
```

	Airline	Date_of_Journe y	Source	Destinatio n	Dep_Tim e	Arrival_Tim e	Duratio n	Total_Stop s
0	Jet Airway s	6/06/2019	Delhi	Cochin	17:30	04:25 07 Jun	10h 55m	1 stop
1	IndiGo	12/05/2019	Kolkata	Banglore	06:20	10:20	4h	1 stop
2	Jet Airway s	21/05/2019	Delhi	Cochin	19:15	19:00 22 May	23h 45m	1 stop
3	Multipl e carrier s	21/05/2019	Delhi	Cochin	08:00	21:00	13h	1 stop
4	Air Asia	24/06/2019	Banglor e	Delhi	23:55	02:45 25 Jun	2h 50m	non-stop

With the help of date time separating minutes hours

```
data_test['Day_of_journey'] =pd.to_datetime(data_test['Date_of_Journey'], format
="%d/%m/%Y").dt.day
data_test['Month_of_journey'] =pd.to_datetime(data_test['Date_of_Journey'], format
="%d/%m/%Y").dt.month
data_test =data_test.drop(['Date_of_Journey'],axis =1)
data_test.head()
```

	Airline	Source	Destinatio n	Dep_Tim e	Arrival_Tim e	Duratio n	Total_Stop s	Day_of_journe y	Month_of_journ ey
0	Jet Airway s	Delhi	Cochin	17:30	04:25 07 Jun	10h 55m	1 stop	6	6
1	IndiGo	Kolkata	Banglore	06:20	10:20	4h	1 stop	12	5
2	Jet Airway s	Delhi	Cochin	19:15	19:00 22 May	23h 45m	1 stop	21	5
3	Multipl e carrier s	Delhi	Cochin	08:00	21:00	13h	1 stop	21	5
4	Air Asia	Banglor e	Delhi	23:55	02:45 25 Jun	2h 50m	non-stop	24	6

Same for other columns

```
data_test['Dep_min'] =pd.to_datetime(data_test['Dep_Time']).dt.minute
data_test['Dep_hour'] =pd.to_datetime(data_test['Dep_Time']).dt.hour
data_test =data_test.drop(['Dep_Time'],axis =1)
data test['Arrival min'] =pd.to datetime(data test['Arrival Time']).dt.minute
data_test['Arrival_hour'] =pd.to_datetime(data_test['Arrival_Time']).dt.hour
data_test =data_test.drop(['Arrival_Time'],axis =1)
duration = list(data_test["Duration"])
for i in range(len(duration)):
  if len(duration[i].split()) != 2:
    if "h" in duration[i]:
      duration[i] = duration[i].strip() + " 0m" # 0m = 0 minute
    else:
      duration[i] = "0h " + duration[i] # 0h = 0 hour
duration_hours = []
duration_mins = []
for i in range(len(duration)):
```

duration_hours.append(int(duration[i].split(sep = "h")[0])) # Extract hours from duration
duration_mins.append(int(duration[i].split(sep = "m")[0].split()[-1])) # Extracts only minutes from
duration

```
data_test["Duration_hours"] = duration_hours
data_test["Duration_mins"] = duration_mins
data_test = data_test.drop(['Duration'],axis = 1)
```

Creating dummies

```
Airline = data_test[["Airline"]]
Airline = pd.get_dummies(Airline, drop_first= True)

Source = data_test[["Source"]]
Source = pd.get_dummies(Source, drop_first= True)

Destination = data_test[["Destination"]]
Destination = pd.get_dummies(Destination, drop_first= True)
```

Replacing the values

data_test.replace({"non-stop":0, "1 stop":1, "2 stops":2, "3 stops":3, "4 stops":4},inplace =True)

	Airl ine	Sou rce	Desti natio n	Total _Stop s	Day_of _journe y	Month_o f_journe y	Dep _mi n	Dep _ho ur	Arriv al_mi n	Arriva I_hou r	Duratio n_hour s	Durati on_mi ns
0	Jet Air wa ys	Del hi	Cochi n	1	6	6	30	17	25	4	10	55
1	Indi Go	Kolk ata	Bangl ore	1	12	5	20	6	20	10	4	0
2	Jet Air wa ys	Del hi	Cochi n	1	21	5	15	19	0	19	23	45
3	Mul tipl e carr iers	Del hi	Cochi n	1	21	5	0	8	0	21	13	0

Airl ine	Sou rce	natio	_Stop	_journe	Month_o f_journe y	_mi	_ho	al_mi	l_hou	n_hour	Durati on_mi ns
Air 4 Asi a	glor	Delhi	0	24	6	55	23	45	2	2	50

Adding the columns using concatenation function

data_test1 =pd.concat([data_test,Airline,Source,Destination], axis=1)

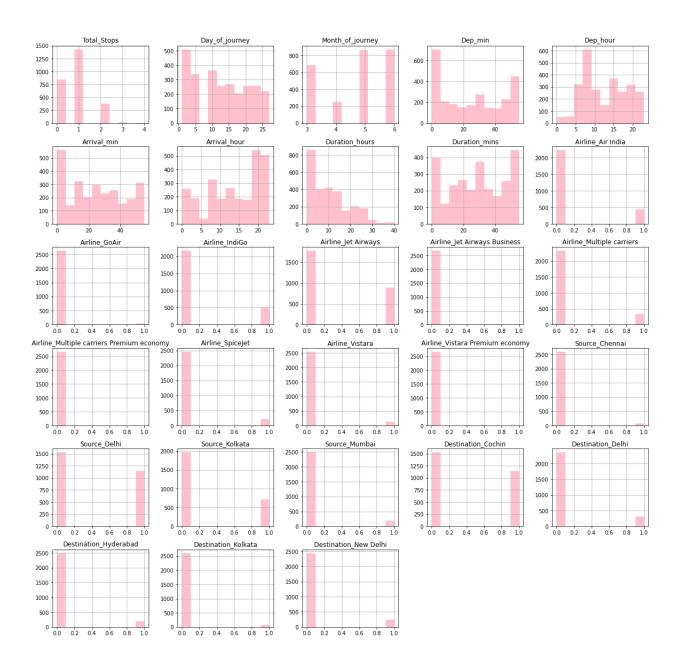
we have these columns data already so drooping the columns data_test1.drop(["Airline","Source","Destination"],axis =1,inplace =True)

Checking the size of the dataset

data_test1.shape
(2671, 28)

Plotting histogram

data_test1.hist(color ="pink" , figsize =(20,20))
plt.show()



Building model

Separating features and label

x =data_train1.drop(['Price'], axis =1)
y =data_train1['Price']

Using Boxplot for better visualization

plt.figure(figsize=(50,50), facecolor='pink')

graph=1

```
for column in x:
    if graph<=30:
        ax=plt.subplot(10,3,graph)
        sns.boxplot(x[column])
    plt.xlabel(column,fontsize=20)</pre>
```

graph+=1 plt.show()



```
And Fitting and transform the scaled data
sc = StandardScaler()
sc.fit transform(x)
Using train test split for further model building
x_train,x_test,y_train,y_test = train_test_split(x,y, test_size=0.2 ,random_state=5)
using linear regression for predictions
Ir = LinearRegression()
lr.fit(x train,y train)
Ir = LinearRegression()
lr.fit(x train,y train)
array([11231.49084379, 4547.90828167, 11436.03552346, ...,
         5338.02095875, 3888.54931379, 4108.6131531 ])
print(r2_score(y_pred,y_test))
0.35664214783409176
Used different model for better prediction
Using Random forest classifier
# Random Forest Classifier
rf =RandomForestClassifier()
rf.fit(x_train,y_train)
y predict =rf.predict(x test)
y predict
array([14781, 4483, 6795, ..., 5224, 3100, 3841], dtype=int64)
rf.score(x train, y train)
0.8821533060269163
rf.score(x test,y test)
0.33879270004679457
```

Using Standard Scaler for scaling

Calculating Metrics

```
MAE: 1485.3093121197942
MSE 7093620.016846046
RMSE: 2663.3850673242964

metrics.r2_score(y_test,y_predict)
0.6884453949370779

Using Decision Tree Classifier

# Decision Tree Classifier
dtc =DecisionTreeClassifier()
dtc.fit(x_train,y_train)

y_prediction =dtc.predict(x_test)
y_prediction
array([14781, 3383, 6795, ..., 5224, 3100, 3841], dtype=int64)

metrics.r2_score(y_test,y_prediction)
0.6535169589522694
```

Conclusion

In this article, we saw how to apply Different libraries to choose the best machine learning algorithm for the task at hand.

We analyzed the dataset and then find the null values, information regarding the dataset removed all the Null values and then used some methods to clean the data. and build a machine learning model further.

They used a standard scaler for the scaling and tried different machine learning models for better prediction or result. One can choose either based on the situation at hand.

thanku