Problem statement

Flight price prediction

In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sn
```

In [2]:

train_df=pd.read_csv(r"C:\Users\chait\Downloads\Data_Train 1.csv")
train_df

Out[2]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Dura
0	IndiGo	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h
1	Air India	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h
2	Jet Airways	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	
3	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h
4	IndiGo	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h
10678	Air Asia	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h
10679	Air India	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h
10680	Jet Airways	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	
10681	Vistara	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h
10682	Air India	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h

In [3]:

test_df=pd.read_csv(r"C:\Users\chait\Downloads\Test_set.csv")
test_df

Out[3]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durat
0	Jet Airways	6/06/2019	Delhi	Cochin	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h £
1	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? MAA ? BLR	06:20	10:20	
2	Jet Airways	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	19:15	19:00 22 May	23h 4
3	Multiple carriers	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	08:00	21:00	
4	Air Asia	24/06/2019	Banglore	Delhi	BLR ? DEL	23:55	02:45 25 Jun	2h ŧ
2666	Air India	6/06/2019	Kolkata	Banglore	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h £
2667	IndiGo	27/03/2019	Kolkata	Banglore	CCU ? BLR	14:20	16:55	2h 3
2668	Jet Airways	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 3
2669	Air India	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	04:00	19:15	15h 1
2670	Multiple carriers	15/06/2019	Delhi	Cochin	DEL ? BOM ? COK	04:55	19:15	14h 2

In [4]:

train_df.head()

Out[4]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration
0	IndiGo	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m
1	Air India	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m
2	Jet Airways	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h
3	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m
4	IndiGo	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m
4		_	_	_	_			

In [5]:

test_df.head()

Out[5]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration
0	Jet Airways	6/06/2019	Delhi	Cochin	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h 55m
1	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? MAA ? BLR	06:20	10:20	4h
2	Jet Airways	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	19:15	19:00 22 May	23h 45m
3	Multiple carriers	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	08:00	21:00	13h
4	Air Asia	24/06/2019	Banglore	Delhi	BLR ? DEL	23:55	02:45 25 Jun	2h 50m
4								

In [6]:

train_df.tail()

Out[6]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Dura
10678	Air Asia	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h
10679	Air India	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h
10680	Jet Airways	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	
10681	Vistara	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h
10682	Air India	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h
4								

In [7]:

test_df.tail()

Out[7]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duratio
2666	Air India	6/06/2019	Kolkata	Banglore	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 55
2667	IndiGo	27/03/2019	Kolkata	Banglore	CCU ? BLR	14:20	16:55	2h 35
2668	Jet Airways	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 35
2669	Air India	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	04:00	19:15	15h 15
2670	Multiple carriers	15/06/2019	Delhi	Cochin	DEL ? BOM ? COK	04:55	19:15	14h 20
4								

In [8]:

train_df.describe()

Out[8]:

	Price
count	10683.000000
mean	9087.064121
std	4611.359167
min	1759.000000
25%	5277.000000
50%	8372.000000
75%	12373.000000
max	79512.000000

In [9]:

test_df.describe()

Out[9]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Dura
count	2671	2671	2671	2671	2671	2671	2671	:
unique	11	44	5	6	100	199	704	
top	Jet Airways	9/05/2019	Delhi	Cochin	DEL ? BOM ? COK	10:00	19:00	2h
freq	897	144	1145	1145	624	62	113	
4								

```
In [10]:
train_df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10683 entries, 0 to 10682
Data columns (total 11 columns):
                     Non-Null Count Dtype
#
    Column
    Airline
                     10683 non-null object
0
 1
    Date_of_Journey 10683 non-null object
 2
    Source
                     10683 non-null object
 3
    Destination
                     10683 non-null object
 4
    Route
                     10682 non-null object
 5
    Dep_Time
                     10683 non-null object
    Arrival_Time
 6
                     10683 non-null object
 7
    Duration
                     10683 non-null object
    Total_Stops
 8
                     10682 non-null object
 9
    Additional_Info 10683 non-null object
                     10683 non-null int64
10 Price
dtypes: int64(1), object(10)
memory usage: 918.2+ KB
In [11]:
test_df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2671 entries, 0 to 2670
Data columns (total 10 columns):
```

#	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

In [12]:

```
train_df.size
```

Out[12]:

117513

```
In [13]:
    test_df.size

Out[13]:
26710

In [14]:
    train_df.shape

Out[14]:
    (10683, 11)

In [15]:
    test_df.shape

Out[15]:
    (2671, 10)
```

Exploratory data analysis

```
In [16]:
```

```
train_df.isnull().sum()
```

Out[16]:

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

```
In [17]:
test_df.isnull().sum()
Out[17]:
Airline
                    0
Date_of_Journey
                    0
Source
                    0
Destination
                    0
Route
                    0
Dep_Time
                    0
                    0
Arrival_Time
Duration
                    0
Total_Stops
                    0
Additional_Info
                    0
dtype: int64
In [18]:
train_df.duplicated().sum()
Out[18]:
220
In [19]:
test_df.duplicated().sum()
Out[19]:
26
In [20]:
train_df.dropna(inplace=True)
In [21]:
train_df.isnull().sum()
Out[21]:
Airline
                    0
Date_of_Journey
                    0
Source
                    0
Destination
                    0
Route
                    0
Dep_Time
                    0
Arrival_Time
                    0
                    0
Duration
Total_Stops
                    0
                    0
Additional_Info
Price
                    0
dtype: int64
```

```
In [22]:
```

```
train_df['Airline'].value_counts()
```

Out[22]:

Airline 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 Vistara Premium economy 3 Trujet 1 Name: count, dtype: int64

In [23]:

```
train_df['Source'].value_counts()
```

Out[23]:

Source

Delhi 4536 Kolkata 2871 Banglore 2197 Mumbai 697 Chennai 381

Name: count, dtype: int64

In [24]:

```
train_df['Destination'].value_counts()
```

Out[24]:

Destination

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

Name: count, dtype: int64

In [25]:

```
train_df['Total_Stops'].value_counts()
```

Out[25]:

Total_Stops
1 stop 5625
non-stop 3491
2 stops 1520
3 stops 45
4 stops 1

Name: count, dtype: int64

In [26]:

```
airline={"Airline":{"Jet Airways":0,"IndiGo":1,"Air India":2,"Multiple carriers":3,
    "SpiceJet":4,"Vistara":5,"Air Asia":6,"GoAir":7,
    "Multiple carriers Premium economy":8,
    "Jet Airways Business":9,"Vistara Premium economy":10,"Trujet":11}}
train_df=train_df.replace(airline)
train_df
```

Out[26]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Dura
0	1	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h :
1	2	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h :
2	0	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	
3	1	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h :
4	1	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 4
10678	6	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h (
10679	2	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h (
10680	0	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	
10681	5	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 4
10682	2	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h :

In [27]:

```
city={"Source":{"Delhi":0,"Kolkata":1,"Banglore":2,
"Mumbai":3,"Chennai":4}}
train_df=train_df.replace(city)
train_df
```

Out[27]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durati
0	1	24/03/2019	2	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 5(
1	2	1/05/2019	1	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 2ŧ
2	0	9/06/2019	0	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	1
3	1	12/05/2019	1	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 2ŧ
4	1	01/03/2019	2	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 4ŧ
10678	6	9/04/2019	1	Banglore	CCU ? BLR	19:55	22:25	2h 3(
10679	2	27/04/2019	1	Banglore	CCU ? BLR	20:45	23:20	2h 3ŧ
10680	0	27/04/2019	2	Delhi	BLR ? DEL	08:20	11:20	
10681	5	01/03/2019	2	New Delhi	BLR ? DEL	11:30	14:10	2h 4(
10682	2	9/05/2019	0	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 2(

In [28]:

```
destination={"Destination":{"Cochin":0,"Banglore":1,"Delhi":2,
"New Delhi":3,"Hyderabad":4,"Kolkata":5}}
train_df=train_df.replace(destination)
train_df
```

Out[28]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durati
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 5(
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 2!
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	1
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 2ŧ
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 4 !
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 3(
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 3ŧ
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 4(
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 2(

In [29]:

```
stops={"Total_Stops":{"non-stop":0,"1 stop":1,"2 stops":2,
"3 stops":3,"4 stops":4}}
train_df=train_df.replace(stops)
train_df
```

Out[29]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durati
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 5(
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 2ŧ
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	1
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 2ŧ
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 4!
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 3(
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 3ŧ
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 4(
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 2(

10682 rows × 11 columns

data visualization

In [30]:

```
import seaborn as sns
df=train_df[['Airline','Source','Destination','Total_Stops','Price']]
sns.heatmap(df.corr(),annot=True)
```

Out[30]:

<Axes: >



Linear regression

In [31]:

```
x=df[['Airline','Source','Destination','Total_Stops']]
y=df['Price']
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=100)
```

In [32]:

```
#Data prediction and Evaluation
from sklearn.linear_model import LinearRegression
regr=LinearRegression()
regr.fit(X_train,y_train)
print(regr.intercept_)
coeff_df=pd.DataFrame(regr.coef_,x.columns,columns=['coefficient'])
coeff_df
```

7211.098088897488

Out[32]:

	coefficient
Airline	-418.483922
Source	-3275.073380
Destination	2505.480291
Total_Stops	3541.798053

In [33]:

```
score=regr.score(X_test,y_test)
print(score)
```

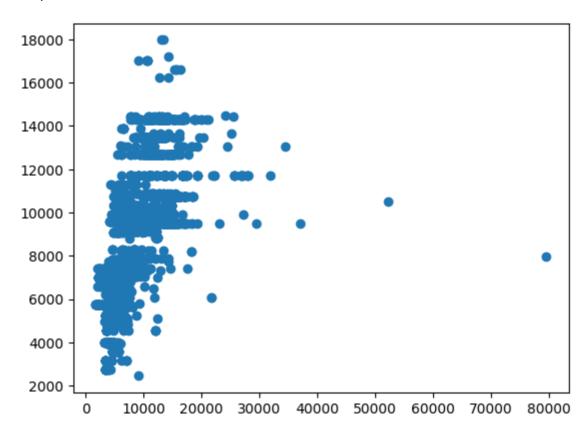
0.4108304890928348

In [34]:

```
predictions=regr.predict(X_test)
plt.scatter(y_test,predictions)
```

Out[34]:

<matplotlib.collections.PathCollection at 0x2ab41d02dd0>



In [35]:

```
x=np.array(df['Price']).reshape(-1,1)
y=np.array(df['Total_Stops']).reshape(-1,1)
df.dropna(inplace=True)
```

C:\Users\chait\AppData\Local\Temp\ipykernel_29576\3039801757.py:3: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

df.dropna(inplace=True)

In [36]:

```
X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
regr.fit(X_train,y_train)
regr.fit(X_train,y_train)
```

Out[36]:

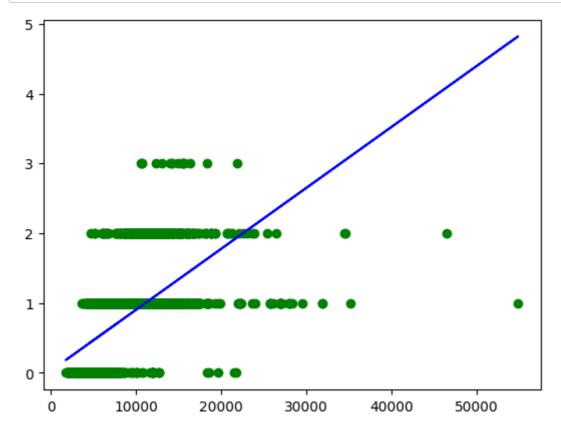
LinearRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

In [37]:

```
y_pred=regr.predict(X_test)
plt.scatter(X_test,y_test,color='g')
plt.plot(X_test,y_pred,color='b')
plt.show()
```



logistic regression

```
In [38]:
```

```
x=np.array(df['Price']).reshape(-1,1)
y=np.array(df['Total_Stops']).reshape(-1,1)
df.dropna(inplace=True)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=1)
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression(max_iter=10000)
import warnings
warnings.simplefilter(action='ignore')
```

C:\Users\chait\AppData\Local\Temp\ipykernel_29576\1264944960.py:3: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)
df.dropna(inplace=True)

In [39]:

```
lr.fit(x_train,y_train)
```

Out[39]:

LogisticRegression(max_iter=10000)

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```
In [40]:
```

```
score=lr.score(x_test,y_test)
print(score)
```

0.7160686427457098

decision tree

In [42]:

```
from sklearn.tree import DecisionTreeClassifier
clf=DecisionTreeClassifier(random_state=0)
clf.fit(x_train,y_train)
```

Out[42]:

DecisionTreeClassifier(random_state=0)

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In [43]:

```
score=clf.score(x_test,y_test)
print(score)
```

0.9369734789391576

Random forest

In [44]:

```
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(X_train,y_train)
```

Out[44]:

RandomForestClassifier()

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In [45]:

```
params={'max_depth':[2,3,5,10,20],
    'min_samples_leaf':[5,10,20,50,100,200],
    'n_estimators':[10,25,30,50,100,200]}
```

In [46]:

```
from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rfc,param_grid=params,cv=2,scoring="accuracy")
grid_search.fit(X_train,y_train)
```

Out[46]:

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

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```
In [47]:
```

```
grid_search.best_score_
```

Out[47]:

0.523605715699528

In [48]:

```
rf_best=grid_search.best_estimator_
rf_best
```

Out[48]:

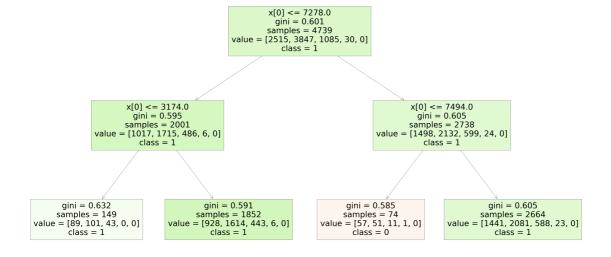
RandomForestClassifier(max_depth=2, min_samples_leaf=5)

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

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In [49]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[4],class_names=['0','1','2','3','4'],filled=True);
```



In [50]:

```
score=rfc.score(x_test,y_test)
print(score)
```

0.4642745709828393

Conclusion

In []:

The Flight price prediction dataset is used to find best model. In the decision tree model has 93% accuracy. In Linear regression 41% accuracy. in Logistic regression 71% accuracy.in random forest 48% accuracy.