

# MINI PROJECT - 2

## Problem Statement: Which model is suitable best for Flight price Prediction Dataset

In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import LogisticRegression
```

## Data collection

In [2]:

```
train_df=pd.read_csv(r"C:\Users\91950\Downloads\Data_Train11.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

10683 rows × 11 columns



In [3]:

```
test_df=pd.read_csv(r"C:\Users\91950\Downloads\Test_set22.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 5
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 5
...	...	...	...	...	...	...	...	
2666	Air India	6/06/2019	Kolkata	Banglore	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 5
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

2671 rows × 10 columns



# Data cleaning and preproceesing

In [4]:

```
train_df.shape
```

Out[4]:

```
(10683, 11)
```

In [5]:

```
test_df.shape
```

Out[5]:

```
(2671, 10)
```

In [6]:

```
train_df.describe
```

Out[6]:

<bound method NDFrame.describe of				Airline	Date_of_Journey	So
urce Destination						
0	IndiGo	24/03/2019	Banglore	New Delhi	\	
1	Air India	1/05/2019	Kolkata	Banglore		
2	Jet Airways	9/06/2019	Delhi	Cochin		
3	IndiGo	12/05/2019	Kolkata	Banglore		
4	IndiGo	01/03/2019	Banglore	New Delhi		
...	...	...	...	...		
10678	Air Asia	9/04/2019	Kolkata	Banglore		
10679	Air India	27/04/2019	Kolkata	Banglore		
10680	Jet Airways	27/04/2019	Banglore	Delhi		
10681	Vistara	01/03/2019	Banglore	New Delhi		
10682	Air India	9/05/2019	Delhi	Cochin		

	Route	Dep_Time	Arrival_Time	Duration	Total_Stops
0	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop
\					
1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops
2	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops
3	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop
4	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop
...	...	...	...	...	...
10678	CCU ? BLR	19:55	22:25	2h 30m	non-stop
10679	CCU ? BLR	20:45	23:20	2h 35m	non-stop
10680	BLR ? DEL	08:20	11:20	3h	non-stop
10681	BLR ? DEL	11:30	14:10	2h 40m	non-stop
10682	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops

	Additional_Info	Price
0	No info	3897
1	No info	7662
2	No info	13882
3	No info	6218
4	No info	13302
...	...	...
10678	No info	4107
10679	No info	4145
10680	No info	7229
10681	No info	12648
10682	No info	11753

[10683 rows x 11 columns]>

In [7]:

test\_df.describe

Out[7]:

<bound method NDFrame.describe of  
Source Destination

	Airline	Date_of_Journey	Source	Destination
0	Jet Airways	6/06/2019	Delhi	Cochin \
1	IndiGo	12/05/2019	Kolkata	Banglore
2	Jet Airways	21/05/2019	Delhi	Cochin
3	Multiple carriers	21/05/2019	Delhi	Cochin
4	Air Asia	24/06/2019	Banglore	Delhi
...	...	...	...	...
2666	Air India	6/06/2019	Kolkata	Banglore
2667	IndiGo	27/03/2019	Kolkata	Banglore
2668	Jet Airways	6/03/2019	Delhi	Cochin
2669	Air India	6/03/2019	Delhi	Cochin
2670	Multiple carriers	15/06/2019	Delhi	Cochin

	Route	Dep_Time	Arrival_Time	Duration	Total_Stops
0	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h 55m	1 stop \
1	CCU ? MAA ? BLR	06:20	10:20	4h	1 stop
2	DEL ? BOM ? COK	19:15	19:00 22 May	23h 45m	1 stop
3	DEL ? BOM ? COK	08:00	21:00	13h	1 stop
4	BLR ? DEL	23:55	02:45 25 Jun	2h 50m	non-stop
...	...	...	...	...	...
2666	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 55m	1 stop
2667	CCU ? BLR	14:20	16:55	2h 35m	non-stop
2668	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 35m	1 stop
2669	DEL ? BOM ? COK	04:00	19:15	15h 15m	1 stop
2670	DEL ? BOM ? COK	04:55	19:15	14h 20m	1 stop

	Additional_Info
0	No info
1	No info
2	In-flight meal not included
3	No info
4	No info
...	...
2666	No info
2667	No info
2668	No info
2669	No info
2670	No info

[2671 rows x 10 columns]&gt;

In [8]:

train\_df.head

Out[8]:

```

<bound method NDFrame.head of
Destination
0      IndiGo      24/03/2019  Bangalore  New Delhi  \
1      Air India    1/05/2019  Kolkata    Bangalore
2      Jet Airways  9/06/2019   Delhi      Cochin
3      IndiGo      12/05/2019  Kolkata    Bangalore
4      IndiGo      01/03/2019  Bangalore  New Delhi
...      ...      ...      ...      ...
10678   Air Asia    9/04/2019   Kolkata    Bangalore
10679   Air India   27/04/2019   Kolkata    Bangalore
10680   Jet Airways 27/04/2019   Bangalore  Delhi
10681   Vistara     01/03/2019   Bangalore  New Delhi
10682   Air India    9/05/2019   Delhi      Cochin

      Route Dep_Time  Arrival_Time  Duration  Total_Stops
0      BLR ? DEL    22:20    01:10 22 Mar    2h 50m    non-stop
\
1      CCU ? IXR ? BBI ? BLR    05:50          13:15    7h 25m    2 stops
2      DEL ? LKO ? BOM ? COK    09:25    04:25 10 Jun    19h    2 stops
3      CCU ? NAG ? BLR    18:05          23:30    5h 25m    1 stop
4      BLR ? NAG ? DEL    16:50          21:35    4h 45m    1 stop
...      ...      ...      ...      ...
10678      CCU ? BLR    19:55          22:25    2h 30m    non-stop
10679      CCU ? BLR    20:45          23:20    2h 35m    non-stop
10680      BLR ? DEL    08:20          11:20     3h    non-stop
10681      BLR ? DEL    11:30          14:10    2h 40m    non-stop
10682  DEL ? GOI ? BOM ? COK    10:55          19:15    8h 20m    2 stops

      Additional_Info  Price
0      No info    3897
1      No info    7662
2      No info   13882
3      No info    6218
4      No info   13302
...      ...      ...
10678      No info    4107
10679      No info    4145
10680      No info    7229
10681      No info   12648
10682      No info   11753

[10683 rows x 11 columns]>

```

In [9]:

test\_df.head

Out[9]:

```
<bound method NDFrame.head of
ource Destination
```

	Airline	Date_of_Journey	S
0	Jet Airways	6/06/2019	Delhi Cochin \
1	IndiGo	12/05/2019	Kolkata Bangalore
2	Jet Airways	21/05/2019	Delhi Cochin
3	Multiple carriers	21/05/2019	Delhi Cochin
4	Air Asia	24/06/2019	Banglore Delhi
...	...	...	...
2666	Air India	6/06/2019	Kolkata Bangalore
2667	IndiGo	27/03/2019	Kolkata Bangalore
2668	Jet Airways	6/03/2019	Delhi Cochin
2669	Air India	6/03/2019	Delhi Cochin
2670	Multiple carriers	15/06/2019	Delhi Cochin

	Route	Dep_Time	Arrival_Time	Duration	Total_Stops
0	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h 55m	1 stop \
1	CCU ? MAA ? BLR	06:20	10:20	4h	1 stop
2	DEL ? BOM ? COK	19:15	19:00 22 May	23h 45m	1 stop
3	DEL ? BOM ? COK	08:00	21:00	13h	1 stop
4	BLR ? DEL	23:55	02:45 25 Jun	2h 50m	non-stop
...	...	...	...	...	...
2666	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 55m	1 stop
2667	CCU ? BLR	14:20	16:55	2h 35m	non-stop
2668	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 35m	1 stop
2669	DEL ? BOM ? COK	04:00	19:15	15h 15m	1 stop
2670	DEL ? BOM ? COK	04:55	19:15	14h 20m	1 stop

	Additional_Info
0	No info
1	No info
2	In-flight meal not included
3	No info
4	No info
...	...
2666	No info
2667	No info
2668	No info
2669	No info
2670	No info

[2671 rows x 10 columns]&gt;



In [10]:

train\_df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10683 entries, 0 to 10682
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   Airline                10683 non-null  object
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
6   Arrival_Time           10683 non-null  object
7   Duration               10683 non-null  object
8   Total_Stops            10682 non-null  object
9   Additional_Info        10683 non-null  object
10  Price                  10683 non-null  int64
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
```

## TO FIND MISSING VALUES

In [12]:

```
train_df.isna().sum()
```

Out[12]:

```
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
```

In [13]:

```
test_df.isna().sum()
```

Out[13]:

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

In [14]:

```
train_df.duplicated().sum()
```

Out[14]:

```
220
```

In [15]:

```
test_df.duplicated().sum()
```

Out[15]:

```
26
```

In [16]:

```
train_df.dropna(inplace=True)
```

In [17]:

```
train_df.isna().sum()
```

Out[17]:

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

In [18]:

```
train_df.shape
```

Out[18]:

```
(10682, 11)
```

## Feature selection

In [19]:

```
train_df.columns
```

Out[19]:

```
Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
       'Dep_Time', 'Arrival_Time', 'Duration', 'Total_Stops',
       'Additional_Info', 'Price'],
      dtype='object')
```

In [20]:

```
test_df.columns
```

Out[20]:

```
Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
       'Dep_Time', 'Arrival_Time', 'Duration', 'Total_Stops',
       'Additional_Info'],
      dtype='object')
```

In [21]:

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

Out[21]:

```
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 [22]:

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

Out[22]:

```
Source
Delhi      4536
Kolkata    2871
Bangalore  2197
Mumbai     697
Chennai    381
Name: count, dtype: int64
```

In [23]:

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

Out[23]:

```
Destination
Cochin      4536
Bangalore   2871
Delhi       1265
New Delhi   932
Hyderabad   697
Kolkata     381
Name: count, dtype: int64
```

In [24]:

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

Out[24]:

```
Total_Stops
1 stop      5625
non-stop    3491
2 stops     1520
3 stops       45
4 stops        1
Name: count, dtype: int64
```

In [25]:

```
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[25]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durat
0	1	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 5
1	2	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 2
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 2
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 3
10679	2	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 3
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 2

10682 rows × 11 columns



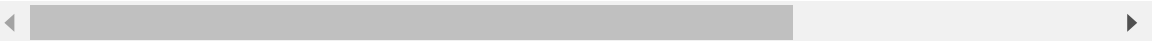
In [26]:

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

Out[26]:

	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 50
1	2	1/05/2019	1	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25
2	0	9/06/2019	0	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	1h
3	1	12/05/2019	1	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25
4	1	01/03/2019	2	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45
...	...	...	...	...	...	...	...	...
10678	6	9/04/2019	1	Banglore	CCU ? BLR	19:55	22:25	2h 30
10679	2	27/04/2019	1	Banglore	CCU ? BLR	20:45	23:20	2h 35
10680	0	27/04/2019	2	Delhi	BLR ? DEL	08:20	11:20	3h
10681	5	01/03/2019	2	New Delhi	BLR ? DEL	11:30	14:10	2h 40
10682	2	9/05/2019	0	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20

10682 rows × 11 columns



In [27]:

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

Out[27]:

	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 50
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	1h
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45
...	...	...	...	...	...	...	...	...
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	3h
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 40
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20

10682 rows × 11 columns





In [28]:

```
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[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 50
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	1h
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45
...	...	...	...	...	...	...	...	...
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	3h
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 40
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20

10682 rows × 11 columns

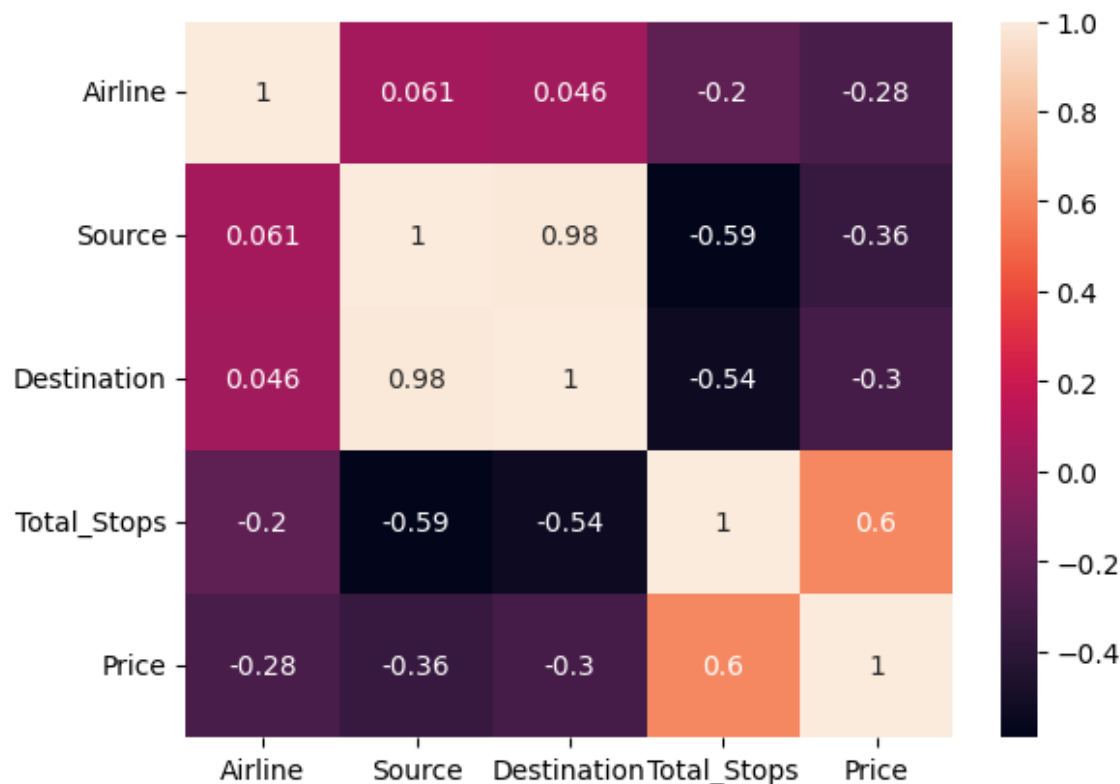


In [29]:

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

Out[29]:

&lt;Axes: &gt;



## Linear regression for training data

In [30]:

```
x=np.array(train_df['Source']).reshape(-1,1)
y=np.array(train_df['Destination']).reshape(-1,1)
```

In [31]:

```
x_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=100)
```

In [32]:

```
from sklearn.linear_model import LinearRegression
regr=LinearRegression()
regr.fit(x_train,y_train)
print(regr.intercept_)
print(regr.coef_)
```

```
[-0.08091813]
[[1.26597749]]
```

In [33]:

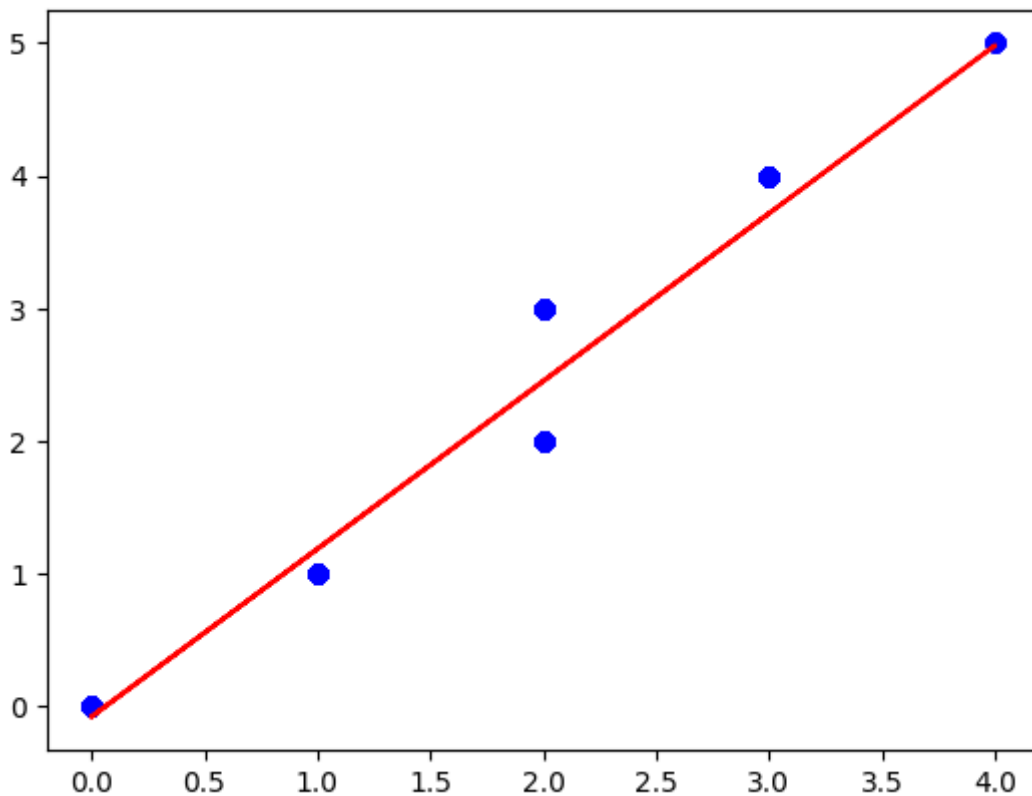
```
X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
regr.fit(X_train,y_train)
regr.fit(X_train,y_train)
```

Out[33]:

```
LinearRegression
LinearRegression()
```

In [34]:

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

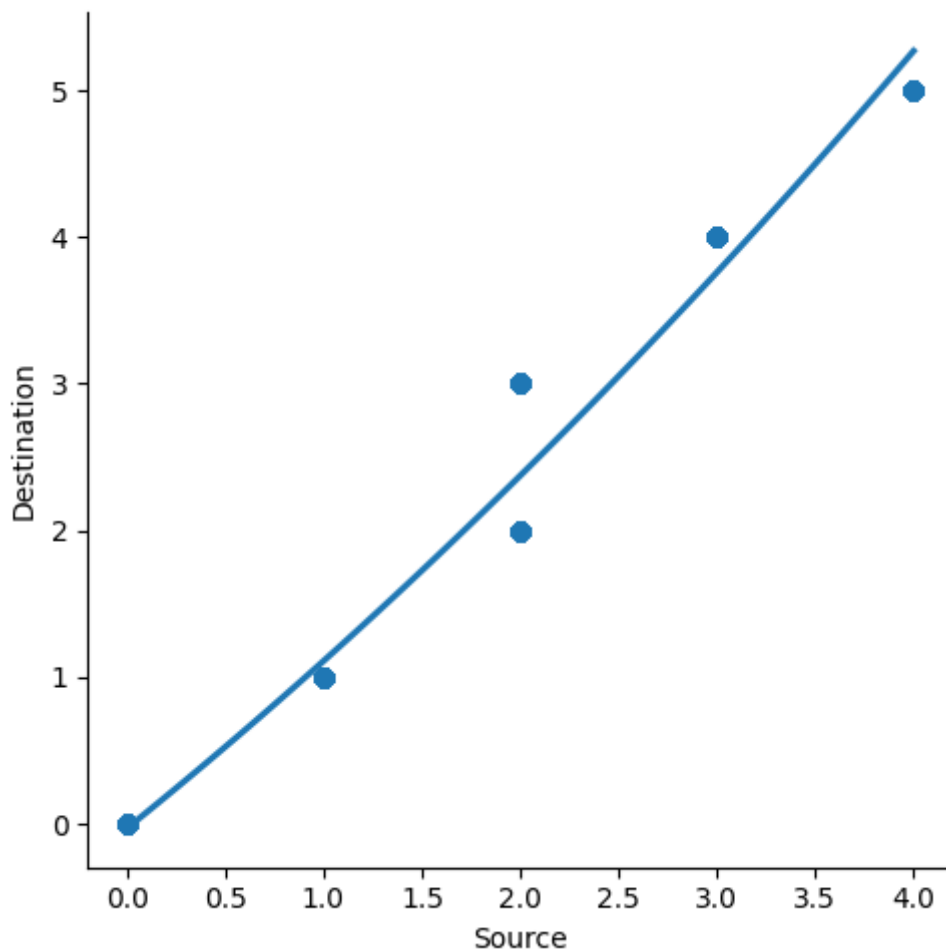


In [35]:

```
sns.lmplot(x="Source",y="Destination",data=train_df,order=2,ci=None)
```

Out[35]:

<seaborn.axisgrid.FacetGrid at 0x1b8f036c190>



In [36]:

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

0.9661997366087269

## ridge

In [37]:

```
from sklearn.linear_model import Ridge, RidgeCV, Lasso
```

In [38]:

```
ridge=Ridge(alpha=2)
ridge.fit(x_train,y_train)
train_score_ridge=ridge.score(x_train,y_train)
test_score_ridge=ridge.score(X_test,y_test)
print("\nLinearRegression\n")
print(train_score_ridge)
print(test_score_ridge)
```

LinearRegression

0.0004353748899809107  
0.04064037732261583

## Lasso

In [39]:

```
lasso=Lasso(alpha=100)
lasso=lasso.fit(x_train,y_train)
train_score_lasso=lasso.score(x_train,y_train)
test_score_lasso=lasso.score(X_test,y_test)
print(train_score_lasso)
print(test_score_lasso)
```

0.0  
-0.0002774640661185046

## elastic

In [40]:

```
from sklearn.linear_model import ElasticNet
```

In [41]:

```
a=ElasticNet()
a.fit(x,y)
print(a.coef_)
print(a.intercept_)
```

[0.6035366]  
[0.5919792]

In [42]:

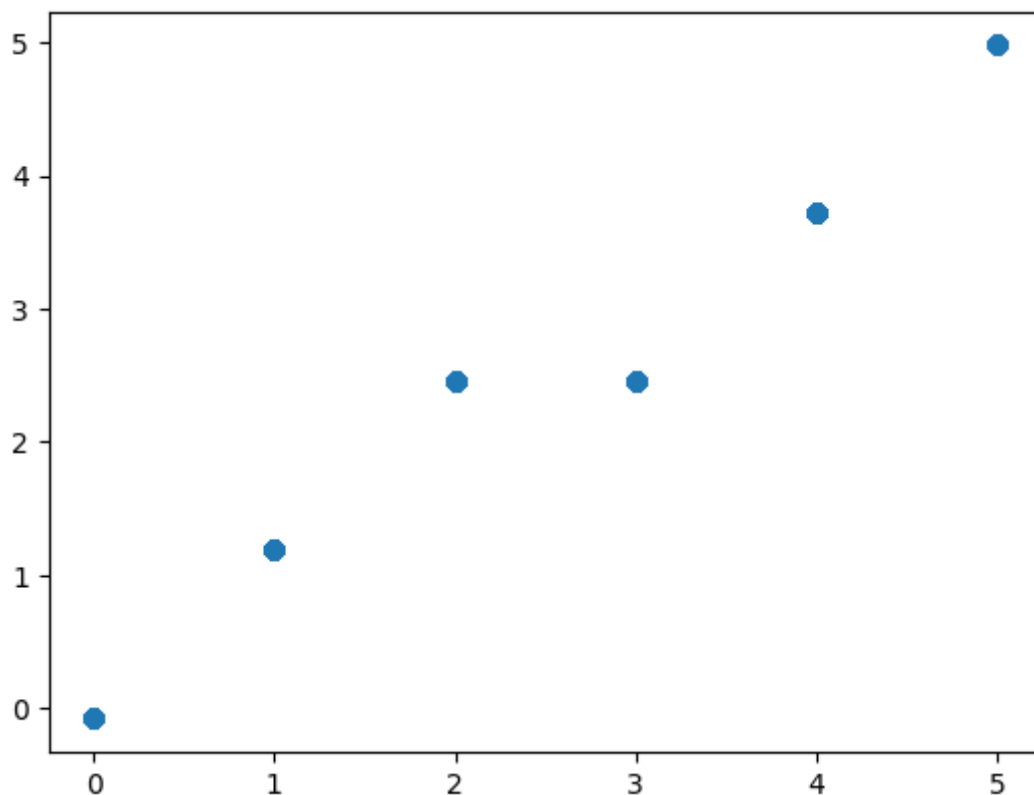
```
predictions=regr.predict(X_test)
```

In [43]:

```
plt.scatter(y_test,predictions)
```

Out[43]:

<matplotlib.collections.PathCollection at 0x1b8f4513f50>



## Logistic regression

In [44]:

```
x=np.array(train_df['Source']).reshape(-1,1)
y=np.array(train_df['Destination']).reshape(-1,1)
train_df.dropna(inplace=True)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=1)
lr=LogisticRegression(max_iter=10000)
```

In [45]:

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

C:\Users\91950\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\utils\validation.py:1143: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples, ), for example using ravel().

```
y = column_or_1d(y, warn=True)
```

Out[45]:

LogisticRegression
LogisticRegression(max_iter=10000)

In [46]:

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

0.9110764430577223

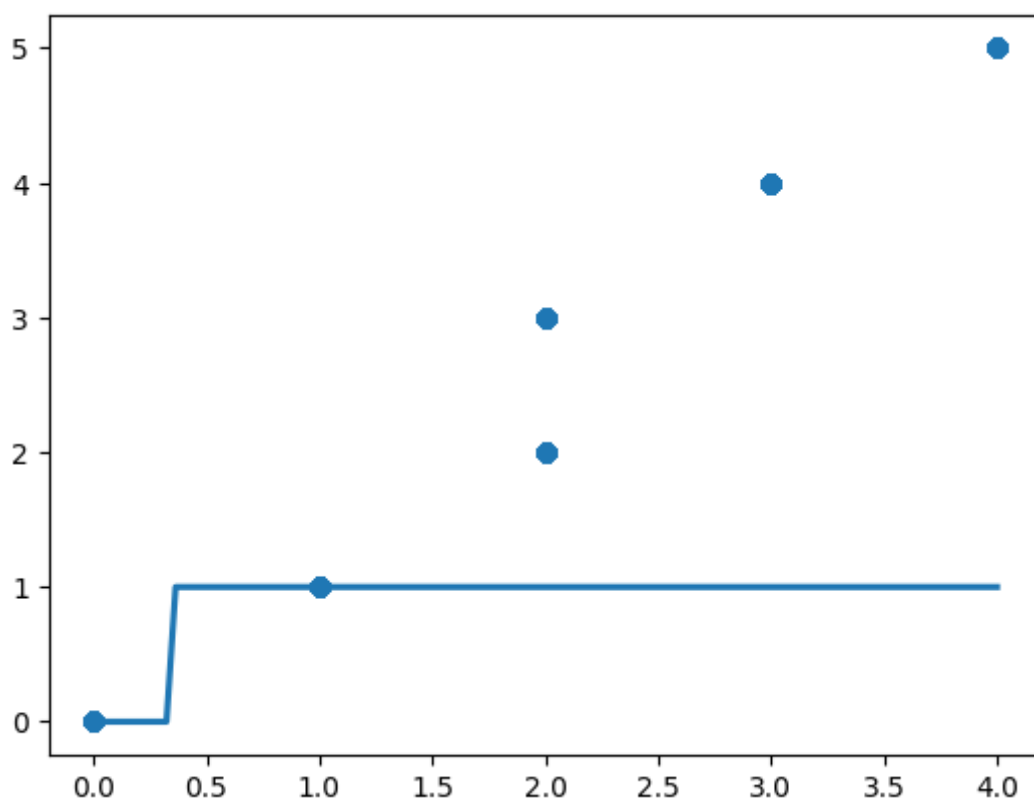
In [47]:

```
sns.regplot(x=x,y=y,data=train_df,logistic=True,ci=None)
```

C:\Users\91950\AppData\Local\Programs\Python\Python311\Lib\site-packages\satsmodels\genmod\family\links.py:198: RuntimeWarning: overflow encountered in exp  
 t = np.exp(-z)

Out[47]:

&lt;Axes: &gt;



## Decision tree

In [48]:

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

Out[48]:

```
DecisionTreeClassifier
DecisionTreeClassifier(random_state=0)
```

In [49]:

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

0.9110764430577223

## Random classifier

In [104]:

```
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(X_train,y_train)
print(rfc.score(x_train,y_train))
print(rfc.score(x_test,y_test))
```

0.4218269359368731

0.431201248049922

In [105]:

```
rf=RandomForestClassifier()
```

In [106]:

```
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 [107]:

```
import warnings
warnings.simplefilter(action='ignore')
```

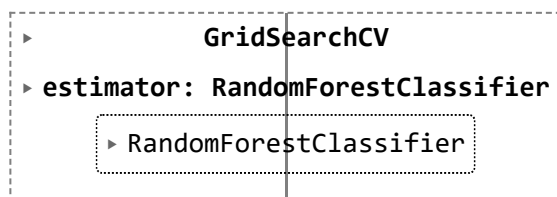
In [108]:

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

In [109]:

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

Out[109]:





In [110]:

```
grid_search.best_score_
```

Out[110]:

0.9134679490013939

In [111]:

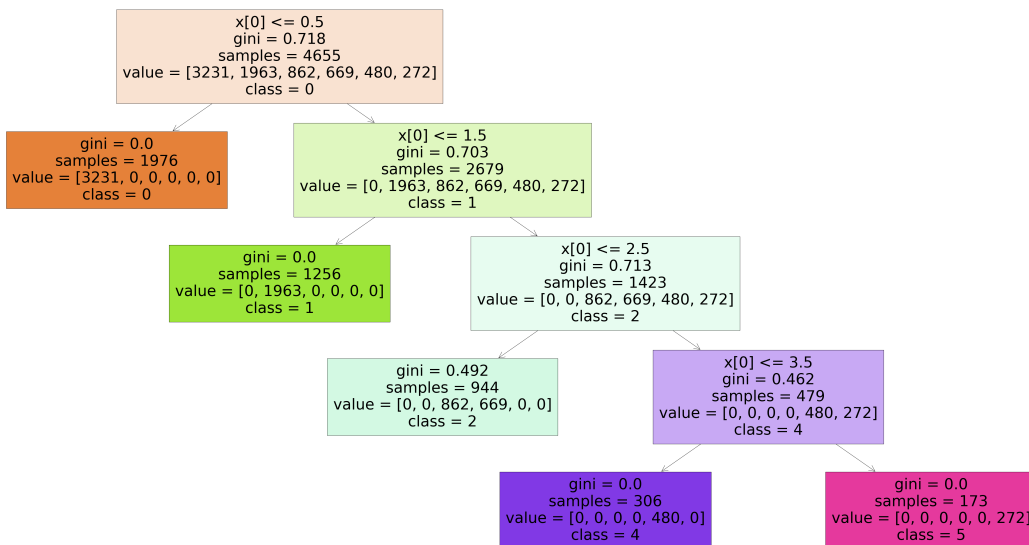
```
rf_best=grid_search.best_estimator_  
rf_best
```

Out[111]:

```
RandomForestClassifier  
RandomForestClassifier(max_depth=5, min_samples_leaf=5, n_estimators=10)
```

In [112]:

```
from sklearn.tree import plot_tree  
plt.figure(figsize=(80,40))  
plot_tree(rf_best.estimators_[6],class_names=['0','1','2','3','4','5','6'],filled=True);
```



In [113]:

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

0.431201248049922

## conclusion

For the above Dataset we use different Types of models, for each and every model we get different Types of Accuracies. For linear regression we obtained 96% accuracy, For Logistic regression we obtained 91% accuracy, For decision tree we obtained 91% accuracy, For Random forest we obtained 43% accuracy. From all the observations we can conclude that Linear Regression model is Best fit.

