

## MINI PROJECT - 2

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

```
In [2]: 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 [3]: train_df=pd.read_csv(r"C:\Users\LENOVO\Desktop\Data_Train11.csv")
train_df
```

Out[3]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	To
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	
...	...	...	...	...	...	...	...	...	
10678	Air Asia	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h 30m	
10679	Air India	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 35m	
10680	Jet Airways	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	3h	
10681	Vistara	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	
10682	Air India	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	

10683 rows × 11 columns



```
In [4]: test_df=pd.read_csv(r"C:\Users\LENOVO\Desktop\Test_set22.csv")
test_df
```

Out[4]:

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

2671 rows × 10 columns



## Data cleaning and preproceesing

```
In [5]: train_df.shape
```

Out[5]: (10683, 11)

```
In [6]: test_df.shape
```

```
Out[6]: (2671, 10)
```

```
In [7]: train_df.describe
```

```
Out[7]: <bound method NDFrame.describe of
estination
```

	Airline	Date_of_Journey	Source	D
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 [8]: test_df.describe
```

```
Out[8]: <bound method NDFrame.describe of
rce Destination
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      Bangalore      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]>
```

```
In [9]: train_df.head
```

```
Out[9]: <bound method NDFrame.head of
nation
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 [10]: test_df.head
```

```
Out[10]: <bound method NDFrame.head of
Destination
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      Bangalore      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]>
```

```
In [11]: 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 [12]: 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 [13]: train_df.isna().sum()
```

```
Out[13]: 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 [14]: test_df.isna().sum()
```

```
Out[14]: 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 [15]: train_df.duplicated().sum()
```

```
Out[15]: 220
```

```
In [16]: test_df.duplicated().sum()
```

```
Out[16]: 26
```

```
In [17]: train_df.dropna(inplace=True)
```

```
In [18]: train_df.isna().sum()
```

```
Out[18]: 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 [19]: train_df.shape
```

```
Out[19]: (10682, 11)
```

## Feature selection

```
In [20]: train_df.columns
```

```
Out[20]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',  
              'Dep_Time', 'Arrival_Time', 'Duration', 'Total_Stops',  
              'Additional_Info', 'Price'],  
              dtype='object')
```

```
In [21]: test_df.columns
```

```
Out[21]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',  
              'Dep_Time', 'Arrival_Time', 'Duration', 'Total_Stops',  
              'Additional_Info'],  
              dtype='object')
```

```
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  
Bangalore  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	Duration	Tot
0	1	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	
1	2	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	
2	0	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	
3	1	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	
4	1	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	
...	...	...	...	...	...	...	...	...	
10678	6	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h 30m	
10679	2	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 35m	
10680	0	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	3h	
10681	5	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	
10682	2	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	

10682 rows × 11 columns



```
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	Duration	Tota
0	1	24/03/2019	2	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	r
1	2	1/05/2019	1	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	
2	0	9/06/2019	0	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	
3	1	12/05/2019	1	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	
4	1	01/03/2019	2	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	
...	...	...	...	...	...	...	...	...	
10678	6	9/04/2019	1	Banglore	CCU ? BLR	19:55	22:25	2h 30m	r
10679	2	27/04/2019	1	Banglore	CCU ? BLR	20:45	23:20	2h 35m	r
10680	0	27/04/2019	2	Delhi	BLR ? DEL	08:20	11:20	3h	r
10681	5	01/03/2019	2	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	r
10682	2	9/05/2019	0	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	

10682 rows × 11 columns



```
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	Duration	Tota
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	r
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25m	
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45m	
...	...	...	...	...	...	...	...	...	
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30m	r
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35m	r
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	3h	r
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 40m	r
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	

10682 rows × 11 columns



```
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	Duration	Tota
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25m	
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45m	
...	...	...	...	...	...	...	...	...	
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30m	
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35m	
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 40m	
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	

10682 rows × 11 columns

visualization

```
In [30]: fdf=train_df[['Airline','Source','Destination','Total_Stops','Price']]
sns.heatmap(fdf.corr(),annot=True)
```

Out[30]: <Axes: >



## Linear regression for training data

```
In [31]: x=np.array(train_df['Source']).reshape(-1,1)
y=np.array(train_df['Destination']).reshape(-1,1)
```

```
In [32]: X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=100)
```

```
In [33]: from sklearn.linear_model import LinearRegression
regr=LinearRegression()
regr.fit(X_train,y_train)
print(regr.intercept_)
print(regr.coef_)
```

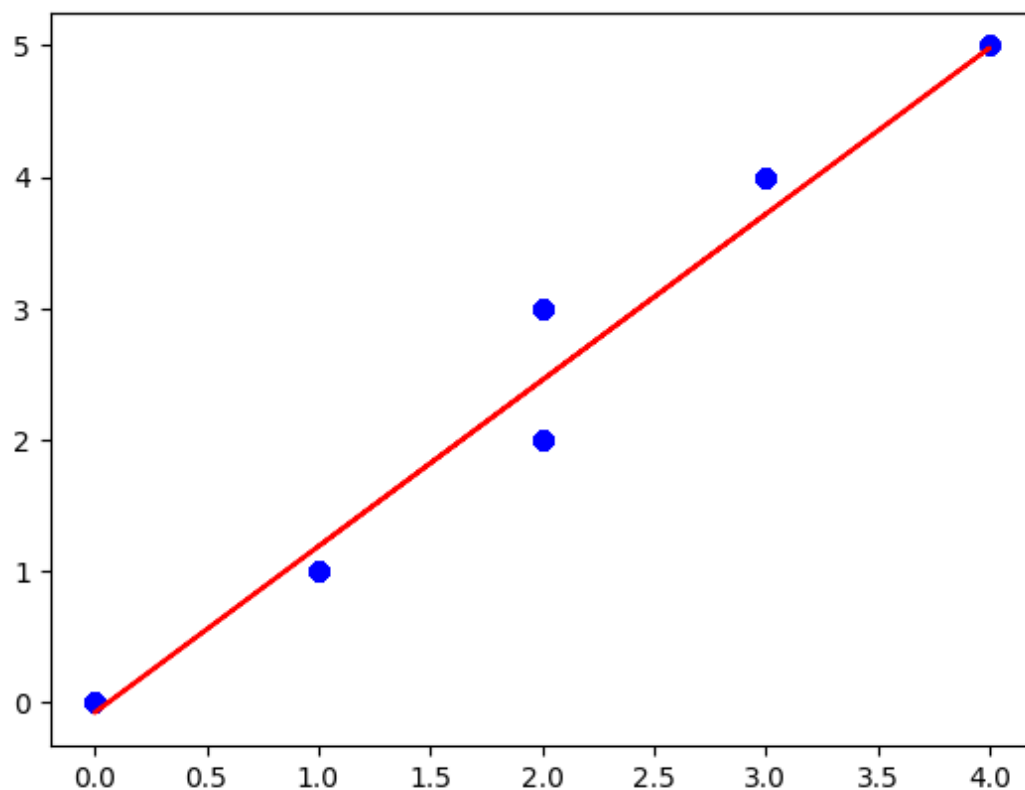
```
[-0.08091813]
[[1.26597749]]
```



```
In [90]: 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)
```

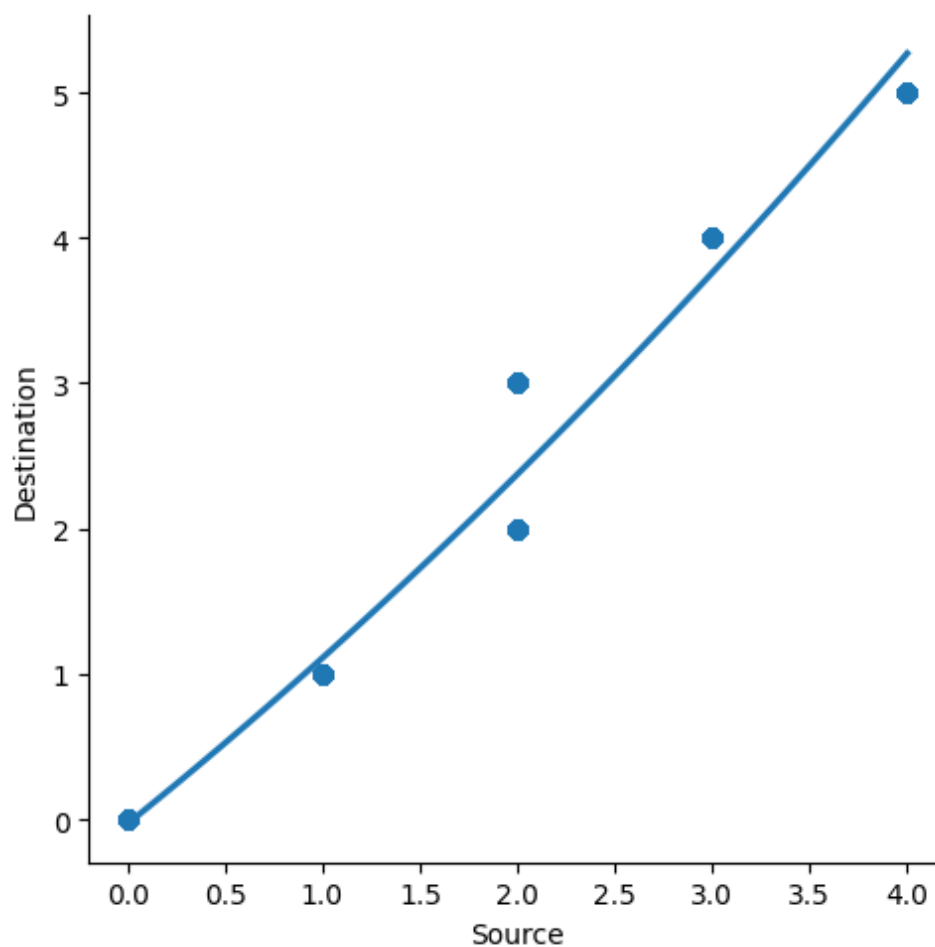
```
Out[90]: ▾ LinearRegression
LinearRegression()
```

```
In [93]: 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 [34]: sns.lmplot(x="Source",y="Destination",data=train_df,order=2,ci=None)
```

```
Out[34]: <seaborn.axisgrid.FacetGrid at 0x1a0bd009e10>
```



```
In [35]: score=regr.score(X_test,y_test)
print(score)
```

```
0.9676993120302174
```

## ridge

```
In [79]: from sklearn.linear_model import Ridge, RidgeCV, Lasso
```

```
In [80]: 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

0.9666541265004386  
0.9655420671601435

## Lasso

```
In [83]: 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  
-8.925310498675287e-06

## elastic

```
In [84]: from sklearn.linear_model import ElasticNet
```

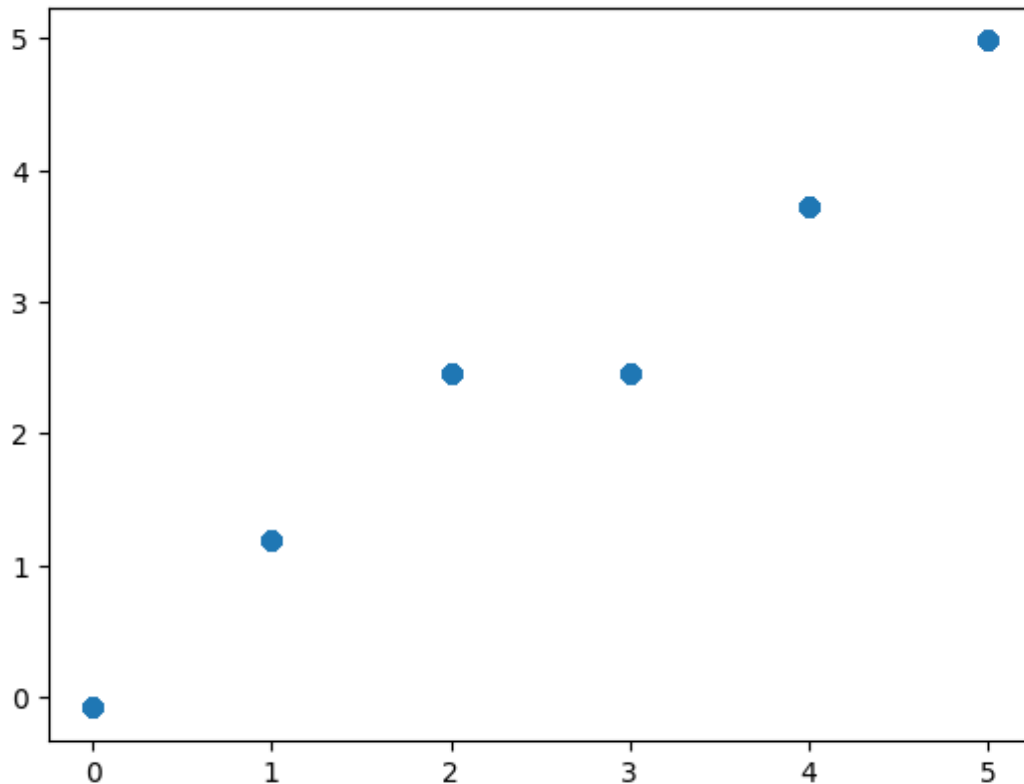
```
In [87]: a=ElasticNet()
a.fit(x,y)
print(a.coef_)
print(a.intercept_)
```

[0.6035366]  
[0.5919792]

```
In [89]: predictions=regr.predict(X_test)
```

```
In [50]: plt.scatter(y_test,predictions)
```

```
Out[50]: <matplotlib.collections.PathCollection at 0x1a0bcfee890>
```



## Logistic regression ¶

```
In [53]: 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 [54]: lr.fit(x_train,y_train)
```

C:\Users\LENOVO\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[54]: LogisticRegression
LogisticRegression(max_iter=10000)
```

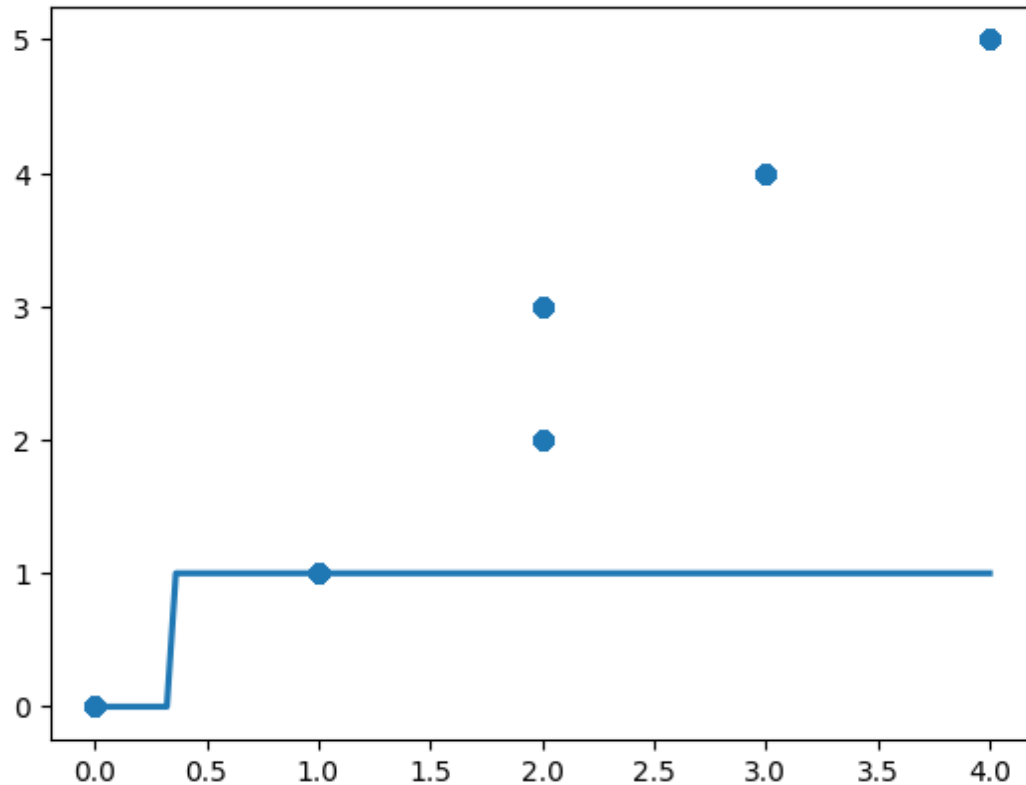
```
In [55]: score=lr.score(x_test,y_test)
print(score)
```

0.9110764430577223

```
In [59]: sns.regplot(x=x,y=y,data=train_df,logistic=True,ci=None)
```

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

Out[59]: <Axes: >



## Decision tree

```
In [60]: from sklearn.tree import DecisionTreeClassifier
train_df=DecisionTreeClassifier(random_state=0)
train_df.fit(x_train,y_train)
```

Out[60]:

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

```
In [61]: score=train_df.score(x_test,y_test)
print(score)
```

0.9110764430577223

# Random classifier

```
In [71]: from sklearn.ensemble import RandomForestClassifier
train_df=RandomForestClassifier()
train_df.fit(X_train,y_train)
```

C:\Users\LENOVO\AppData\Local\Temp\ipykernel\_6684\3479789618.py:3: 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().

```
train_df.fit(X_train,y_train)
```

```
Out[71]: ▾ RandomForestClassifier
RandomForestClassifier()
```

```
In [72]: 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 [73]: from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rfc,param_grid=params,cv=2,scoring="accuracy")
```

```
In [74]: grid_search.fit(X_train,y_train)
y was passed when a 1d array was expected. Please change the shape of y to (n_
samples,), for example using ravel().
estimator.fit(X_train, y_train, **fit_params)
C:\Users\LENOVO\AppData\Local\Programs\Python\Python311\Lib\site-packages\skle
arn\model_selection\_validation.py:686: 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().
estimator.fit(X_train, y_train, **fit_params)
C:\Users\LENOVO\AppData\Local\Programs\Python\Python311\Lib\site-packages\skle
arn\model_selection\_validation.py:686: 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().
estimator.fit(X_train, y_train, **fit_params)
C:\Users\LENOVO\AppData\Local\Programs\Python\Python311\Lib\site-packages\skle
arn\model_selection\_validation.py:686: DataConversionWarning: A column-vector
```

```
In [75]: grid_search.best_score_
```

```
Out[75]: 0.42182694348222594
```



**accuracy,For Random forest we obtained 42%  
accuracy,For decision tree we obtained 91%  
acuracy.From all the observations we can  
conclude that LinearRegression model is Best fit**

▪

In [ ]: