# In [1]:

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

# In [2]:

 $\label{traindf} traindf=pd.read\_csv(r"C:\Users\prajapath\ Arjun\OneDrive\Documents\Copy\ of\ Data\_Train.csv(runder) and the contract of the$ 

# In [3]:

traindf

# Out[3]:

	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 × 1	1 columns						
4								

# In [4]:

 $testdf=pd.read\_csv(r"C:\Users\prajapath\ Arjun\OneDrive\Documents\Copy\ of\ Test\_set.csv")\\ testdf$ 

# Out[4]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durat
0	Jet Airways	6/06/2019	De <b>l</b> hi	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	De <b>l</b> hi	Cochin	DEL → BOM → COK	19:15	19:00 22 May	23h 4
3	Multiple carriers	21/05/2019	De <b>l</b> hi	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	De <b>l</b> hi	Cochin	DEL → BOM → COK	21:50	04:25 07 Mar	6h 3
2669	Air India	6/03/2019	De <b>l</b> hi	Cochin	DEL → BOM → COK	04:00	19:15	15h 1
2670	Multiple carriers	15/06/2019	De <b>l</b> hi	Cochin	DEL → BOM → COK	04:55	19:15	14h 2

2671 rows × 10 columns

# In [5]:

# traindf.head()

# Out[5]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration
0	IndiGo	24/03/2019	Banglore	New De <b>l</b> hi	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	$\begin{array}{c} BLR \\ \to \\ NAG \\ \to \\ DEL \end{array}$	16:50	21:35	4h 45m
4								

# In [6]:

testdf.head()

# Out[6]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration
0	Jet Airways	6/06/2019	De <b>l</b> hi	Cochin	DEL → BOM → COK	17:30	04:25 07 Jun	10h 55m
1	IndiGo	12/05/2019	Kolkata	Banglore	$\begin{array}{c} CCU \\ \to \\ MAA \\ \to \\ BLR \end{array}$	06:20	10:20	4h
2	Jet Airways	21/05/2019	De <b>l</b> hi	Cochin	DEL → BOM → COK	19:15	19:00 22 May	23h 45m
3	Multiple carriers	21/05/2019	De <b>l</b> hi	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

# In [7]:

traindf.tail()

# Out[7]:

	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 [8]:

testdf.tail()

# Out[8]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duratic
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	De <b>l</b> hi	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 [9]:

traindf.describe()

# Out[9]:

	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 [10]:
```

```
testdf.describe()
```

# Out[10]:

		Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Dura
-	ount	2671	2671	2671	2671	2671	2671	2671	2
ur	nique	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 [11]:

```
traindf.shape
```

# Out[11]:

(10683, 11)

#### In [12]:

```
testdf.shape
```

# Out[12]:

(2671, 10)

# In [13]:

```
traindf.columns
```

#### Out[13]:

# In [14]:

```
testdf.columns
```

# Out[14]:

# In [15]:

```
traindf.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10683 entries, 0 to 10682
Data columns (total 11 columns):
# Column Non-Null Count

#	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
16	Price	10683 non-null	int64

dtypes: int64(1), object(10)
memory usage: 918.2+ KB

# In [16]:

testdf.info()

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

Duca	COTAMILE (COCAT T	0 0014111113/1	
#	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 [17]:
```

```
traindf.isnull().sum()
Out[17]:
Airline
                    0
Date_of_Journey
                    0
                    0
Source
Destination
                    0
                    1
Route
Dep_Time
                    0
                    0
Arrival Time
Duration
                    0
Total_Stops
                    1
Additional_Info
                    0
                    0
Price
dtype: int64
In [18]:
testdf.isnull().sum()
Out[18]:
Airline
                    0
Date_of_Journey
                    0
                    0
Source
Destination
                    0
                    0
Route
Dep Time
                    0
Arrival_Time
                    0
Duration
                    0
Total_Stops
                    0
Additional_Info
dtype: int64
In [19]:
traindf.dropna(inplace=True)
In [20]:
traindf.isnull().sum()
Out[20]:
Airline
                    0
Date_of_Journey
                    0
                    0
Source
Destination
                    0
Route
                    0
Dep_Time
                    0
                    0
Arrival_Time
Duration
                    0
Total_Stops
                    0
Additional_Info
                    0
                    0
Price
dtype: int64
```

```
In [21]:
```

```
traindf.shape
```

# Out[21]:

(10682, 11)

# In [22]:

```
traindf['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 1 Trujet

Name: count, dtype: int64

# In [23]:

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

# Out[23]:

Source

Delhi 4536 Kolkata 2871 Banglore 2197 Mumbai 697 Chennai 381

Name: count, dtype: int64

# In [24]:

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

# Out[24]:

Destination

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

Name: count, dtype: int64

# In [27]:

```
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}}
traindf=traindf.replace(airline)
traindf
```

# Out[27]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Durat
0	1	24/03/2019	Banglore	New De <b>l</b> hi	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
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 ∠
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	De <b>l</b> hi	BLR → DEL	08:20	11:20	
10681	5	01/03/2019	Banglore	New De <b>l</b> hi	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 x 1	1 columns						
100021								•

```
In [28]:
```

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

# Out[28]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duratio
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	09:25	04:25 10 Jun	1!
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
		9/04/2019		 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	;
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 2C
10682	rows × 1	1 columns						
4						_		

# In [29]:

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

# Out[29]:

DEL  CCU  IXR  1 2 1/05/2019 1 1 → 05:50 13:15 7h  BBI  BLR  DEL  DEL	13:15 7h 2			→ DEL CCU → IXR → BBI → BLR					0
1 2 1/05/2019 1 1 → 05:50 13:15 7h  BBI  BLR  DEL			05:50	→ IXR → BBI → BLR	1	1	1/05/2019	2	
$\rightarrow$	25 10 Jun 1	04:25 10 Jun		DEI			1700/2013	2	1
			09:25	→ LKO → BOM →	0	0	9/06/2019	0	2
CCU → 3 1 12/05/2019 1 1 NAG 18:05 23:30 5h → BLR	23:30 5h 29	23:30	18:05	$\begin{matrix} \rightarrow \\ NAG \\ \rightarrow \end{matrix}$	1	1	12/05/2019	1	3
BLR → 1 01/03/2019 2 3 NAG 16:50 21:35 4h DEL	21:35 4h 4	21:35	16:50	$\begin{matrix} \rightarrow \\ NAG \\ \rightarrow \end{matrix}$	3	2	01/03/2019	1	4
<b></b>									
10678 6 9/04/2019 1 1 $\xrightarrow{CCU}$ 19:55 22:25 2h BLR	22:25 2h 30	22:25	19:55	$\rightarrow$	1	1	9/04/2019	6	10678
10679 2 27/04/2019 1 1 $\rightarrow$ 20:45 23:20 2h BLR	23:20 2h 3	23:20	20:45	$\rightarrow$	1	1	27/04/2019	2	10679
10680 0 27/04/2019 2 BLR 2 → 08:20 11:20 DEL	11:20	11:20	08:20	$\rightarrow$	2	2	27/04/2019	0	10680
10681 5 01/03/2019 2 3 → 11:30 14:10 2h DEL	14:10 2h 40	14:10	11:30	$\rightarrow$	3	2	01/03/2019	5	10681
DEL → GOI 10682 2 9/05/2019 0 0 → 10:55 19:15 8h BOM → COK	19:15 8h 20	19:15	10:55	$\begin{array}{c} \rightarrow \\ GOI \\ \rightarrow \\ BOM \\ \rightarrow \end{array}$	0	0	9/05/2019	2	10682
10682 rows × 11 columns									

```
In [30]:
```

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

# Out[30]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duratio	
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	09:25	04:25 10 Jun	1!	
3	1	12/05/2019	1	1	$\begin{array}{c} CCU \\ \to \\ NAG \\ \to \\ BLR \end{array}$	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	;	
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									
1 Control of the Cont									

# In [31]:

traindf

# Out[31]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duratio
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	1!
3	1	12/05/2019	1	1	$\begin{array}{c} CCU \\ \to \\ NAG \\ \to \\ BLR \end{array}$	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	:
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 [32]:

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

# Out[32]:

#### <Axes: >



# In [33]:

```
x=fdf[['Airline','Source','Destination','Total_Stops']]
y=fdf['Price']
```

# In [34]:

```
#Linear Regression
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 [35]:

```
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
```

# 7211.098088897486

# Out[35]:

# Airline -418.483922 Source -3275.073380 Destination 2505.480291

3541.798053

# In [36]:

Total\_Stops

```
#Linear Rgeression
score=regr.score(X_test,y_test)
print(score)
```

#### 0.41083048909283504

# In [37]:

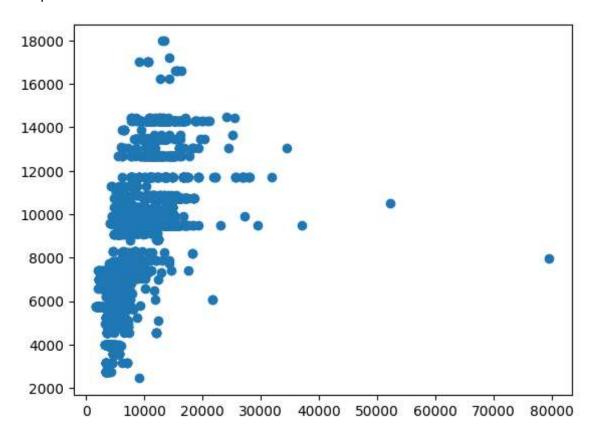
```
predictions=regr.predict(X_test)
```

#### In [38]:

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

#### Out[38]:

<matplotlib.collections.PathCollection at 0x1e9c74881d0>



#### In [39]:

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

C:\Users\prajapath Arjun\AppData\Local\Temp\ipykernel\_14740\521034954.py:

3: SettingWithCopyWarning:

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)

fdf.dropna(inplace=True)

# In [40]:

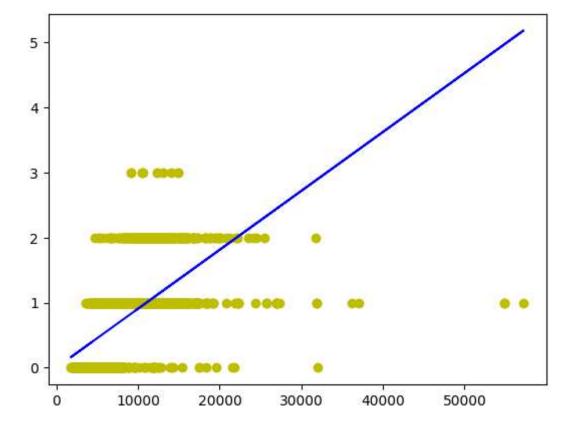
```
X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
regr.fit(X_train,y_train)
regr.fit(X_train,y_train)
```

# Out[40]:

```
LinearRegression
LinearRegression()
```

# In [41]:

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



# In [42]:

```
#Logistic Regression
x=np.array(fdf['Price']).reshape(-1,1)
y=np.array(fdf['Total_Stops']).reshape(-1,1)
fdf.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)
```

C:\Users\prajapath Arjun\AppData\Local\Temp\ipykernel\_14740\3604832714.py:

4: SettingWithCopyWarning:

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)

fdf.dropna(inplace=True)

#### In [43]:

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

C:\Users\prajapath Arjun\AppData\Local\Programs\Python\Python311\Lib\sitepackages\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[43]:

```
LogisticRegression
LogisticRegression(max_iter=10000)
```

#### In [44]:

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

#### 0.7160686427457098

# In [45]:

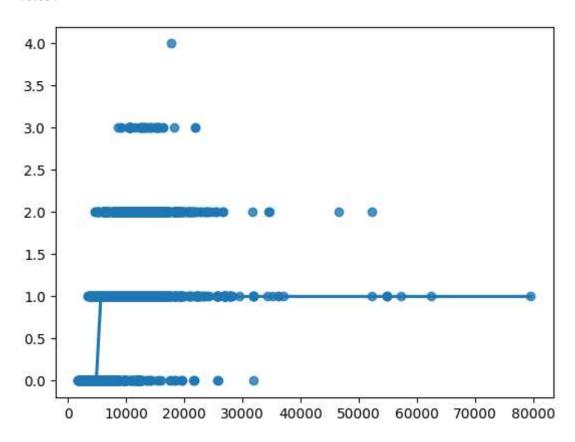
```
sns.regplot(x=x,y=y,data=fdf,logistic=True,ci=None)
```

C:\Users\prajapath Arjun\AppData\Local\Programs\Python\Python311\Lib\sitepackages\statsmodels\genmod\families\links.py:198: RuntimeWarning: overflo
w encountered in exp

t = np.exp(-z)

# Out[45]:

<Axes: >



# In [46]:

# #Decision tree

from sklearn.tree import DecisionTreeClassifier
clf=DecisionTreeClassifier(random\_state=0)
clf.fit(x\_train,y\_train)

# Out[46]:

DecisionTreeClassifier
DecisionTreeClassifier(random\_state=0)

#### In [47]:

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

#### 0.9369734789391576

# In [48]:

```
#Random forest classifier
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(X_train,y_train)
```

C:\Users\prajapath Arjun\AppData\Local\Temp\ipykernel\_14740\1232785509.py:
4: 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().
 rfc.fit(X\_train,y\_train)

#### Out[48]:

```
r RandomForestClassifier
RandomForestClassifier()
```

# In [49]:

```
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 [50]:

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

#### In [51]:

```
grid_search.fit(X_train,y_train)
arning. A column-vector y was passed when a la array was expected, ried
se change the shape of y to (n samples,), for example using ravel().
  estimator.fit(X train, y train, **fit params)
C:\Users\prajapath Arjun\AppData\Local\Programs\Python\Python311\Lib\si
te-packages\sklearn\model_selection\_validation.py:686: DataConversionW
arning: A column-vector y was passed when a 1d array was expected. Plea
se change the shape of y to (n samples,), for example using ravel().
  estimator.fit(X train, y train, **fit params)
C:\Users\prajapath Arjun\AppData\Local\Programs\Python\Python311\Lib\si
te-packages\sklearn\model selection\ validation.py:686: DataConversionW
arning: A column-vector y was passed when a 1d array was expected. Plea
se change the shape of y to (n samples,), for example using ravel().
  estimator.fit(X_train, y_train, **fit_params)
C:\Users\prajapath Arjun\AppData\Local\Programs\Python\Python311\Lib\si
te-packages\sklearn\model_selection\_validation.py:686: DataConversionW
arning: A column-vector y was passed when a 1d array was expected. Plea
se change the shape of y to (n_samples,), for example using ravel().
  estimator.fit(X_train, y_train, **fit_params)
C:\Users\prajapath Arjun\AppData\Local\Programs\Python\Python311\Lib\si
te-packages\sklearn\model_selection\_validation.py:686: DataConversionW
In [52]:
grid_search.best_score_
Out[52]:
0.523605715699528
In [53]:
rf_best=grid_search.best_estimator_
rf_best
Out[53]:
                           RandomForestClassifier
RandomForestClassifier(max depth=2, min samples leaf=50, n estimators=10)
```

# In [54]:

```
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);
```

```
x[0] \le 19551.5

gini = 0.597

samples = 4697

value = [2507, 3894, 1032, 42, 2]

class = 1
```

```
x[0] \le 3812.0

gini = 0.597

samples = 4619

value = [2434, 3840, 1020, 42, 2]

class = 1
```

```
gini = 0.566

samples = 78

value = [73, 54, 12, 0, 0]

class = 0
```

```
\begin{array}{c} \text{gini} = 0.582 \\ \text{samples} = 322 \\ \text{value} = [141, 294, 87, 0, 0] \\ \text{class} = 1 \end{array}
```

```
gini = 0.597
samples = 4297
value = [2293, 3546, 933, 42, 2]
class = 1
```

# In [55]:

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

#### 0.4424336973478939

#### In [ ]: