

# Mini Project on Flite

## 1.Problem Statement:which model is suitable best fit model for Flite

In [1]:

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

## Data Collection

## Read the Data

In [2]:

```
train_df=pd.read_csv(r"C:\Users\RAMADEVI SURIPAKA\OneDrive\Documents\train data.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\RAMADEVI SURIPAKA\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 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 preprocessing

In [4]:

```
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 [5]:

```
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 [6]:

```
train_df.describe()
```

Out[6]:

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

In [7]:

```
test_df.describe()
```

Out[7]:

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

In [8]:

```
train_df.columns
```

Out[8]:

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

In [9]:

```
test_df.columns
```

Out[9]:

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

In [10]:

train\_df.head()

Out[10]:

	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

In [11]:

test\_df.head()

Out[11]:

	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

In [12]:

```
train_df.shape
```

Out[12]:

```
(10683, 11)
```

In [13]:

```
test_df.shape
```

Out[13]:

```
(2671, 10)
```

## To find duplicate values

In [14]:

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

Out[14]:

```
220
```

In [15]:

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

Out[15]:

```
26
```

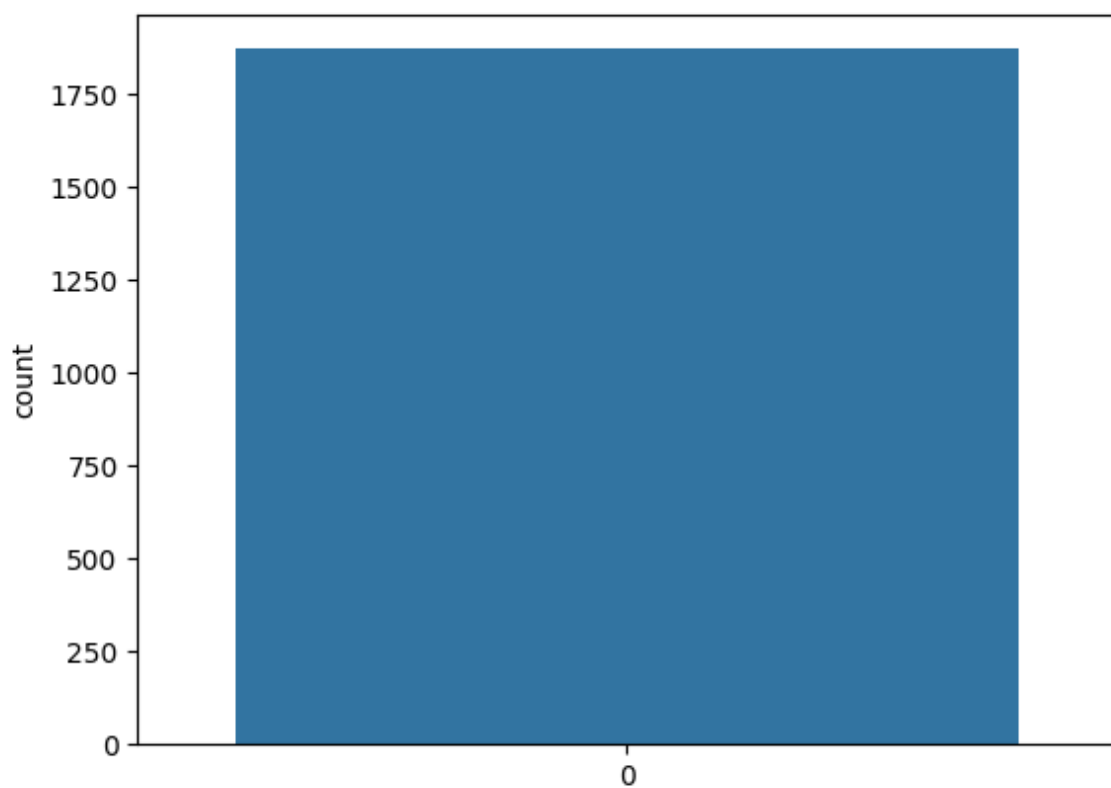
## 3.Data Visualization:Visualize the unique counts

In [16]:

```
sns.countplot(train_df['Price'].unique())
```

Out[16]:

<Axes: ylabel='count'>





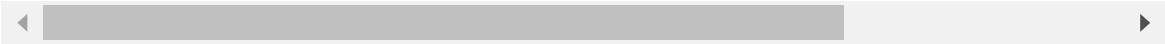
In [17]:

```
additionalinfo={"Additional_Info":{"No info":0,"In-flight meal not included":1,"No check  
test_df=test_df.replace(additionalinfo)  
test_df
```

Out[17]:

	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



find null values

Drop the column

In [ ]:

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

In [19]:

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

Out[19]:

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 [20]:

```
test_df.isnull().sum()
```

Out[20]:

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 [21]:

```
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[21]:

	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

10683 rows × 11 columns



In [22]:

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

Out[22]:

	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

10683 rows × 11 columns



In [23]:

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

Out[23]:

	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

10683 rows × 11 columns



In [24]:

```
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[24]:

	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

10683 rows × 11 columns



In [25]:

```
train_df
```

Out[25]:

	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	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

10683 rows × 11 columns



In [26]:

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

Out[26]:

<Axes: >



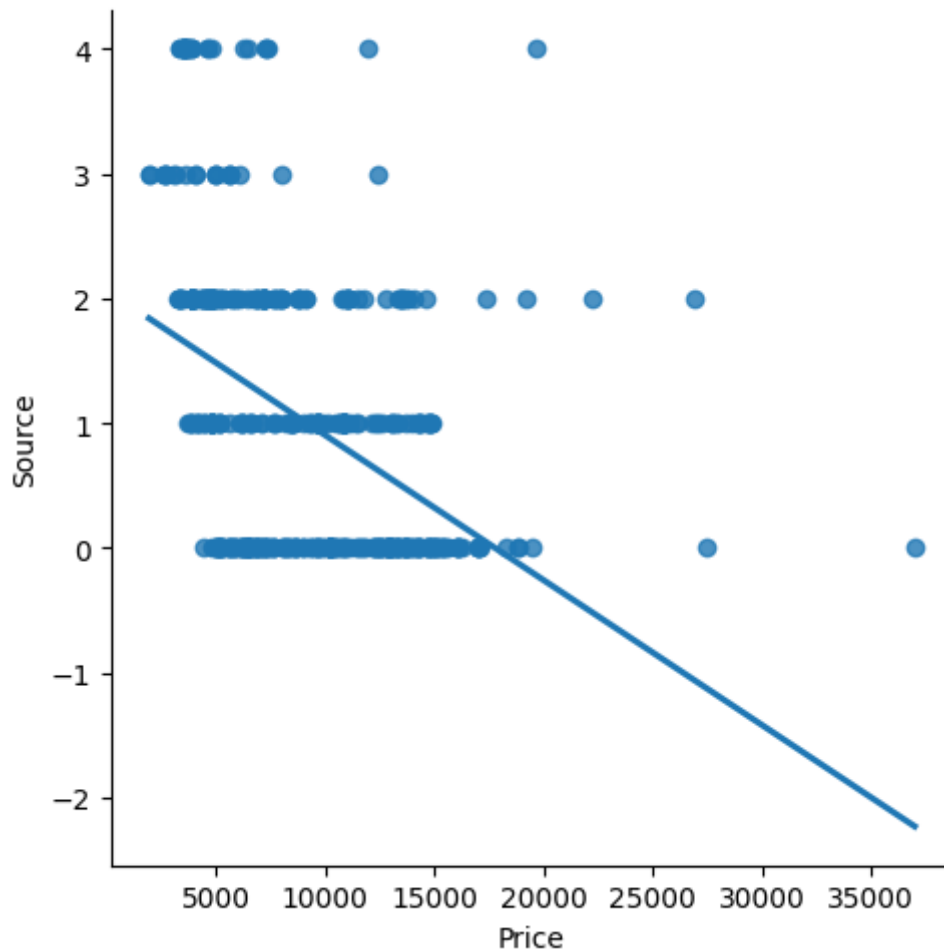


In [29]:

```
train_df500=train_df[:][:500]  
sns.lmplot(x="Price",y="Source",data=train_df500,order=1,ci=None)
```

Out[29]:

<seaborn.axisgrid.FacetGrid at 0x154bab89d20>

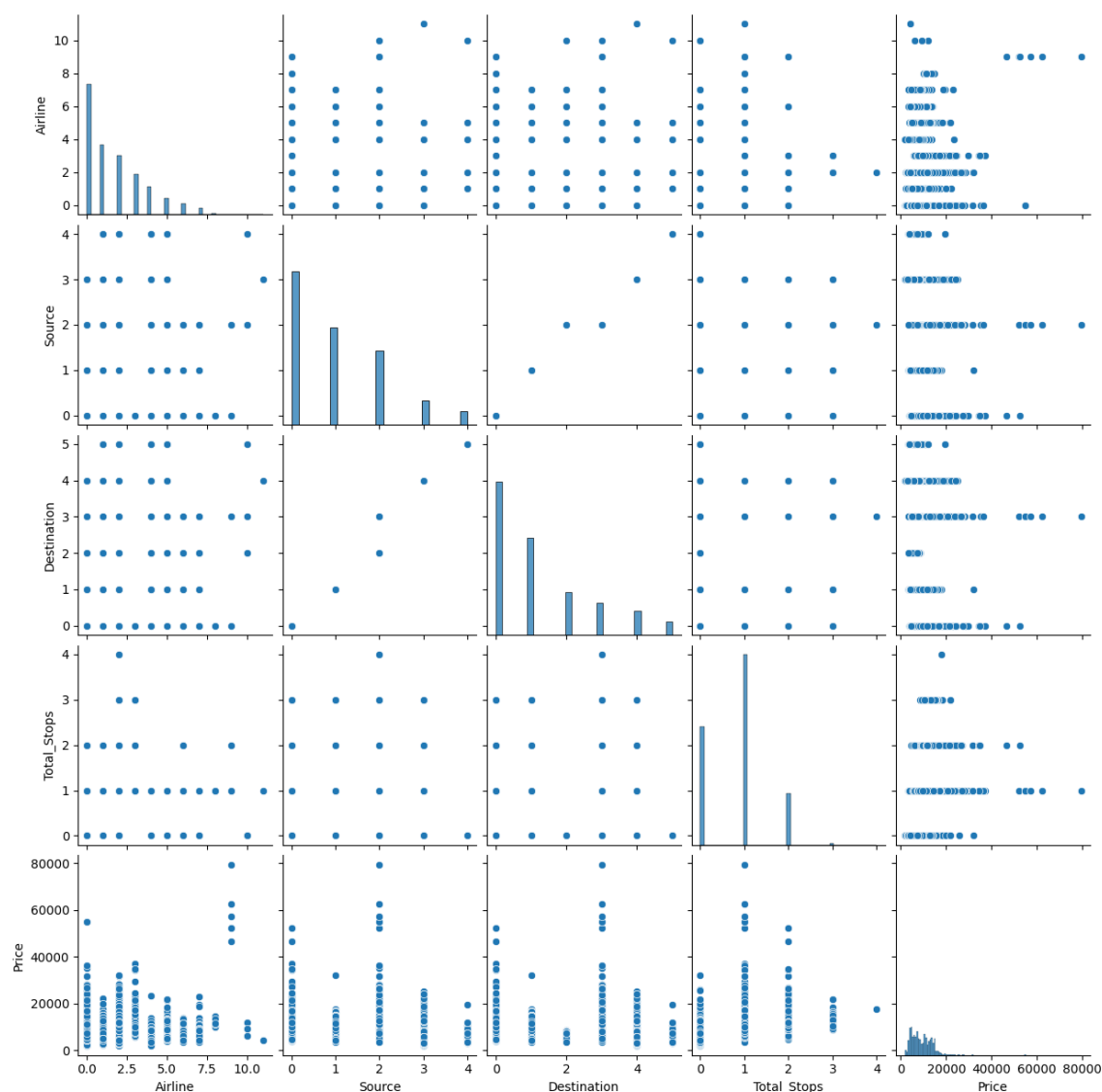


In [31]:

```
sns.pairplot(train_df)
```

Out[31]:

```
<seaborn.axisgrid.PairGrid at 0x154bddd9d0>
```



In [35]:

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

In [ ]:

## 4.Data Modeling:Using LinearRegression

In [55]:

```
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 [80]:

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

Out[80]:

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 [69]:

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

0.38389415387222314

In [70]:

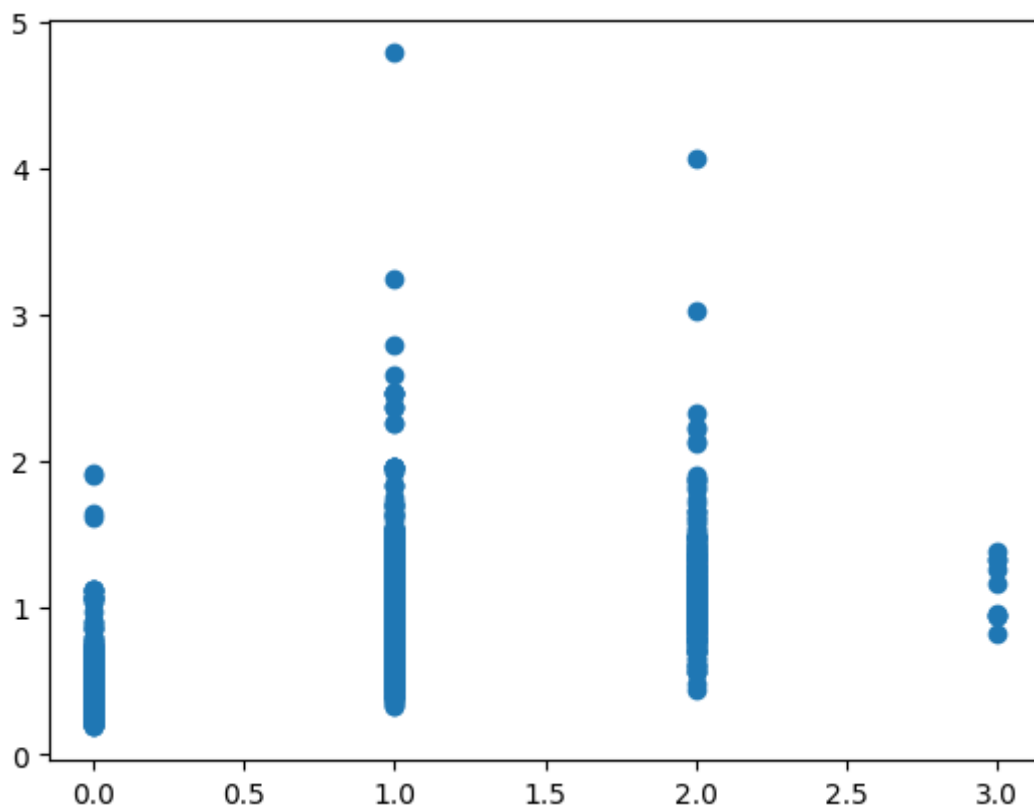
```
predictions=regr.predict(x_test)
```

In [71]:

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

Out[71]:

<matplotlib.collections.PathCollection at 0x154c347dc90>



In [76]:

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

C:\Users\RAMADEVI SURIPAKA\AppData\Local\Temp\ipykernel\_10164\143116723.p

y: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) ([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))

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

In [73]:

```
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[73]:

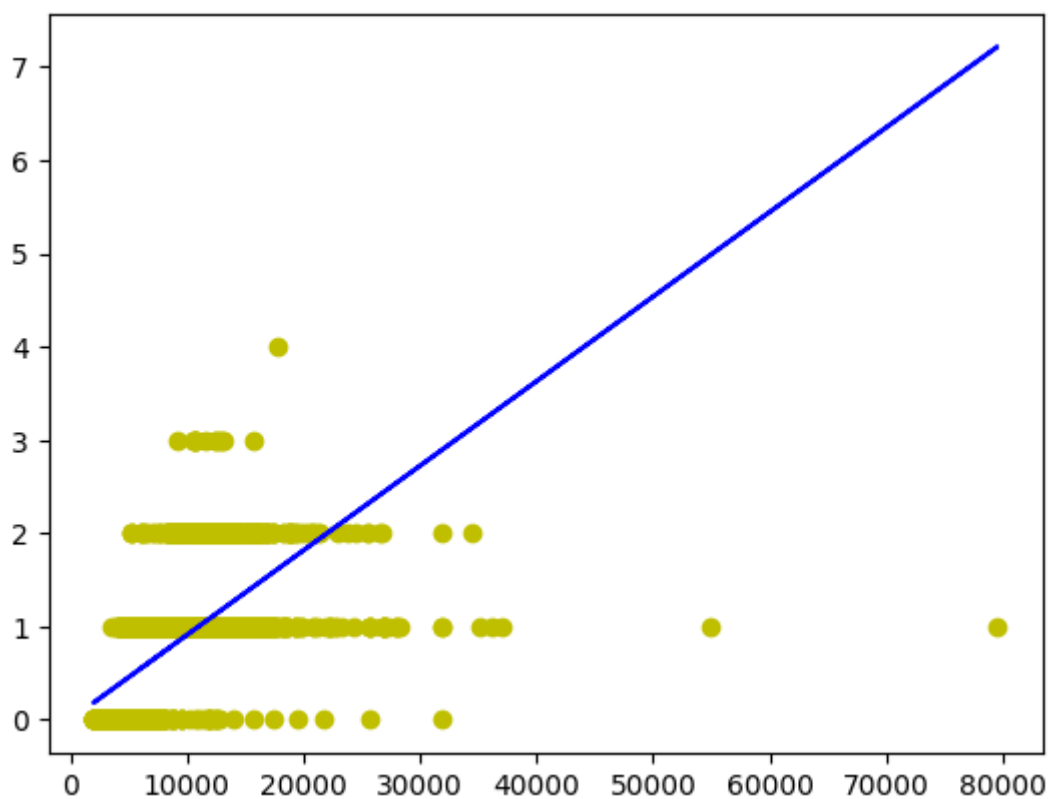
LinearRegression()

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 [74]:

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



## Logistic Regression

In [52]:

```
#Logistic Regression
x=np.array(train_df['Price']).reshape(-1,1)
y=np.array(train_df['Total_Stops']).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)
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression(max_iter=10000)
```

C:\Users\RAMADEVI SURIPAKA\AppData\Local\Temp\ipykernel\_10164\601506973.p

y: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) ([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))

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

In [77]:

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

C:\Users\RAMADEVI SURIPAKA\AppData\Local\Programs\Python\Python310\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[77]:

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

**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 [65]:

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

0.7160686427457098

In [67]:

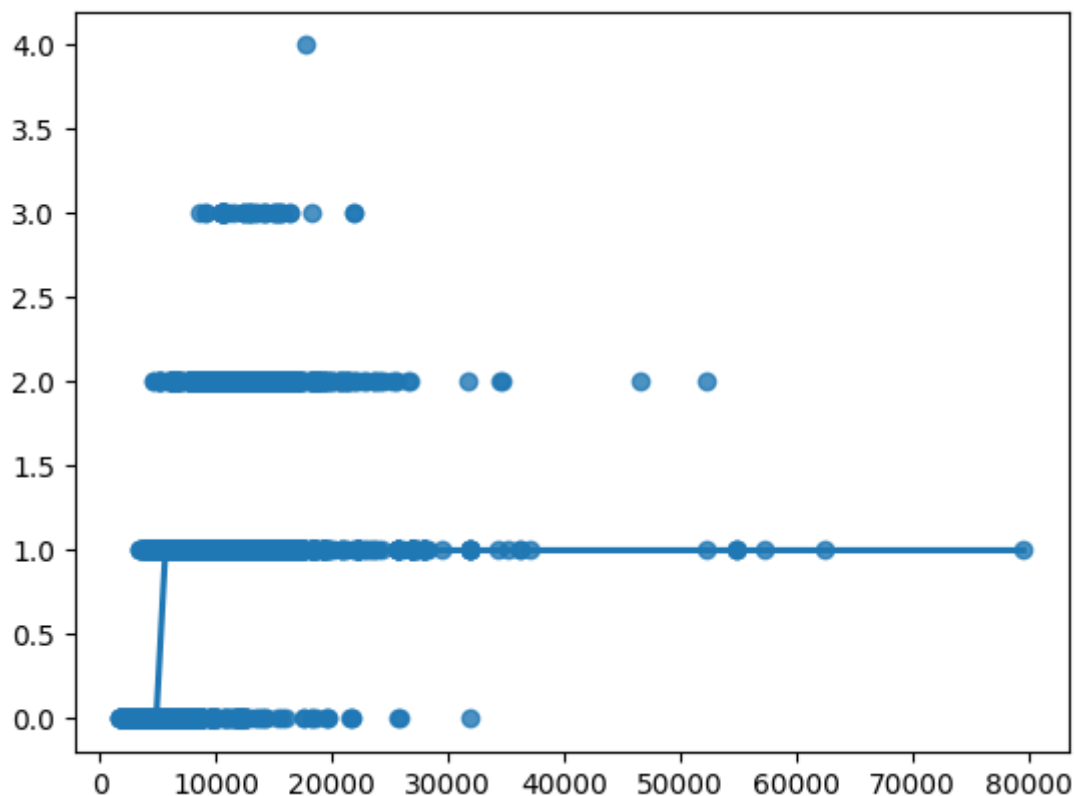
```
sns.regplot(x=x,y=y,data=train_df,logistic=True,ci=None)
```

C:\Users\RAMADEVI SURIPAKA\AppData\Local\Programs\Python\Python310\lib\site-packages\statsmodels\genmod\link.py:198: RuntimeWarning: overflow encountered in exp

```
t = np.exp(-z)
```

Out[67]:

&lt;Axes: &gt;



## Decision Tree

In [81]:

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

Out[81]:

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

In [82]:

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

0.4449297971918877

## Random Forest

In [83]:

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

C:\Users\RAMADEVI SURIPAKA\AppData\Local\Temp\ipykernel\_10164\4104924521.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().

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

Out[83]:

```
RandomForestClassifier()
RandomForestClassifier()
```

In [84]:

```
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 [85]:

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



In [86]:

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

```

lease change the shape of y to (n_samples,), for example using ravel().
    estimator.fit(X_train, y_train, **fit_params)
C:\Users\RAMADEVI SURIPAKA\AppData\Local\Programs\Python\Python310\lib
\site-packages\sklearn\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\RAMADEVI SURIPAKA\AppData\Local\Programs\Python\Python310\lib
\site-packages\sklearn\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\RAMADEVI SURIPAKA\AppData\Local\Programs\Python\Python310\lib
\site-packages\sklearn\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\RAMADEVI SURIPAKA\AppData\Local\Programs\Python\Python310\lib
\site-packages\sklearn\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)

```

In [87]:

```
grid_search.best_score_
```

Out[87]:

```
0.8790962138842513
```

In [88]:

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

Out[88]:

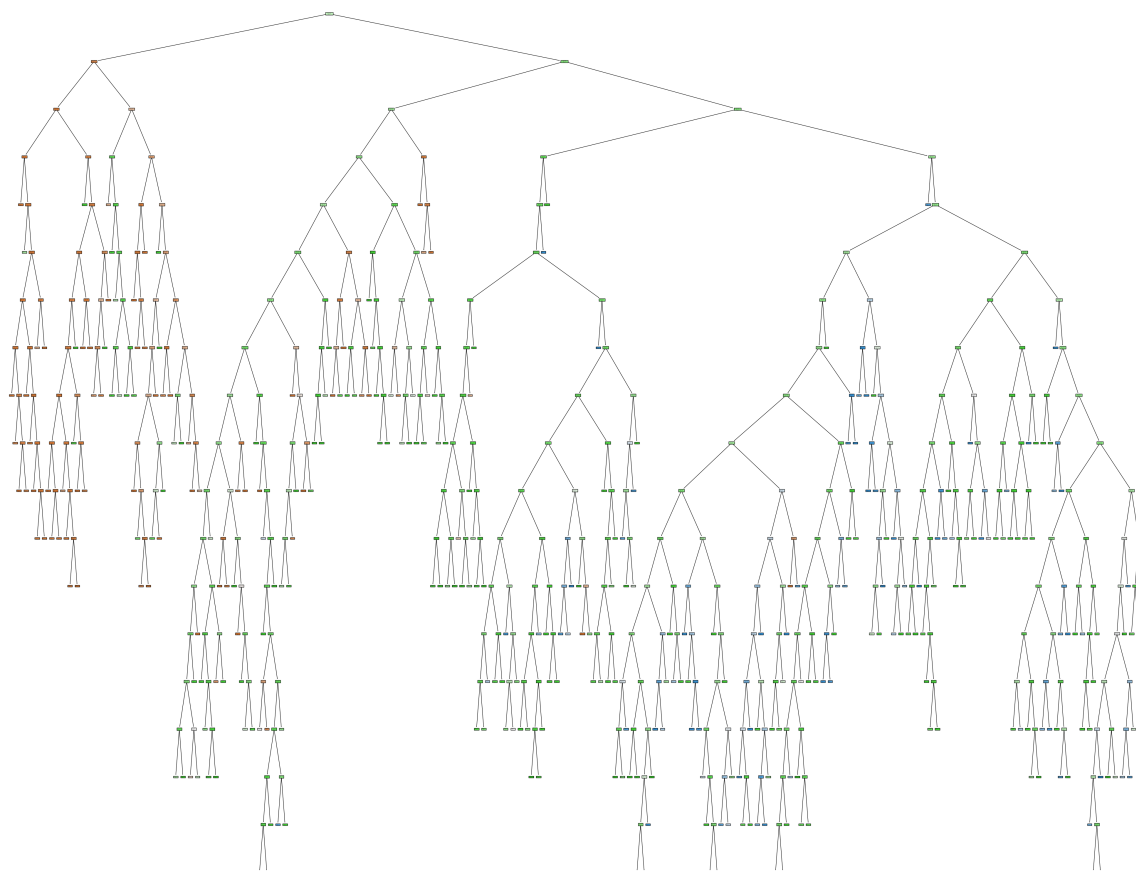
```

▼                                RandomForestClassifier
RandomForestClassifier(max_depth=20, min_samples_leaf=5, n_estimators=25)

```

In [96]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(50,40))
plot_tree(rf_best.estimators_[5],class_names=['0','1','2'],filled=True);
```



## 5.Prediction and Evaluation

In [93]:

```
prediction=lr.predict(X_test)
```

In [94]:

```
from sklearn.metrics import r2_score
model=LinearRegression()
model.fit(X_train,y_train)
y_pred=model.predict(X_test)
r2=r2_score(y_test,y_pred)
print("R2_score:",r2)
```

R2\_score: 0.3513581227540905

## Conclusion

Finallyn In the project I have done Lenear,Logistic,Decision Tree,Random Forest models.I got the different types of accuracys, I concluded that random Forest model was the best fit model for the Flite Price I got the accuracy value 87.

In [ ]: