

MINI PROJECT

PROBLEM STATEMENT :

model is suitable for Flight Price Prediction

Importing Packages

```
In [1]: ▶ import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns
```

Read The Data

```
In [2]: ▶ traindf=pd.read_csv(r"C:\Users\munigreeshma\OneDrive\Desktop\greeshma\Data_Train.csv")
traindf
```

Out[2]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	IndiGo	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	Air India	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops	No info	7662
2	Jet Airways	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop	No info	6218
4	IndiGo	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop	No info	13302
...
10678	Air Asia	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h 30m	non-stop	No info	4107
10679	Air India	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 35m	non-stop	No info	4145
10680	Jet Airways	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	3h	non-stop	No info	7229
10681	Vistara	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	non-stop	No info	12648
10682	Air India	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops	No info	11753

10683 rows × 11 columns

```
In [3]: testdf=pd.read_csv(r"C:\Users\munigreeshma\OneDrive\Desktop\greeshma\Test_set.csv")
testdf
```

Out[3]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
0	Jet Airways	6/06/2019	Delhi	Cochin	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h 55m	1 stop	No info
1	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? MAA ? BLR	06:20	10:20	4h	1 stop	No info
2	Jet Airways	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	19:15	19:00 22 May	23h 45m	1 stop	In-flight meal not included
3	Multiple carriers	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	08:00	21:00	13h	1 stop	No info
4	Air Asia	24/06/2019	Banglore	Delhi	BLR ? DEL	23:55	02:45 25 Jun	2h 50m	non-stop	No info
...
2666	Air India	6/06/2019	Kolkata	Banglore	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 55m	1 stop	No info
2667	IndiGo	27/03/2019	Kolkata	Banglore	CCU ? BLR	14:20	16:55	2h 35m	non-stop	No info
2668	Jet Airways	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 35m	1 stop	No info
2669	Air India	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	04:00	19:15	15h 15m	1 stop	No info
2670	Multiple carriers	15/06/2019	Delhi	Cochin	DEL ? BOM ? COK	04:55	19:15	14h 20m	1 stop	No info

2671 rows × 10 columns

In [4]: ▶ trairndf

Out[4]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	IndiGo	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	Air India	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops	No info	7662
2	Jet Airways	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop	No info	6218
4	IndiGo	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop	No info	13302
...
10678	Air Asia	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h 30m	non-stop	No info	4107
10679	Air India	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 35m	non-stop	No info	4145
10680	Jet Airways	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	3h	non-stop	No info	7229
10681	Vistara	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	non-stop	No info	12648
10682	Air India	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops	No info	11753

10683 rows × 11 columns

In [5]: ▶ testdf

Out[5]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
0	Jet Airways	6/06/2019	Delhi	Cochin	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h 55m	1 stop	No info
1	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? MAA ? BLR	06:20	10:20	4h	1 stop	No info
2	Jet Airways	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	19:15	19:00 22 May	23h 45m	1 stop	In-flight meal not included
3	Multiple carriers	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	08:00	21:00	13h	1 stop	No info
4	Air Asia	24/06/2019	Banglore	Delhi	BLR ? DEL	23:55	02:45 25 Jun	2h 50m	non-stop	No info
...
2666	Air India	6/06/2019	Kolkata	Banglore	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 55m	1 stop	No info
2667	IndiGo	27/03/2019	Kolkata	Banglore	CCU ? BLR	14:20	16:55	2h 35m	non-stop	No info
2668	Jet Airways	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 35m	1 stop	No info
2669	Air India	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	04:00	19:15	15h 15m	1 stop	No info
2670	Multiple carriers	15/06/2019	Delhi	Cochin	DEL ? BOM ? COK	04:55	19:15	14h 20m	1 stop	No info

2671 rows × 10 columns

DATA COLLECTION AND PREPROCESSING

```
In [6]: ▶ traindf.head()
```

Out[6]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	IndiGo	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	Air India	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops	No info	7662
2	Jet Airways	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop	No info	6218
4	IndiGo	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop	No info	13302

```
In [7]: ▶ testdf.head()
```

Out[7]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
0	Jet Airways	6/06/2019	Delhi	Cochin	DEL ? BOM ? COK	17:30	04:25 07 Jun	10h 55m	1 stop	No info
1	IndiGo	12/05/2019	Kolkata	Banglore	CCU ? MAA ? BLR	06:20	10:20	4h	1 stop	No info
2	Jet Airways	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	19:15	19:00 22 May	23h 45m	1 stop	In-flight meal not included
3	Multiple carriers	21/05/2019	Delhi	Cochin	DEL ? BOM ? COK	08:00	21:00	13h	1 stop	No info
4	Air Asia	24/06/2019	Banglore	Delhi	BLR ? DEL	23:55	02:45 25 Jun	2h 50m	non-stop	No info

```
In [8]: ▶ traindf.tail()
```

Out[8]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
10678	Air Asia	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h 30m	non-stop	No info	4107
10679	Air India	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 35m	non-stop	No info	4145
10680	Jet Airways	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	3h	non-stop	No info	7229
10681	Vistara	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	non-stop	No info	12648
10682	Air India	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops	No info	11753

```
In [9]: ▶ testdf.tail()
```

Out[9]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
2666	Air India	6/06/2019	Kolkata	Banglore	CCU ? DEL ? BLR	20:30	20:25 07 Jun	23h 55m	1 stop	No info
2667	IndiGo	27/03/2019	Kolkata	Banglore	CCU ? BLR	14:20	16:55	2h 35m	non-stop	No info
2668	Jet Airways	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	21:50	04:25 07 Mar	6h 35m	1 stop	No info
2669	Air India	6/03/2019	Delhi	Cochin	DEL ? BOM ? COK	04:00	19:15	15h 15m	1 stop	No info
2670	Multiple carriers	15/06/2019	Delhi	Cochin	DEL ? BOM ? COK	04:55	19:15	14h 20m	1 stop	No info

```
In [10]: ▶ traindf.describe()
```

Out[10]:

	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 [11]: ▶ testdf.describe()
```

Out[11]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info
count	2671	2671	2671	2671	2671	2671	2671	2671	2671	2671
unique	11	44	5	6	100	199	704	320	5	6
top	Jet Airways	9/05/2019	Delhi	Cochin	DEL ? BOM ? COK	10:00	19:00	2h 50m	1 stop	No info
freq	897	144	1145	1145	624	62	113	122	1431	2148

```
In [12]: ▶ traindf.shape
```

Out[12]: (10683, 11)

```
In [13]: ▶ testdf.shape
```

Out[13]: (2671, 10)


```
In [14]: ▶ traindf.columns
```

```
Out[14]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',  
              'Dep_Time', 'Arrival_Time', 'Duration', 'Total_Stops',  
              'Additional_Info', 'Price'],  
              dtype='object')
```

```
In [15]: ▶ testdf.columns
```

```
Out[15]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',  
              'Dep_Time', 'Arrival_Time', 'Duration', 'Total_Stops',  
              'Additional_Info'],  
              dtype='object')
```

```
In [16]: ▶ traindf.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 [17]: ► testdf.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
```

Checking Whether Any NULL Values Are Present Or Not

```
In [18]: ► traindf.isnull().sum()
```

```
Out[18]: 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 [19]: testdf.isnull().sum()
```

```
Out[19]: 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
```

Removing Null Values In Train Data

```
In [20]: traindf.dropna(inplace=True)
```

```
In [21]: traindf.isnull().sum()
```

```
Out[21]: 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 [22]: ▶ traindf.shape
```

```
Out[22]: (10682, 11)
```

Conversion Of Datatype Of Values From String To Numerical Values

```
In [23]: ▶ traindf['Airline'].value_counts()
```

```
Out[23]: 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 [24]: ▶ traindf['Source'].value_counts()
```

```
Out[24]: Source
Delhi      4536
Kolkata    2871
Bangalore  2197
Mumbai     697
Chennai    381
Name: count, dtype: int64
```

```
In [25]: ▶ traaindf['Destination'].value_counts()
```

```
Out[25]: Destination
Cochin      4536
Banglore    2871
Delhi       1265
New Delhi   932
Hyderabad   697
Kolkata     381
Name: count, dtype: int64
```

```
In [26]: ▶ traaindf['Total_Stops'].value_counts()
```

```
Out[26]: Total_Stops
1 stop      5625
non-stop    3491
2 stops     1520
3 stops      45
4 stops       1
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	Duration	Total_Stops	Additional_Info	Price
0	1	24/03/2019	Banglore	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	2	1/05/2019	Kolkata	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops	No info	7662
2	0	9/06/2019	Delhi	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	1	12/05/2019	Kolkata	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop	No info	6218
4	1	01/03/2019	Banglore	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop	No info	13302
...
10678	6	9/04/2019	Kolkata	Banglore	CCU ? BLR	19:55	22:25	2h 30m	non-stop	No info	4107
10679	2	27/04/2019	Kolkata	Banglore	CCU ? BLR	20:45	23:20	2h 35m	non-stop	No info	4145
10680	0	27/04/2019	Banglore	Delhi	BLR ? DEL	08:20	11:20	3h	non-stop	No info	7229
10681	5	01/03/2019	Banglore	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	non-stop	No info	12648
10682	2	9/05/2019	Delhi	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops	No info	11753

10682 rows × 11 columns

```
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	Duration	Total_Stops	Additional_Info	Price
0	1	24/03/2019	2	New Delhi	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	2	1/05/2019	1	Banglore	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops	No info	7662
2	0	9/06/2019	0	Cochin	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	1	12/05/2019	1	Banglore	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop	No info	6218
4	1	01/03/2019	2	New Delhi	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop	No info	13302
...
10678	6	9/04/2019	1	Banglore	CCU ? BLR	19:55	22:25	2h 30m	non-stop	No info	4107
10679	2	27/04/2019	1	Banglore	CCU ? BLR	20:45	23:20	2h 35m	non-stop	No info	4145
10680	0	27/04/2019	2	Delhi	BLR ? DEL	08:20	11:20	3h	non-stop	No info	7229
10681	5	01/03/2019	2	New Delhi	BLR ? DEL	11:30	14:10	2h 40m	non-stop	No info	12648
10682	2	9/05/2019	0	Cochin	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops	No info	11753

10682 rows × 11 columns

```
In [29]: destination={"Destination":{"Cochin":0,"Banglore":1,"Delhi":2,
    "New Delhi":3,"Hyderabad":4,"Kolkata":5}}
    traindf=traindf.replace(destination)
    traindf
```

Out[29]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2 stops	No info	7662
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1 stop	No info	6218
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1 stop	No info	13302
...
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30m	non-stop	No info	4107
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35m	non-stop	No info	4145
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	3h	non-stop	No info	7229
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 40m	non-stop	No info	12648
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2 stops	No info	11753

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	Duration	Total_Stops	Additional_Info	Price
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	0	No info	3897
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2	No info	7662
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2	No info	13882
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1	No info	6218
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1	No info	13302
...
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30m	0	No info	4107
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35m	0	No info	4145
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	3h	0	No info	7229
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 40m	0	No info	12648
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2	No info	11753

10682 rows × 11 columns

In [31]: ▶ traindf

Out[31]:

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	1	24/03/2019	2	3	BLR ? DEL	22:20	01:10 22 Mar	2h 50m	0	No info	3897
1	2	1/05/2019	1	1	CCU ? IXR ? BBI ? BLR	05:50	13:15	7h 25m	2	No info	7662
2	0	9/06/2019	0	0	DEL ? LKO ? BOM ? COK	09:25	04:25 10 Jun	19h	2	No info	13882
3	1	12/05/2019	1	1	CCU ? NAG ? BLR	18:05	23:30	5h 25m	1	No info	6218
4	1	01/03/2019	2	3	BLR ? NAG ? DEL	16:50	21:35	4h 45m	1	No info	13302
...
10678	6	9/04/2019	1	1	CCU ? BLR	19:55	22:25	2h 30m	0	No info	4107
10679	2	27/04/2019	1	1	CCU ? BLR	20:45	23:20	2h 35m	0	No info	4145
10680	0	27/04/2019	2	2	BLR ? DEL	08:20	11:20	3h	0	No info	7229
10681	5	01/03/2019	2	3	BLR ? DEL	11:30	14:10	2h 40m	0	No info	12648
10682	2	9/05/2019	0	0	DEL ? GOI ? BOM ? COK	10:55	19:15	8h 20m	2	No info	11753


10682 rows × 11 columns

DATA VISUALIZATION

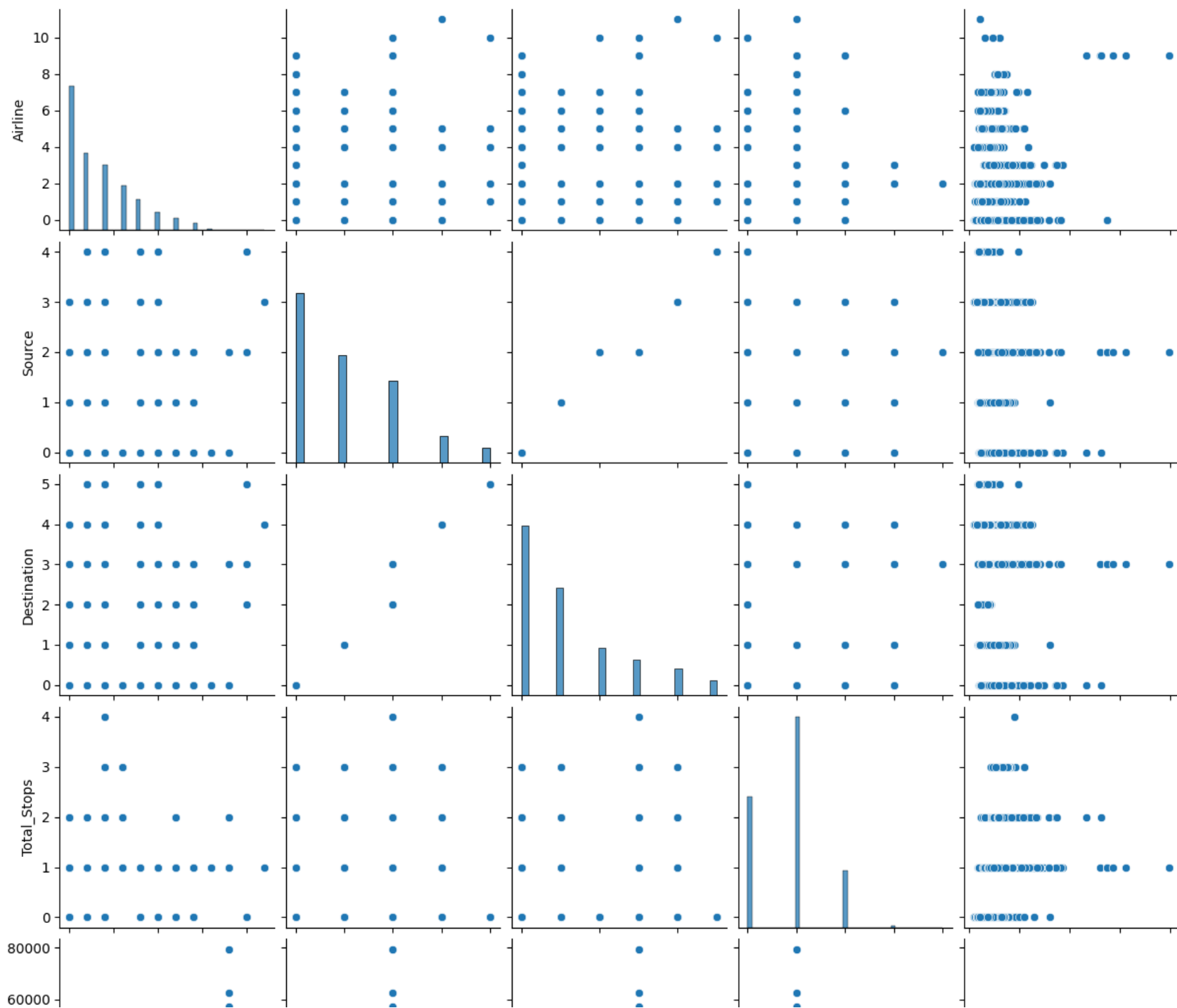
```
In [32]: ▶ import seaborn as sns
df=traindf[['Airline','Source','Destination','Total_Stops','Price']]
sns.heatmap(df.corr(),annot=True)
```

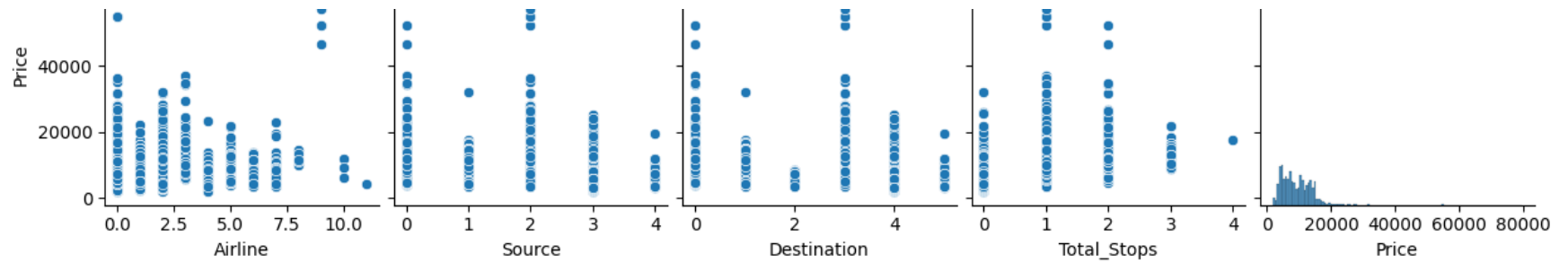
Out[32]: <Axes: >



In [33]:  sns.pairplot(df)

Out[33]: <seaborn.axisgrid.PairGrid at 0x248184a1190>





DATA MODELLING

LINEAR REGRESSION

```
In [34]: ▶ import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [35]: ▶ 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 [36]: ► #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
```

7211.098088897471

Out[36]:

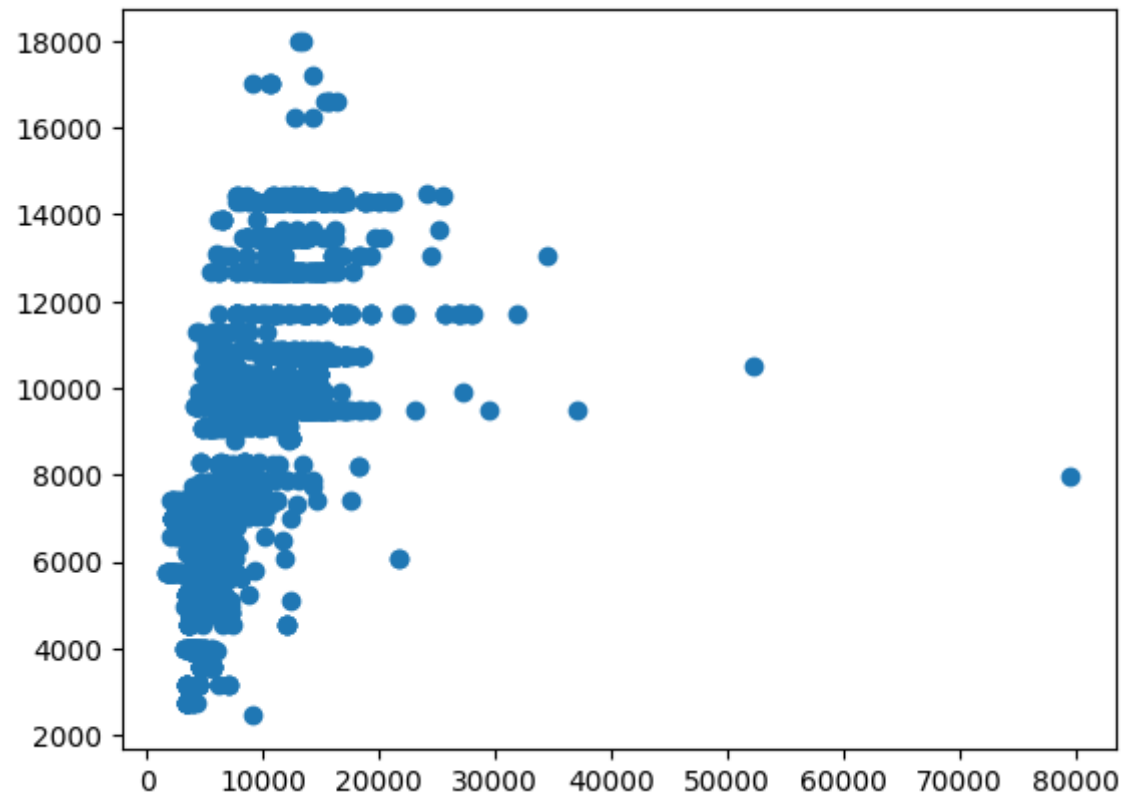
	coefficient
Airline	-418.483922
Source	-3275.073380
Destination	2505.480291
Total_Stops	3541.798053

```
In [37]: ► score=regr.score(X_test,y_test)
print(score)
```

0.41083048909283415


```
In [38]: ► predictions=regr.predict(X_test)  
plt.scatter(y_test,predictions)
```

```
Out[38]: <matplotlib.collections.PathCollection at 0x2481a668d90>
```



```
In [39]: ▶ x=np.array(df['Price']).reshape(-1,1)
y=np.array(df['Total_Stops']).reshape(-1,1)
df.dropna(inplace=True)
```

C:\Users\munigreeshma\AppData\Local\Temp\ipykernel_38160\3039801757.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)
df.dropna(inplace=True)

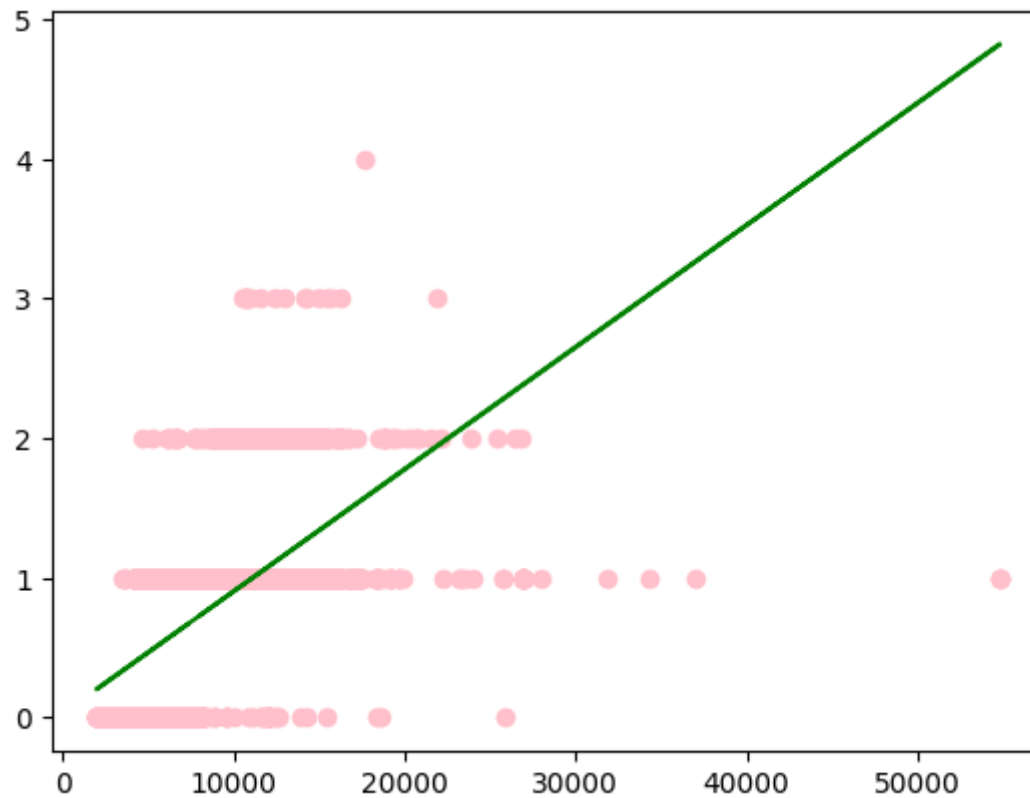
```
In [40]: ▶ X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
regr.fit(X_train,y_train)
regr.fit(X_train,y_train)
```

Out[40]:

▼ LinearRegression

LinearRegression()

```
In [41]: ► y_pred=regr.predict(X_test)
plt.scatter(X_test,y_test,color='pink')
plt.plot(X_test,y_pred,color='g')
plt.show()
```



LOGISTIC REGRESSION

```
In [42]: x=np.array(df['Price']).reshape(-1,1)
y=np.array(df['Destination']).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\munigreeshma\AppData\Local\Temp\ipykernel_38160\1131727007.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)
df.dropna(inplace=True)

```
In [43]: lr.fit(x_train,y_train)
```

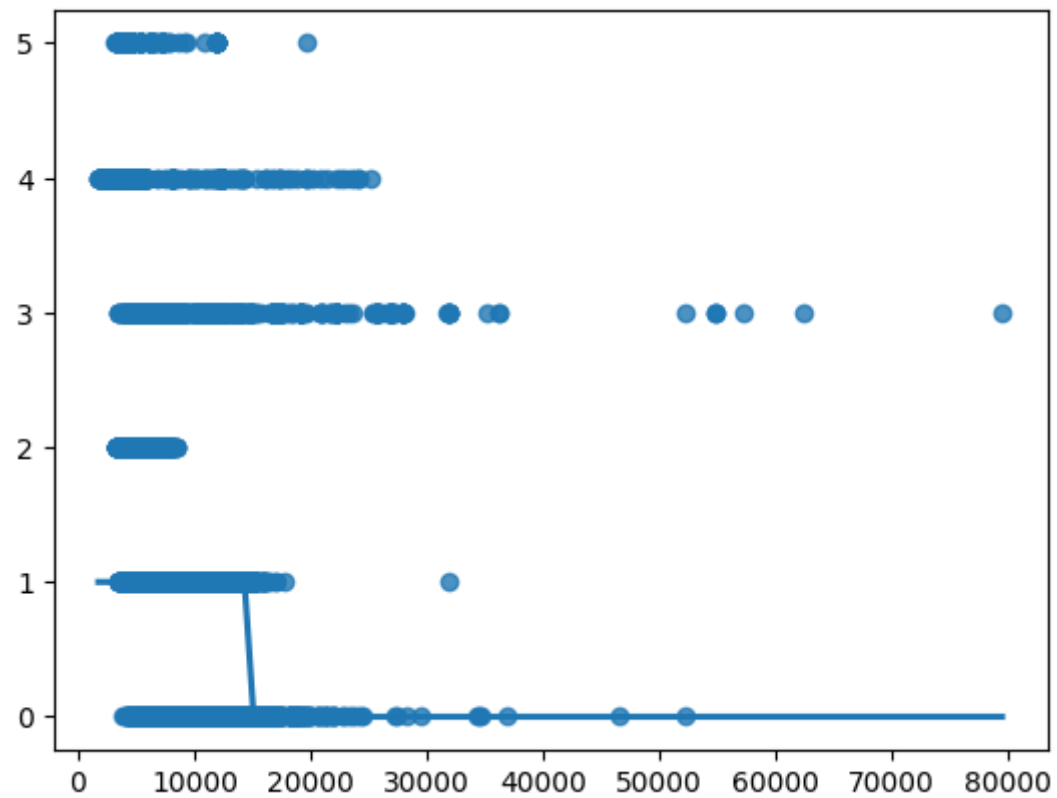
```
Out[43]: LogisticRegression
LogisticRegression(max_iter=10000)
```

```
In [44]: score=lr.score(x_test,y_test)
print(score)
```

0.431201248049922

```
In [45]: sns.regplot(x=x,y=y,data=df,logistic=True,ci=None)
```

Out[45]: <Axes: >



DECISION TREE


```
In [50]: ➤ 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[50]:  ▸      GridSearchCV
          ▸ estimator: RandomForestClassifier
            ▸ RandomForestClassifier
```

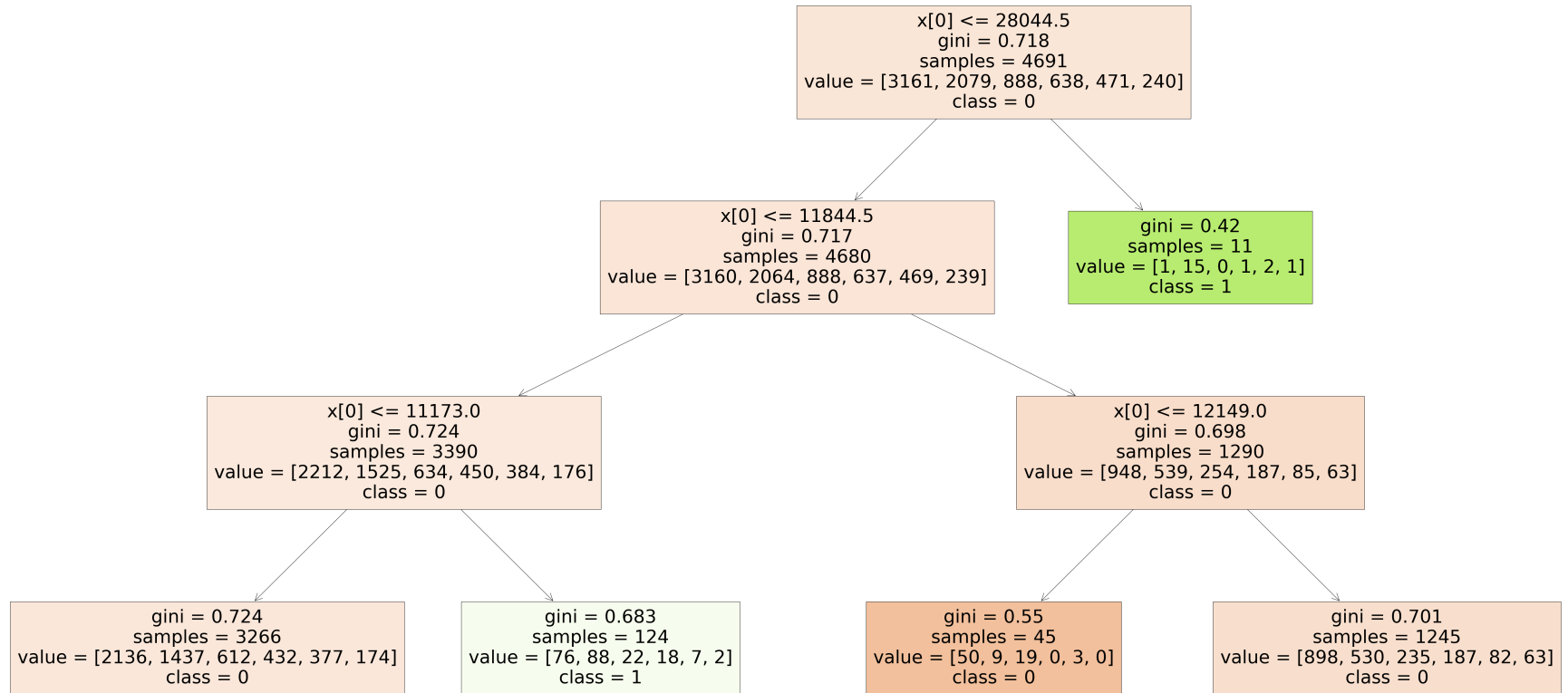
```
In [51]: ➤ grid_search.best_score_
```

```
Out[51]: 0.42343186527099785
```

```
In [52]: ➤ rf_best=grid_search.best_estimator_
rf_best
```

```
Out[52]: ▾      RandomForestClassifier
          RandomForestClassifier(max_depth=3, min_samples_leaf=10, n_estimators=25)
```

```
In [53]: ▶ 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 [54]: ▶ score=rfc.score(x_test,y_test)
print(score)
```

0.33510140405616223

CONCLUSION

In the Flight Price Prediction dataset project ,we have to find the best model after applying linear regression,logistic regression,decision tree and random forest we have observed that the DECISION TREE is the best model because the dataset i got is 93%accuracy.In Linear Regression i got 41% accuracy,In Logistic Regression i got 43% accuracy,In Random Forest i got 33% accuracy. By implementing all models on the dataset, i conclude that DECISION TREE is the best ,model for Flight Price Prediction dataset,because it got high accuracy by comparing all otherws models.