

In [1]:

```

1 #impoting libraries
2 import pandas as pd
3 import numpy as np
4 import matplotlib.pyplot as plt
5 import seaborn as sns

```

In [2]:

```

1 #Loading data and viewing null values if any
2 train_data = pd.read_excel("D:/Airline Tickets/Data_Train.xlsx")
3 print(train_data.isnull().sum())
4 print(train_data[train_data.isna().any(axis=1)])

```

```

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

```

	Airline	Date_of_Journey	Source	Destination	Route	Dep_Time	\
9039	Air India	6/05/2019	Delhi	Cochin	NaN	09:45	

	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
9039	09:25 07 May	23h 40m	NaN	No info	7480

In [3]:

```

1 #deleting null value records
2 train_data.dropna(inplace=True)
3 print(train_data.isnull().sum())

```

```

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 [4]:

```

1 #copying dataset to perform featuriazation
2 data = train_data.copy()
3 print(data.shape)
4 print(data.dtypes)

```

```

(10682, 11)
Airline          object
Date_of_Journey  object
Source           object
Destination      object
Route            object
Dep_Time         object
Arrival_Time     object
Duration         object
Total_Stops      object
Additional_Info   object
Price            int64
dtype: object

```

In [5]:

```

1 #changing datatypes
2 #data['Date_of_Journey'] = pd.to_datetime(data['Date_of_Journey'])
3 #data['Dep_Time'] = pd.to_datetime(data['Dep_Time'])
4 #data['Arrival_Time'] = pd.to_datetime(data['Arrival_Time'])
5
6 #function to change data type
7 def datetime(col):
8     data[col]=pd.to_datetime(data[col])
9
10 for feature in ['Date_of_Journey', 'Dep_Time', 'Arrival_Time']:
11     datetime(feature)
12
13 print(data.dtypes)
14

```

```

Airline          object
Date_of_Journey  datetime64[ns]
Source           object
Destination      object
Route            object
Dep_Time         datetime64[ns]
Arrival_Time     datetime64[ns]
Duration         object
Total_Stops      object
Additional_Info   object
Price            int64
dtype: object

```

C:\Users\Pradeep\AppData\Local\Temp\ipykernel_10760\1318083259.py:8: UserWarning: Parsing dates in DD/MM/YYYY format when dayfirst=False (the default) was specified. This may lead to inconsistently parsed dates! Specify a format to ensure consistent parsing.

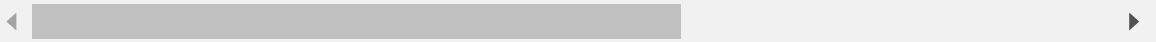
```
data[col]=pd.to_datetime(data[col])
```

In [6]:

```
1 #extracting year,month,day from Date_of_Journey column
2 data['journey_day']=data['Date_of_Journey'].dt.day
3 data['journey_month']=data['Date_of_Journey'].dt.month
4 data['journey_year']=data['Date_of_Journey'].dt.year
5 data.drop('Date_of_Journey',axis=1,inplace=True)
6 data.head()
```

Out[6]:

	Airline	Source	Destination	Route	Dep_Time	Arrival_Time	Duration	Total_Stops	Adc
0	IndiGo	Banglore	New Delhi	BLR → DEL	2023-05-13 22:20:00	2023-03-22 01:10:00	2h 50m	non-stop	
1	Air India	Kolkata	Banglore	CCU → IXR → BBI → BLR	2023-05-13 05:50:00	2023-05-13 13:15:00	7h 25m	2 stops	
2	Jet Airways	Delhi	Cochin	DEL → LKO → BOM → COK	2023-05-13 09:25:00	2023-06-10 04:25:00	19h	2 stops	
3	IndiGo	Kolkata	Banglore	CCU → NAG → BLR	2023-05-13 18:05:00	2023-05-13 23:30:00	5h 25m	1 stop	
4	IndiGo	Banglore	New Delhi	BLR → NAG → DEL	2023-05-13 16:50:00	2023-05-13 21:35:00	4h 45m	1 stop	



In [7]:

```
1 #removing day,month,year from Dep_Time and Arrival_Time column
2 def extract_hour_min(df,col):
3     df[col+'_hour']=df[col].dt.hour
4     df[col+'_minute']=df[col].dt.minute
5     df.drop(col,axis=1,inplace=True)
6
7
8 extract_hour_min(data,'Dep_Time')
9 extract_hour_min(data,'Arrival_Time')
10 data.head()
```

Out[7]:

	Airline	Source	Destination	Route	Duration	Total_Stops	Additional_Info	Price	journ
0	IndiGo	Banglore	New Delhi	BLR → DEL	2h 50m	non-stop	No info	3897	
1	Air India	Kolkata	Banglore	CCU → IXR → BBI → BLR	7h 25m	2 stops	No info	7662	
2	Jet Airways	Delhi	Cochin	DEL → LKO → BOM → COK	19h	2 stops	No info	13882	
3	IndiGo	Kolkata	Banglore	CCU → NAG → BLR	5h 25m	1 stop	No info	6218	
4	IndiGo	Banglore	New Delhi	BLR → NAG → DEL	4h 45m	1 stop	No info	13302	

In [8]:

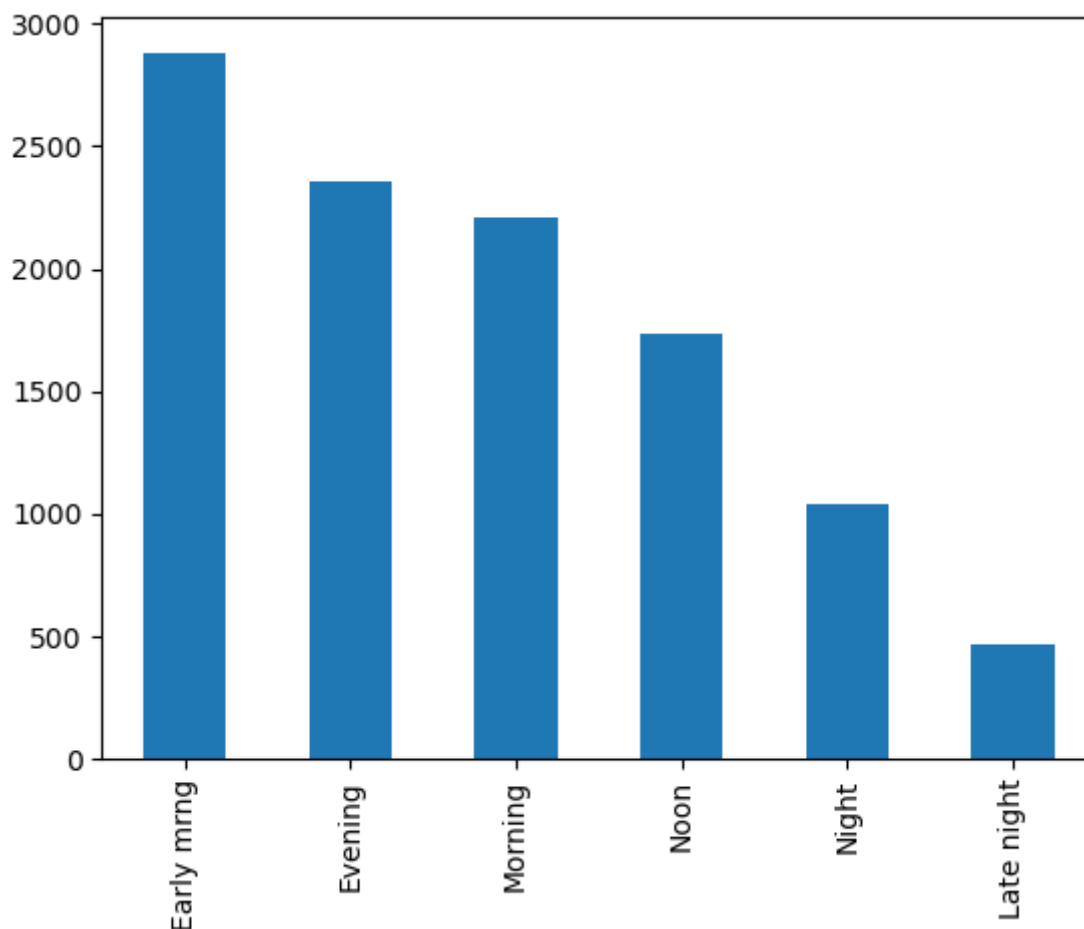
```
1 #data analysis for departure time and arrival time
2 def flight_time(x):
3     if ( x> 4) and (x<=8 ):
4         return 'Early mrng'
5
6     elif ( x>8 ) and (x<=12 ):
7         return 'Morning'
8
9     elif ( x>12 ) and (x<=16 ):
10        return 'Noon'
11
12    elif ( x>16 ) and (x<=20 ):
13        return 'Evening'
14
15    elif ( x>20 ) and (x<=24 ):
16        return 'Night'
17    else:
18        return 'Late night'
```

In [9]:

```
1 data['Dep_Time_hour'].apply(flight_time).value_counts().plot(kind='bar')
```

Out[9]:

<AxesSubplot: >

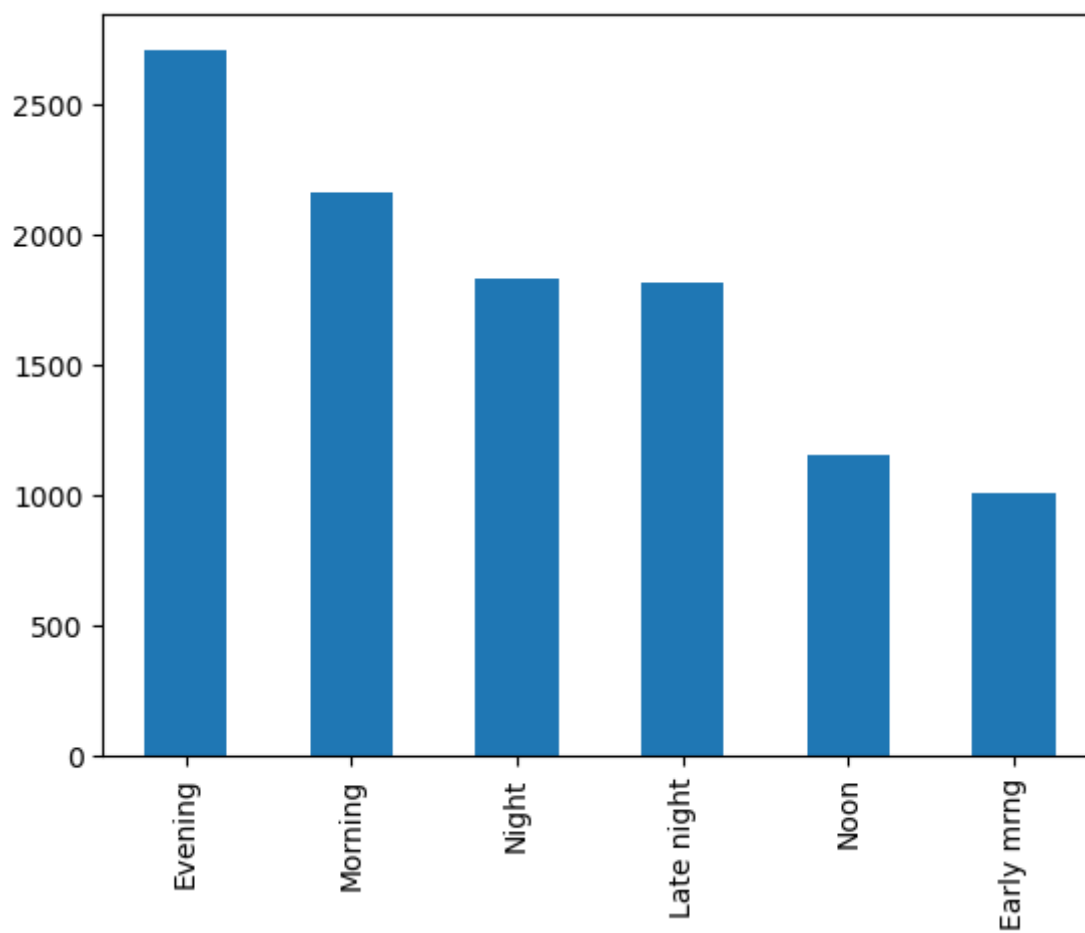


In [10]:

```
1 data['Arrival_Time_hour'].apply(flight_time).value_counts().plot(kind='bar')
```

Out[10]:

<AxesSubplot: >



In [11]:

```
1 data.head()
```

Out[11]:

	Airline	Source	Destination	Route	Duration	Total_Stops	Additional_Info	Price	journ
0	IndiGo	Banglore	New Delhi	BLR → DEL	2h 50m	non-stop	No info	3897	
1	Air India	Kolkata	Banglore	CCU → IXR → BBI → BLR	7h 25m	2 stops	No info	7662	
2	Jet Airways	Delhi	Cochin	DEL → LKO → BOM → COK	19h	2 stops	No info	13882	
3	IndiGo	Kolkata	Banglore	CCU → NAG → BLR	5h 25m	1 stop	No info	6218	
4	IndiGo	Banglore	New Delhi	BLR → NAG → DEL	4h 45m	1 stop	No info	13302	

In [12]:

```
1 #Processing Duration column since we need to make it uniform Like 5h 5m
2 def duration(x):
3     if 'h' not in x:
4         x='0h '+x
5     elif 'm' not in x:
6         x=x+' 0m'
7     return x
8
9 data['Duration']=data['Duration'].apply(duration)
10 data.head()
11
```

Out[12]:

	Airline	Source	Destination	Route	Duration	Total_Stops	Additional_Info	Price	journ
0	IndiGo	Banglore	New Delhi	BLR → DEL	2h 50m	non-stop	No info	3897	
1	Air India	Kolkata	Banglore	CCU → IXR → BBI → BLR	7h 25m	2 stops	No info	7662	
2	Jet Airways	Delhi	Cochin	DEL → LKO → BOM → COK	19h 0m	2 stops	No info	13882	
3	IndiGo	Kolkata	Banglore	CCU → NAG → BLR	5h 25m	1 stop	No info	6218	
4	IndiGo	Banglore	New Delhi	BLR → NAG → DEL	4h 45m	1 stop	No info	13302	

In [13]:

```
1 #splitting hour and minute from Duration column
2 #print(data['Duration'][0].split(' ')[0]) #splitting hr only but it contains hr stri
3 #print(int(data['Duration'][0].split(' ')[0][0:-1])) #[0:-1] where it will exclude l
4 #print(data['Duration'][0].split(' ')[1]) #splitting min only but it contains min st
5 #print(int(data['Duration'][0].split(' ')[1][0:-1])) #[0:-1] where it will exclude l
6
7 #new column for duration hour and duration minute
8 data['Duration_hours']=data['Duration'].apply(lambda x:int(x.split(' ')[0][0:-1]))
9 data['Duration_mins']=data['Duration'].apply(lambda x:int(x.split(' ')[1][0:-1]))
10 data.head()
11
```

Out[13]:

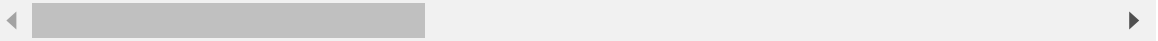
	Airline	Source	Destination	Route	Duration	Total_Stops	Additional_Info	Price	journ
0	IndiGo	Banglore	New Delhi	BLR → DEL	2h 50m	non-stop	No info	3897	
1	Air India	Kolkata	Banglore	CCU → IXR → BBI → BLR	7h 25m	2 stops	No info	7662	
2	Jet Airways	Delhi	Cochin	DEL → LKO → BOM → COK	19h 0m	2 stops	No info	13882	
3	IndiGo	Kolkata	Banglore	CCU → NAG → BLR	5h 25m	1 stop	No info	6218	
4	IndiGo	Banglore	New Delhi	BLR → NAG → DEL	4h 45m	1 stop	No info	13302	

In [14]:

```
1 #duration in mins
2 data['Duration_total_mins']=data['Duration'].str.replace('h','*60').str.replace(' ',
3 data.head()
4 #example of eval
5 #eval('2*60+50*1') #converts string expression into python expression
```

Out[14]:

	Airline	Source	Destination	Route	Duration	Total_Stops	Additional_Info	Price	journ
0	IndiGo	Banglore	New Delhi	BLR → DEL	2h 50m	non-stop	No info	3897	
1	Air India	Kolkata	Banglore	CCU → IXR → BBI → BLR	7h 25m	2 stops	No info	7662	
2	Jet Airways	Delhi	Cochin	DEL → LKO → BOM → COK	19h 0m	2 stops	No info	13882	
3	IndiGo	Kolkata	Banglore	CCU → NAG → BLR	5h 25m	1 stop	No info	6218	
4	IndiGo	Banglore	New Delhi	BLR → NAG → DEL	4h 45m	1 stop	No info	13302	

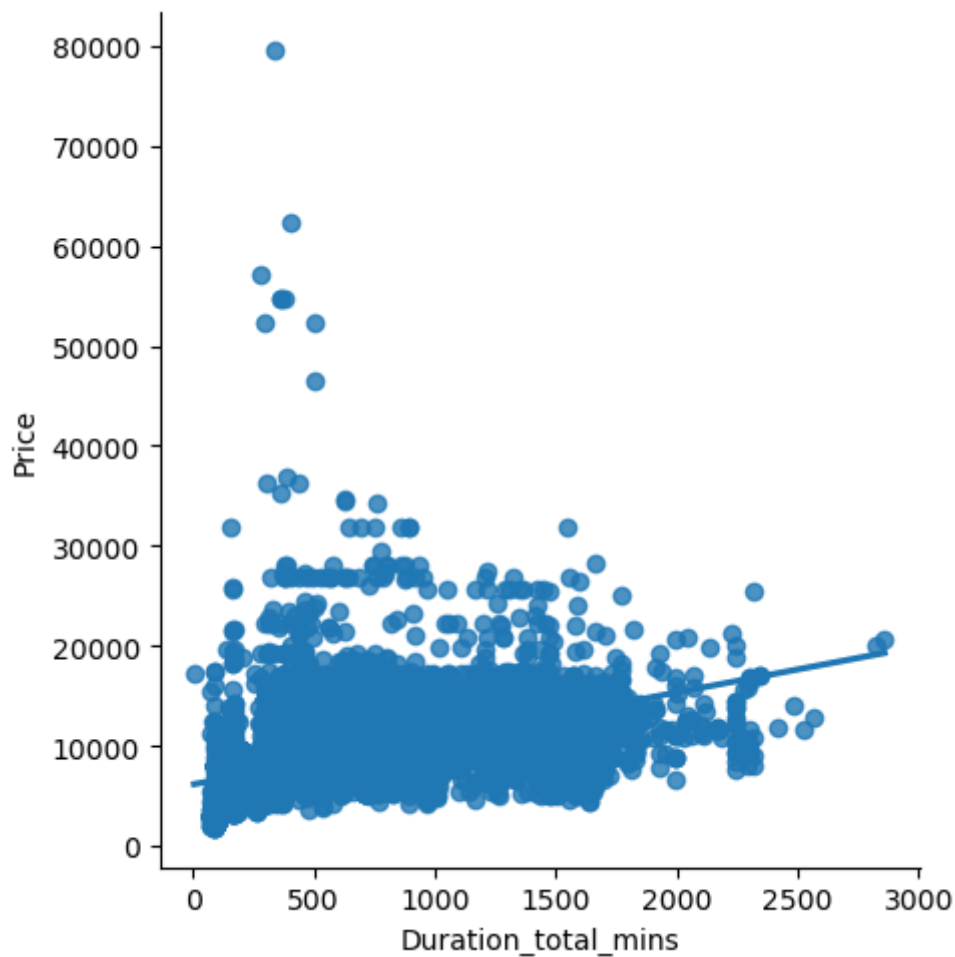


In [15]:

```
1 #Price Vs Duration
2 sns.lmplot(x='Duration_total_mins',y='Price',data=data)
3 #As duration time increase the price also increases
```

Out[15]:

<seaborn.axisgrid.FacetGrid at 0x2ddcea68f40>



In [16]:

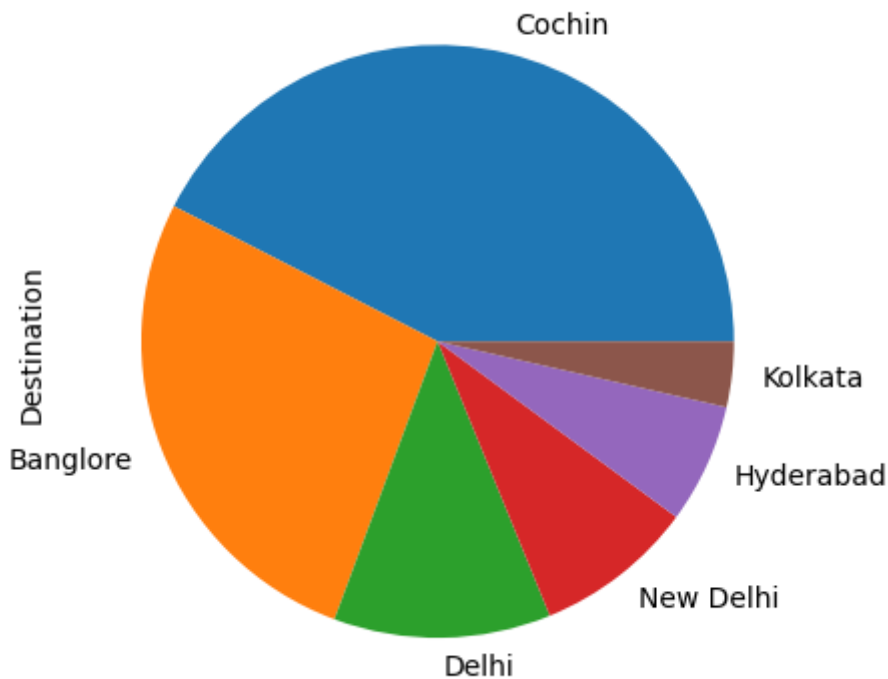
```

1 #max final destination
2 #data['Destination'].unique()
3 #data['Destination'].value_counts().plot(kind='bar')
4 data['Destination'].value_counts().plot(kind='pie')
5 #plt.axis('off')

```

Out[16]:

<AxesSubplot: ylabel='Destination'>



In [17]:

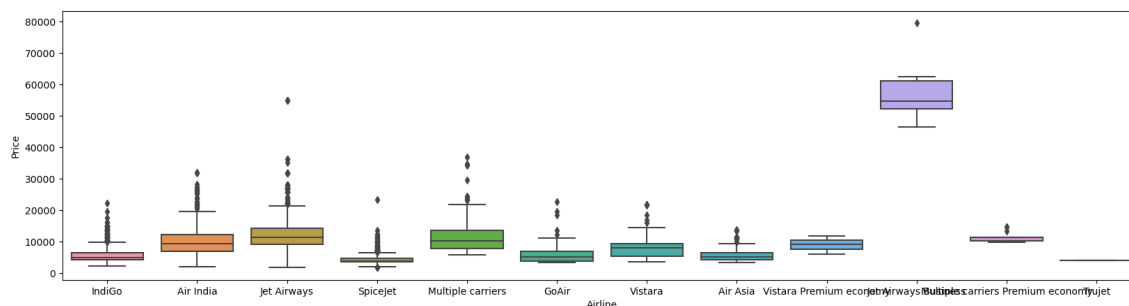
```

1 #airline and their price distribution using boxplot
2 plt.figure(figsize=(20,5))
3 sns.boxplot(y='Price',x='Airline',data=data)

```

Out[17]:

<AxesSubplot: xlabel='Airline', ylabel='Price'>



In [18]:

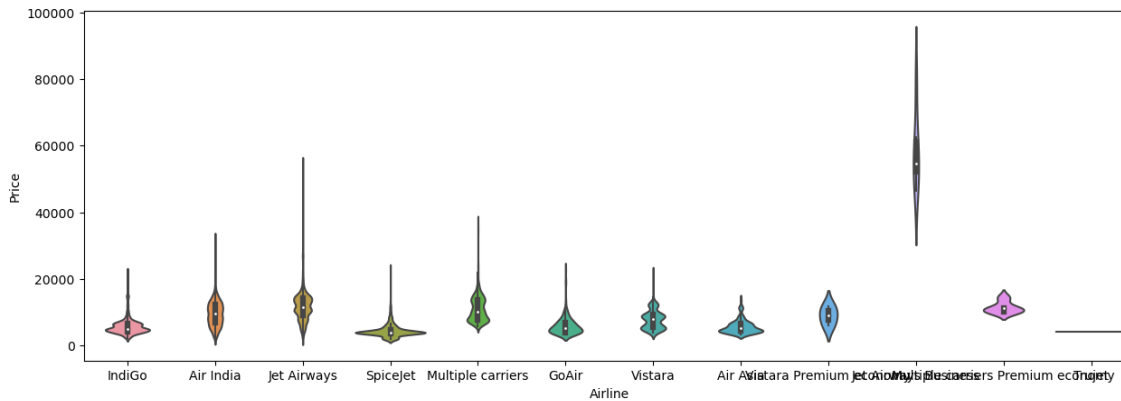
```

1 #airline and their price distribution using violinplot(box+distribution)
2 plt.figure(figsize=(15,5))
3 sns.violinplot(y='Price',x='Airline',data=data)
4

```

Out[18]:

<AxesSubplot: xlabel='Airline', ylabel='Price'>



In [19]:

```

1 #viewing additional_info records value percentage
2 np.round(data['Additional_Info'].value_counts()/len(data)*100,2)

```

Out[19]:

No info	78.11
In-flight meal not included	18.55
No check-in baggage included	3.00
1 Long layover	0.18
Change airports	0.07
Business class	0.04
No Info	0.03
1 Short layover	0.01
Red-eye flight	0.01
2 Long layover	0.01

Name: Additional_Info, dtype: float64

In [20]:

```

1 #removing features
2 #additional info has more No Info
3 #Route since it not a great feature
4 #Duration_total_mins as already we have Duration_hour Duration_min
5 #journey_year as it is same value for all record
6 data.drop(columns=['Additional_Info', 'Route', 'Duration_total_mins', 'journey_year'],a
7 data.head()

```

Out[20]:

	Airline	Source	Destination	Duration	Total_Stops	Price	journey_day	journey_month
0	IndiGo	Banglore	New Delhi	2h 50m	non-stop	3897	24	3
1	Air India	Kolkata	Banglore	7h 25m	2 stops	7662	5	1
2	Jet Airways	Delhi	Cochin	19h 0m	2 stops	13882	6	9
3	IndiGo	Kolkata	Banglore	5h 25m	1 stop	6218	5	12
4	IndiGo	Banglore	New Delhi	4h 45m	1 stop	13302	3	1

In [21]:

```

1 #separating categorical and numerical columns
2 cat_col=[col for col in data.columns if data[col].dtype=='object']
3 num_col=[col for col in data.columns if data[col].dtype!='object']
4 #num_col=[col for col in data.columns if data[col].dtype=='int64']
5 print("categorical columns are : ",cat_col)
6 print("\nNumerical columns are : ",num_col)

```

categorical columns are : ['Airline', 'Source', 'Destination', 'Duration', 'Total_Stops']

Numerical columns are : ['Price', 'journey_day', 'journey_month', 'Dep_Time_hour', 'Dep_Time_minute', 'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours', 'Duration_mins']

In [22]:

```

1 #one-hot encoding without using python package for source column
2 for category in data['Source'].unique():
3     data['Source_'+category]=data['Source'].apply(lambda x: 1 if x==category else 0)
4
5 data.head()

```

Out[22]:

	Airline	Source	Destination	Duration	Total_Stops	Price	journey_day	journey_month
0	IndiGo	Banglore	New Delhi	2h 50m	non-stop	3897	24	3
1	Air India	Kolkata	Banglore	7h 25m	2 stops	7662	5	1
2	Jet Airways	Delhi	Cochin	19h 0m	2 stops	13882	6	9
3	IndiGo	Kolkata	Banglore	5h 25m	1 stop	6218	5	12
4	IndiGo	Banglore	New Delhi	4h 45m	1 stop	13302	3	1

In [23]:

```

1 #Target guided encoding on airline column
2 '''
3 airlines = data['Airline'].unique()
4 dict1={key:index for index,key in enumerate(airlines,0)}
5 dict1
6 In the above code it is not ordered so we are ordered based on price
7 '''
8 airlines=data.groupby(['Airline'])['Price'].mean().sort_values().index
9 dict={key:index for index,key in enumerate(airlines,0)}
10 print(dict)
11 '''
12 In the above code we have ordered the name of airline based on mean price from low
13 '''
14 data['Airline']=data['Airline'].map(dict)
15 data.head()

```

```
{'Trujet': 0, 'SpiceJet': 1, 'Air Asia': 2, 'IndiGo': 3, 'GoAir': 4, 'Vistara': 5, 'Vistara Premium economy': 6, 'Air India': 7, 'Multiple carriers': 8, 'Multiple carriers Premium economy': 9, 'Jet Airways': 10, 'Jet Airways Business': 11}
```

Out[23]:

	Airline	Source	Destination	Duration	Total_Stops	Price	journey_day	journey_month
0	3	Banglore	New Delhi	2h 50m	non-stop	3897	24	3
1	7	Kolkata	Banglore	7h 25m	2 stops	7662	5	1
2	10	Delhi	Cochin	19h 0m	2 stops	13882	6	9
3	3	Kolkata	Banglore	5h 25m	1 stop	6218	5	12
4	3	Banglore	New Delhi	4h 45m	1 stop	13302	3	1

In [24]:

```
1 data['Destination'].unique()
```

Out[24]:

```
array(['New Delhi', 'Banglore', 'Cochin', 'Kolkata', 'Delhi', 'Hyderaba  
d'],  
      dtype=object)
```

In [25]:

```
1 #we have delhi as two destination so we should replace it  
2 data['Destination'].replace('New Delhi', 'Delhi', inplace=True)  
3 data['Destination'].unique()
```

Out[25]:

```
array(['Delhi', 'Banglore', 'Cochin', 'Kolkata', 'Hyderabad'],  
      dtype=object)
```

In [26]:

```
1 #Target guided encoding on destination column  
2 dest=data.groupby(['Destination'])['Price'].mean().sort_values().index  
3 dict1={key:index for index, key in enumerate(dest,0)}  
4 print(dict1)  
5 data['Destination']=data['Destination'].map(dict1)  
6 data.head()
```

```
{'Kolkata': 0, 'Hyderabad': 1, 'Delhi': 2, 'Banglore': 3, 'Cochin': 4}
```

Out[26]:

	Airline	Source	Destination	Duration	Total_Stops	Price	journey_day	journey_month
0	3	Banglore	2	2h 50m	non-stop	3897	24	3
1	7	Kolkata	3	7h 25m	2 stops	7662	5	1
2	10	Delhi	4	19h 0m	2 stops	13882	6	9
3	3	Kolkata	3	5h 25m	1 stop	6218	5	12
4	3	Banglore	2	4h 45m	1 stop	13302	3	1

In [27]:

```
1 data['Total_Stops'].unique()
```

Out[27]:

```
array(['non-stop', '2 stops', '1 stop', '3 stops', '4 stops'],  
      dtype=object)
```


In [28]:

```
1 #manual encoding on total_stops
2 stops = {'non-stop':0, '1 stop':1, '2 stops':2, '3 stops':3, '4 stops':4}
3 data['Total_Stops']=data['Total_Stops'].map(stops)
4 data.head()
```

Out[28]:

	Airline	Source	Destination	Duration	Total_Stops	Price	journey_day	journey_month
0	3	Banglore	2	2h 50m	0	3897	24	3
1	7	Kolkata	3	7h 25m	2	7662	5	1
2	10	Delhi	4	19h 0m	2	13882	6	9
3	3	Kolkata	3	5h 25m	1	6218	5	12
4	3	Banglore	2	4h 45m	1	13302	3	1

In [29]:

```

1 #detecting outliers using plotting
2 def plot(df,col):
3     fig,(ax1,ax2,ax3)=plt.subplots(3,1)
4     sns.distplot(df[col],ax=ax1)
5     sns.boxplot(df[col],ax=ax2,orient="h")
6     sns.histplot(df[col],ax=ax3)
7
8 plot(data,'Price')

```

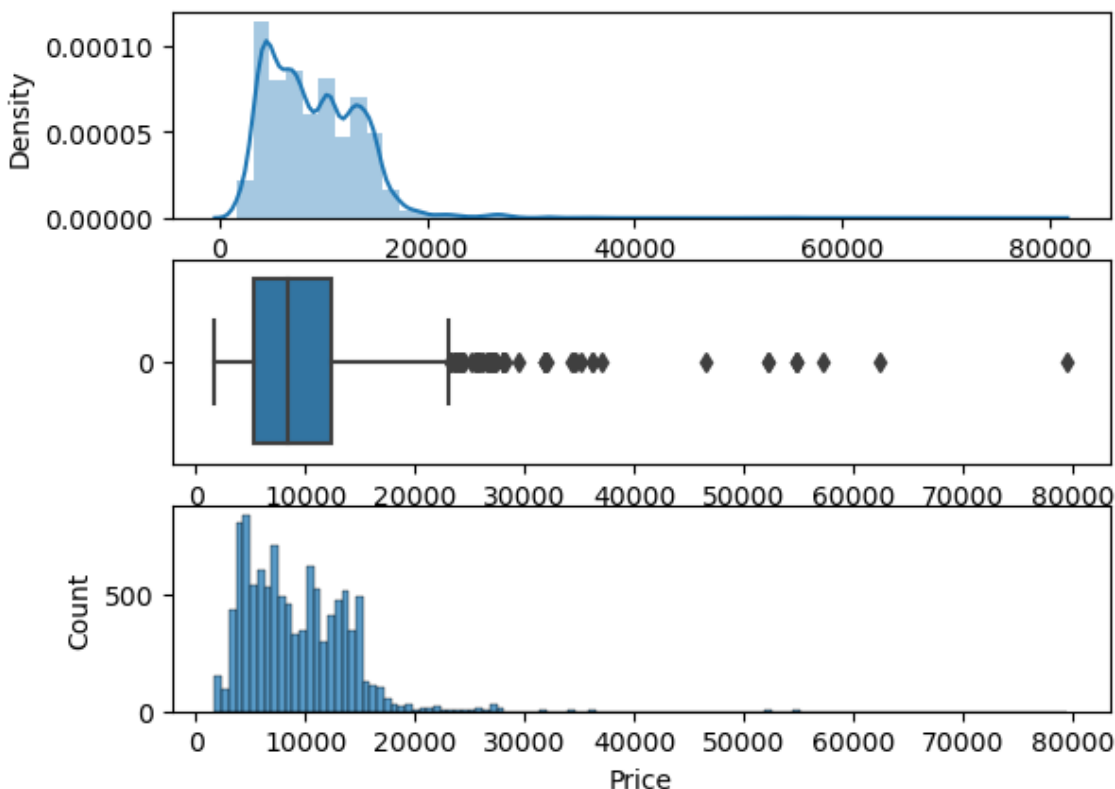
C:\Users\Pradeep\AppData\Local\Temp\ipykernel_10760\3914968476.py:4: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(df[col],ax=ax1)
```



In [30]:

```
1 #outliers using IQR
2 per25 = data['Price'].quantile(0.25)
3 per75 = data['Price'].quantile(0.75)
4 iqr = per75 - per25
5 print("low_iqr",per25-1.5*iqr)
6 print("high_iqr",per75+1.5*iqr)
7 print("Number of outliers",len(data[data['Price']>per75+(1.5*iqr)]))
```

low_iqr -5367.0

high_iqr 23017.0

Number of outliers 94

In [31]:

```

1 #imputing outliers with median,we can't use mean since it will have a impact from ou
2 data['Price']=np.where(data['Price']>=25000,data['Price'].median(),data['Price'])
3 plot(data,'Price')

```

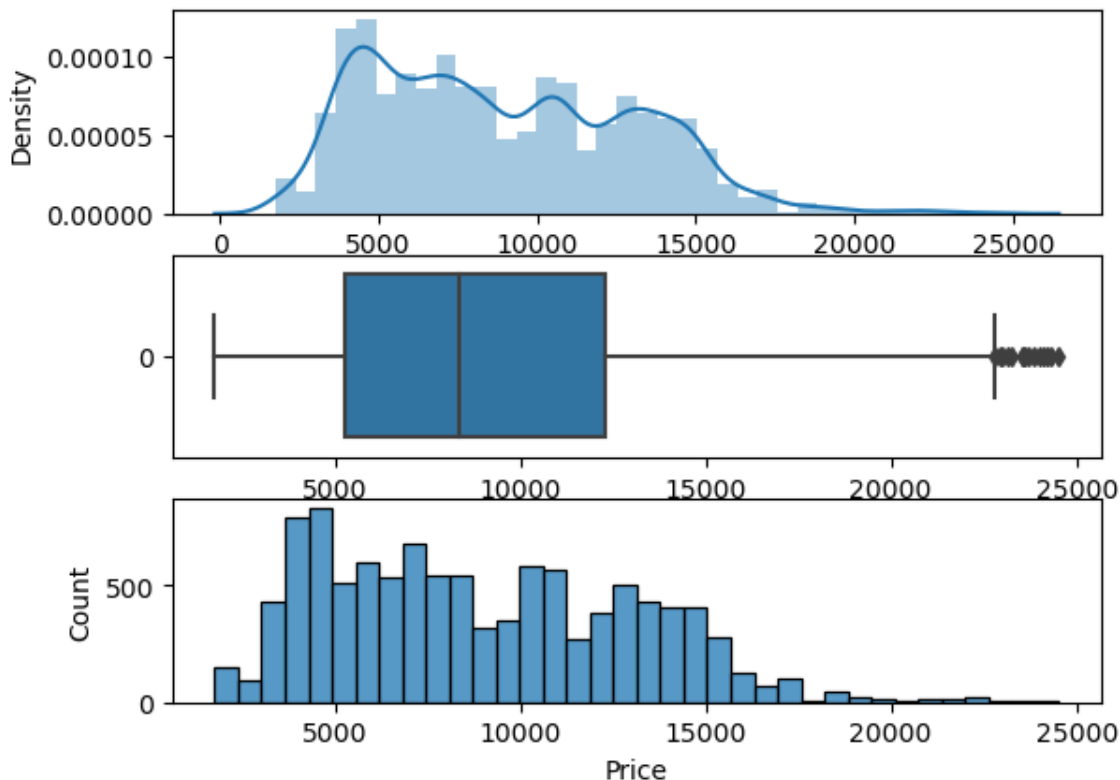
C:\Users\Pradeep\AppData\Local\Temp\ipykernel_10760\3914968476.py:4: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(df[col],ax=ax1)
```



In [32]:

```
1 data.head()
```

Out[32]:

	Airline	Source	Destination	Duration	Total_Stops	Price	journey_day	journey_month
0	3	Banglore	2	2h 50m	0	3897.0	24	3
1	7	Kolkata	3	7h 25m	2	7662.0	5	1
2	10	Delhi	4	19h 0m	2	13882.0	6	9
3	3	Kolkata	3	5h 25m	1	6218.0	5	12
4	3	Banglore	2	4h 45m	1	13302.0	3	1

In [33]:

```
1 #removing features and viewing datatypes if we have any object column
2 data.drop(columns=['Source', 'Duration'],axis=1,inplace=True)
3 data.dtypes
```

Out[33]:

```
Airline          int64
Destination      int64
Total_Stops      int64
Price            float64
journey_day      int64
journey_month    int64
Dep_Time_hour    int64
Dep_Time_minute  int64
Arrival_Time_hour int64
Arrival_Time_minute int64
Duration_hours   int64
Duration_mins    int64
Source_Banglore  int64
Source_Kolkata   int64
Source_Delhi     int64
Source_Chennai   int64
Source_Mumbai    int64
dtype: object
```

In [34]:

```

1 #feature importance
2 from sklearn.feature_selection import mutual_info_regression
3 X=data.drop(['Price'],axis=1)
4 y=data['Price']
5 print(mutual_info_regression(X,y))
6 imp=pd.DataFrame(mutual_info_regression(X,y),index=X.columns)
7 imp.columns=['importance']
8 imp.sort_values(by='importance',ascending=False)

```

```

[0.96355969 0.99811016 0.79041637 0.19582242 0.23101291 0.3347241
 0.25849524 0.40084984 0.34004384 0.46542898 0.34104411 0.38022903
 0.45346214 0.52603833 0.12611241 0.20313238]

```

Out[34]:

	importance
Destination	0.995398
Airline	0.965107
Total_Stops	0.787540
Source_Delhi	0.525961
Duration_hours	0.463217
Source_Kolkata	0.457689
Arrival_Time_hour	0.398395
Source_Banglore	0.377982
Arrival_Time_minute	0.341728
Dep_Time_hour	0.341074
Duration_mins	0.330745
Dep_Time_minute	0.262337
journey_month	0.235800
Source_Mumbai	0.200794
journey_day	0.193360
Source_Chennai	0.130994

In [35]:

```

1 #RandomForestRegressor ML model
2 from sklearn.ensemble import RandomForestRegressor
3 from sklearn.model_selection import train_test_split
4 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25)
5 ml_model=RandomForestRegressor()
6 model=ml_model.fit(X_train,y_train)
7 y_pred=model.predict(X_test)
8 y_pred

```

Out[35]:

```
array([ 6766.97,  9780.89,  4860.18, ..., 2775.05, 18214.05, 10316.28])
```

In [36]:

```

1 #comparing actual vs predicted value
2 z = y_test.values
3 data_pred_df = pd.DataFrame(y_pred,columns=["Predicted Price"])
4 data_pred_df['Actual Price'] = z
5 data_pred_df

```

Out[36]:

	Predicted Price	Actual Price
0	6766.970000	4544.0
1	9780.890000	9646.0
2	4860.180000	4409.0
3	8034.290000	6610.0
4	7348.348000	12681.0
...
2666	8501.370000	8085.0
2667	8586.991833	7064.0
2668	2775.050000	2754.0
2669	18214.050000	18275.0
2670	10316.280000	14714.0

2671 rows × 2 columns

In [37]:

```

1 #saving ML model
2 import pickle
3 file = open(r'D:/Airline Tickets/random_ml.pkl','wb')
4 pickle.dump(model,file)
5 model = open(r'D:/Airline Tickets/random_ml.pkl','rb')
6 forest = pickle.load(model)
7 forest.predict(X_test)

```

Out[37]:

```
array([ 6766.97,  9780.89,  4860.18, ..., 2775.05, 18214.05, 10316.28])
```

In [38]:

```

1 # MAPE function for evaluation
2 def mape(y_true,y_pred):
3     y_true,y_pred=np.array(y_true),np.array(y_pred)
4
5     return np.mean(np.abs((y_true-y_pred)/y_true))*100
6
7 mape(y_test,forest.predict(X_test))

```

Out[38]:

13.383292237272823

In [39]:

```
1  # ML pipeline
2  def predict(ml_model):
3
4      model=ml_model.fit(X_train,y_train)
5      print('Training_score: {}'.format(model.score(X_train,y_train)))
6      y_prediction=model.predict(X_test)
7      print('Predictions are : {}'.format(y_prediction))
8
9
10     from sklearn import metrics
11     r2_score=metrics.r2_score(y_test,y_prediction)
12     print('r2_score: {}'.format(r2_score))
13     print('MSE : ', metrics.mean_squared_error(y_test,y_prediction))
14     print('MAE : ', metrics.mean_absolute_error(y_test,y_prediction))
15     print('RMSE : ', np.sqrt(metrics.mean_squared_error(y_test,y_prediction)))
16     print('MAPE : ', mape(y_test,y_prediction))
17
18
19
20 predict(RandomForestRegressor())
```

Training_score: 0.9522617589594621

Predictions are : [6763.504 9718.17 4802.99 ... 2776.

72

18213.71 10684.32333333]

r2_score: 0.7930845931869165

MSE : 3526433.944144381

MAE : 1201.976624953201

RMSE : 1877.8801729994332

MAPE : 13.395135057196775

In [40]:

```

1  #hypertuning ML model
2  from sklearn.model_selection import RandomizedSearchCV
3
4  reg_rf=RandomForestRegressor()
5
6  # Number of trees in random forest
7  n_estimators=[int(x) for x in np.linspace(start=1000,stop=1200,num=24)]
8
9  # Number of features to consider at every split
10 max_features=["auto", "sqrt"]
11
12 # Maximum number of levels in tree
13 max_depth=[int(x) for x in np.linspace(start=5,stop=30,num=10)]
14
15 # Minimum number of samples required to split a node
16 min_samples_split=[5,10,15,100]
17
18 print(n_estimators)
19 print(max_features)
20 print(max_depth)
21 print(min_samples_split)
22
23 # Create the grid or hyper-parameter space
24 random_grid={
25     'n_estimators':n_estimators,
26     'max_features':max_features,
27     'max_depth':max_depth,
28     'min_samples_split':min_samples_split
29 }
30 }
31
32
33 rf_Random=RandomizedSearchCV(reg_rf,param_distributions=random_grid,cv=3,verbose=2,r
34
35 rf_Random.fit(X_train,y_train)
36
37 ### to get your best model..
38 print(rf_Random.best_params_)
39
40 pred2=rf_Random.predict(X_test)
41
42 from sklearn import metrics
43 metrics.r2_score(y_test,pred2)

```

```

[1000, 1008, 1017, 1026, 1034, 1043, 1052, 1060, 1069, 1078, 1086, 1095, 1
104, 1113, 1121, 1130, 1139, 1147, 1156, 1165, 1173, 1182, 1191, 1200]
['auto', 'sqrt']
[5, 7, 10, 13, 16, 18, 21, 24, 27, 30]
[5, 10, 15, 100]
Fitting 3 folds for each of 10 candidates, totalling 30 fits

```

```

C:\Users\Pradeep\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py:4
13: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and wi
ll be removed in 1.3. To keep the past behaviour, explicitly set `max_feat
ures=1.0` or remove this parameter as it is also the default value for Ran
domForestRegressors and ExtraTreesRegressors.
warn(

```

```
{'n_estimators': 1017, 'min_samples_split': 15, 'max_features': 'auto', 'max_depth': 27}
```

Out[40]:

0.822327208304843

In []:

1	
---	--