

Airline Fare Prediction

February 6, 2025

1 Airline Fare Prediction 2025

1.1 Importing Libraries

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

```
[2]: train_data = pd.read_excel(r"E:\DS Projects\1. Predict Fare of Airlines Tickets_
    using Machine Learning\Data_Train.xlsx")
```

```
[3]: train_data.head(4)
```

```
[3]:      Airline Date_of_Journey  Source Destination  Route \
0      IndiGo      24/03/2019  Bangalore  New Delhi      BLR → DEL
1    Air India      1/05/2019  Kolkata   Bangalore  CCU → IXR → BBI → BLR
2  Jet Airways      9/06/2019    Delhi    Cochin    DEL → LKO → BOM → COK
3      IndiGo     12/05/2019  Kolkata   Bangalore      CCU → NAG → BLR
```

	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	05:50	13:15	7h 25m	2 stops	No info	7662
2	09:25	04:25 10 Jun	19h	2 stops	No info	13882
3	18:05	23:30	5h 25m	1 stop	No info	6218

```
[4]: train_data.tail(4)
```

```
[4]:      Airline Date_of_Journey  Source Destination \
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	
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
10679	No info	4145
10680	No info	7229
10681	No info	12648
10682	No info	11753

1.2 Data Cleaning: Missing Values

```
[5]: train_data.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
```

```
[6]: train_data.isnull().sum()
```

```
[6]: 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
```

```
[7]: train_data['Total_Stops'].isnull()
```

```
[7]: 0      False
      1      False
      2      False
      3      False
      4      False
      ...
      10678   False
      10679   False
      10680   False
      10681   False
      10682   False
      Name: Total_Stops, Length: 10683, dtype: bool
```

```
[8]: train_data[train_data['Total_Stops'].isnull()]
```

```
[8]:      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
```

```
[9]: train_data.dropna(inplace=True)
```

```
[10]: train_data.isnull().sum()
```

```
[10]: 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
```

```
[11]: train_data.dtypes
```

```
[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
```

```
[12]: train_data.info(memory_usage="deep")
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 10682 entries, 0 to 10682
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   Airline                10682 non-null  object
1   Date_of_Journey        10682 non-null  object
2   Source                 10682 non-null  object
3   Destination            10682 non-null  object
4   Route                  10682 non-null  object
5   Dep_Time               10682 non-null  object
6   Arrival_Time           10682 non-null  object
7   Duration                10682 non-null  object
8   Total_Stops            10682 non-null  object
9   Additional_Info        10682 non-null  object
10  Price                  10682 non-null  int64
dtypes: int64(1), object(10)
memory usage: 6.3 MB
```

1.3 Pre-Processing & Extraction of Derived Attributes

```
[13]: data = train_data.copy()
```

```
[14]: data.columns
```

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

```
[15]: data.head(2)
```

```
[15]:
```

	Airline	Date_of_Journey	Source	Destination	Route \
0	IndiGo	24/03/2019	Bangalore	New Delhi	BLR → DEL
1	Air India	1/05/2019	Kolkata	Bangalore	CCU → IXR → BBI → BLR

	Dep_Time	Arrival_Time	Duration	Total_Stops	Additional_Info	Price
0	22:20	01:10 22 Mar	2h 50m	non-stop	No info	3897
1	05:50	13:15	7h 25m	2 stops	No info	7662

```
[16]: data.dtypes
```

```
[16]: 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
```

```
[17]: import warnings
      from warnings import filterwarnings
      filterwarnings("ignore")
```

1.4 Changing into Date Data type

```
[18]: def change_into_Datetime(col):
      data[col] = pd.to_datetime(data[col])
```

```
[19]: data.columns
```

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

```
[20]: for feature in ['Dep_Time', 'Arrival_Time', 'Date_of_Journey']:
      change_into_Datetime(feature)
```

```
[21]: data.dtypes
```

```
[21]: 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
```

1.5 Separating Day, Month and Year

```
[22]: data["Journey_day"] = data['Date_of_Journey'].dt.day
```

```
[23]: data["Journey_month"] = data['Date_of_Journey'].dt.month
```

```
[24]: data["Journey_year"] = data['Date_of_Journey'].dt.year
```

```
[25]: data.head(3)
```

```
[25]:
```

	Airline	Date_of_Journey	Source	Destination	Route	\
0	IndiGo	2019-03-24	Banglore	New Delhi	BLR → DEL	
1	Air India	2019-05-01	Kolkata	Banglore	CCU → IXR → BBI → BLR	
2	Jet Airways	2019-06-09	Delhi	Cochin	DEL → LKO → BOM → COK	

	Dep_Time	Arrival_Time	Duration	Total_Stops	\
0	2025-02-06 22:20:00	2025-03-22 01:10:00	2h 50m	non-stop	
1	2025-02-06 05:50:00	2025-02-06 13:15:00	7h 25m	2 stops	
2	2025-02-06 09:25:00	2025-06-10 04:25:00	19h	2 stops	

	Additional_Info	Price	Journey_day	Journey_month	Journey_year
0	No info	3897	24	3	2019
1	No info	7662	1	5	2019
2	No info	13882	9	6	2019

1.6 Cleaning Dep_Time and Arrival_Time & Extracting Derived Attributes (Hour and Mins)

```
[26]: def extract_hour_min(df , col):  
      df[col+"_hour"] = df[col].dt.hour  
      df[col+"_minute"] = df[col].dt.minute  
      return df.head(3)
```

```
[27]: data.columns
```

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

```
[28]: extract_hour_min(data, "Dep_Time")
```

```
[28]:
```

	Airline	Date_of_Journey	Source	Destination	Route	\
0	IndiGo	2019-03-24	Banglore	New Delhi	BLR → DEL	
1	Air India	2019-05-01	Kolkata	Banglore	CCU → IXR → BBI → BLR	
2	Jet Airways	2019-06-09	Delhi	Cochin	DEL → LKO → BOM → COK	

		Dep_Time	Arrival_Time	Duration	Total_Stops	\
0	2025-02-06	22:20:00	2025-03-22 01:10:00	2h 50m	non-stop	
1	2025-02-06	05:50:00	2025-02-06 13:15:00	7h 25m	2 stops	
2	2025-02-06	09:25:00	2025-06-10 04:25:00	19h	2 stops	

	Additional_Info	Price	Journey_day	Journey_month	Journey_year	\
0	No info	3897	24	3	2019	
1	No info	7662	1	5	2019	
2	No info	13882	9	6	2019	

	Dep_Time_hour	Dep_Time_minute
0	22	20
1	5	50
2	9	25

```
[29]: extract_hour_min(data, "Arrival_Time")
```

```
[29]:
```

	Airline	Date_of_Journey	Source	Destination	Route	\
0	IndiGo	2019-03-24	Bangalore	New Delhi	BLR → DEL	
1	Air India	2019-05-01	Kolkata	Bangalore	CCU → IXR → BBI → BLR	
2	Jet Airways	2019-06-09	Delhi	Cochin	DEL → LKO → BOM → COK	

		Dep_Time	Arrival_Time	Duration	Total_Stops	\
0	2025-02-06	22:20:00	2025-03-22 01:10:00	2h 50m	non-stop	
1	2025-02-06	05:50:00	2025-02-06 13:15:00	7h 25m	2 stops	
2	2025-02-06	09:25:00	2025-06-10 04:25:00	19h	2 stops	

	Additional_Info	Price	Journey_day	Journey_month	Journey_year	\
0	No info	3897	24	3	2019	
1	No info	7662	1	5	2019	
2	No info	13882	9	6	2019	

	Dep_Time_hour	Dep_Time_minute	Arrival_Time_hour	Arrival_Time_minute
0	22	20	1	10
1	5	50	13	15
2	9	25	4	25

```
[30]: cols_to_drop = ['Arrival_Time', 'Dep_Time']
data.drop(cols_to_drop , axis=1 , inplace=True)
```

```
[31]: data.head(3)
```

```
[31]:
```

	Airline	Date_of_Journey	Source	Destination	Route	\
0	IndiGo	2019-03-24	Bangalore	New Delhi	BLR → DEL	
1	Air India	2019-05-01	Kolkata	Bangalore	CCU → IXR → BBI → BLR	
2	Jet Airways	2019-06-09	Delhi	Cochin	DEL → LKO → BOM → COK	

	Duration	Total_Stops	Additional_Info	Price	Journey_day	Journey_month	\
0	2h 50m	non-stop	No info	3897	24	3	
1	7h 25m	2 stops	No info	7662	1	5	
2	19h	2 stops	No info	13882	9	6	

	Journey_year	Dep_Time_hour	Dep_Time_minute	Arrival_Time_hour	\
0	2019	22	20	1	
1	2019	5	50	13	
2	2019	9	25	4	

	Arrival_Time_minute
0	10
1	15
2	25

```
[32]: data.shape
```

```
[32]: (10682, 16)
```

1.7 Analysis: Most of the flights take-off

```
[33]: data.columns
```

```
[33]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
        'Duration', 'Total_Stops', 'Additional_Info', 'Price', 'Journey_day',
        'Journey_month', 'Journey_year', 'Dep_Time_hour', 'Dep_Time_minute',
        'Arrival_Time_hour', 'Arrival_Time_minute'],
        dtype='object')
```

```
[34]: def flight_dep_time(x):
        if (x>4) and (x<=8):
            return "Early Morning"

        elif (x>8) and (x<=12):
            return "Morning"

        elif (x>12) and (x<=16):
            return "Noon"

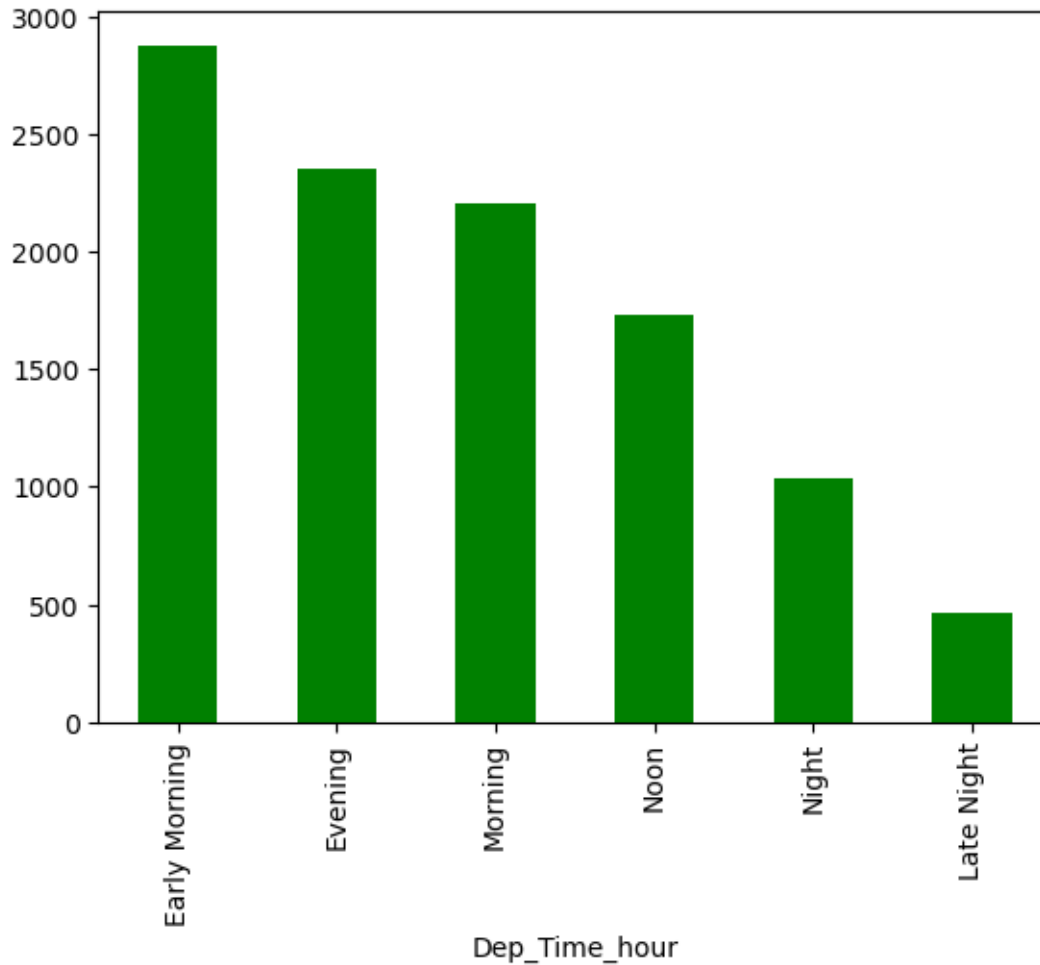
        elif (x>16) and (x<=20):
            return "Evening"

        elif (x>20) and (x<=24):
            return "Night"
        else:
            return "Late Night"
```



```
[35]: data['Dep_Time_hour'].apply(flight_dep_time).value_counts().plot(kind="bar" ,  
    ↪color="g")
```

```
[35]: <Axes: xlabel='Dep_Time_hour'>
```



```
[36]: import plotly  
import cufflinks as cf  
from cufflinks.offline import go_offline  
from plotly.offline import plot, iplot, init_notebook_mode , download_plotlyjs  
init_notebook_mode(connected=True)  
cf.go_offline()
```

```
[37]: data['Dep_Time_hour'].apply(flight_dep_time).value_counts().iplot(kind="bar")
```

```
[38]: data['Dep_Time_hour'].apply(flight_dep_time).value_counts()
```

```
[38]: Dep_Time_hour
      Early Morning    2880
      Evening          2357
      Morning          2209
      Noon             1731
      Night            1040
      Late Night        465
      Name: count, dtype: int64
```

1.8 Pre-Processing on Duration Feature

```
[39]: def preprocess_duration(x):
      if 'h' not in x:
          x = '0h' + ' ' + x
      elif 'm' not in x:
          x = x + ' ' + '0m'

      return x
```

```
[40]: data['Duration'] = data['Duration'].apply(preprocess_duration)
```

```
[41]: data['Duration']
```

```
[41]: 0      2h 50m
      1      7h 25m
      2     19h 0m
      3      5h 25m
      4      4h 45m
      ...
      10678   2h 30m
      10679   2h 35m
      10680    3h 0m
      10681   2h 40m
      10682   8h 20m
      Name: Duration, Length: 10682, dtype: object
```

```
[42]: data['Duration'][0]
```

```
[42]: '2h 50m'
```

```
[43]: '2h 50m'.split(' ')
```

```
[43]: ['2h', '50m']
```

```
[44]: '2h 50m'.split(' ')[0]
```

```
[44]: '2h'
```

```
[45]: '2h 50m'.split(' ')[0][0:-1]
```

```
[45]: '2'
```

```
[46]: type('2h 50m'.split(' ')[0][0:-1])
```

```
[46]: str
```

```
[47]: int('2h 50m'.split(' ')[0][0:-1])
```

```
[47]: 2
```

```
[48]: int('2h 50m'.split(' ')[1][0:-1])
```

```
[48]: 50
```

1.9 Using Lambda as Anonymous function

```
[49]: data['Duration_hours'] = data['Duration'].apply(lambda x: int(x.split(' ')[0][0:  
↪-1]))
```

```
[50]: data['Duration_mins'] = data['Duration'].apply(lambda x: int(x.split(' ')[1][0:  
↪-1]))
```

```
[51]: data.head(2)
```

```
[51]:
```

	Airline	Date_of_Journey	Source	Destination	Route	\
0	IndiGo	2019-03-24	Banglore	New Delhi	BLR → DEL	
1	Air India	2019-05-01	Kolkata	Banglore	CCU → IXR → BBI → BLR	

	Duration	Total_Stops	Additional_Info	Price	Journey_day	Journey_month	\
0	2h 50m	non-stop	No info	3897	24	3	
1	7h 25m	2 stops	No info	7662	1	5	

	Journey_year	Dep_Time_hour	Dep_Time_minute	Arrival_Time_hour	\
0	2019	22	20	1	
1	2019	5	50	13	

	Arrival_Time_minute	Duration_hours	Duration_mins
0	10	2	50
1	15	7	25

1.10 Analysis: Wheteher Duration Impacts On Price Or Not?

```
[52]: data['Duration']
```

```
[52]: 0      2h 50m  
      1      7h 25m
```

```

2      19h 0m
3      5h 25m
4      4h 45m
...
10678   2h 30m
10679   2h 35m
10680   3h 0m
10681   2h 40m
10682   8h 20m
Name: Duration, Length: 10682, dtype: object

```

```
[53]: 2*60
```

```
[53]: 120
```

```
[54]: '2*60'
```

```
[54]: '2*60'
```

```
[55]: eval('2*60')
```

```
[55]: 120
```

```
[56]: data['Duration'].str.replace('h' , "*60").str.replace(' ' , '+').str.
      ↪replace('m' , "*1")
```

```

[56]: 0      2*60+50*1
      1      7*60+25*1
      2     19*60+0*1
      3      5*60+25*1
      4      4*60+45*1
...
10678   2*60+30*1
10679   2*60+35*1
10680   3*60+0*1
10681   2*60+40*1
10682   8*60+20*1
Name: Duration, Length: 10682, dtype: object

```

```
[57]: data['Duration_total_mins'] = data['Duration'].str.replace('h' , "*60").str.
      ↪replace(' ' , '+').str.replace('m' , "*1").apply(eval)
```

```
[58]: data['Duration_total_mins']
```

```

[58]: 0      170
      1     445
      2    1140
      3     325

```

```

4          285
...
10678      150
10679      155
10680      180
10681      160
10682      500
Name: Duration_total_mins, Length: 10682, dtype: int64

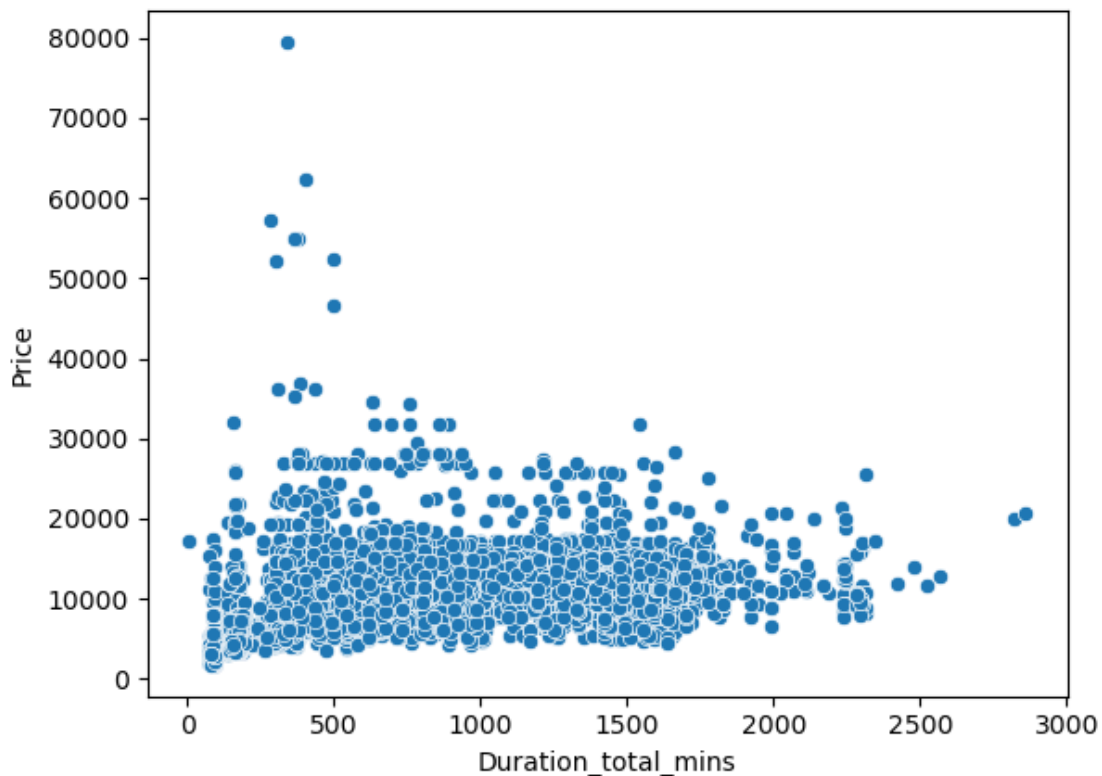
```

```
[59]: data.columns
```

```
[59]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
        'Duration', 'Total_Stops', 'Additional_Info', 'Price', 'Journey_day',
        'Journey_month', 'Journey_year', 'Dep_Time_hour', 'Dep_Time_minute',
        'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours',
        'Duration_mins', 'Duration_total_mins'],
        dtype='object')
```

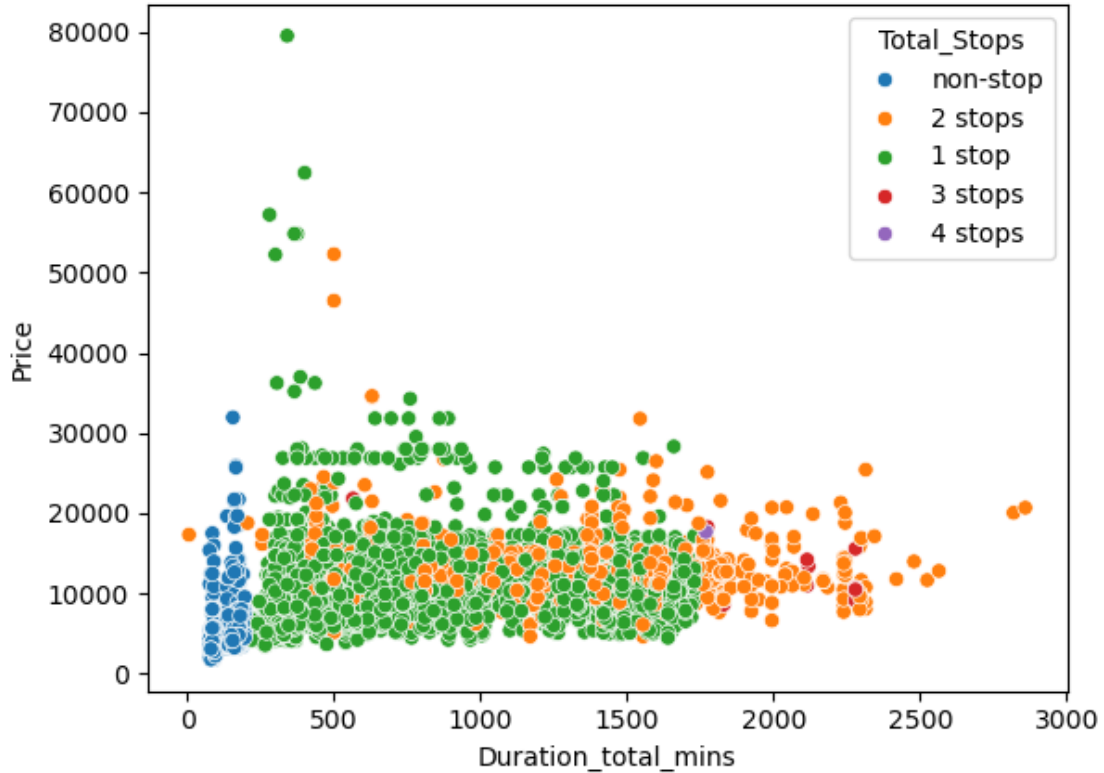
```
[60]: sns.scatterplot(x="Duration_total_mins" , y="Price" , data=data)
```

```
[60]: <Axes: xlabel='Duration_total_mins', ylabel='Price'>
```



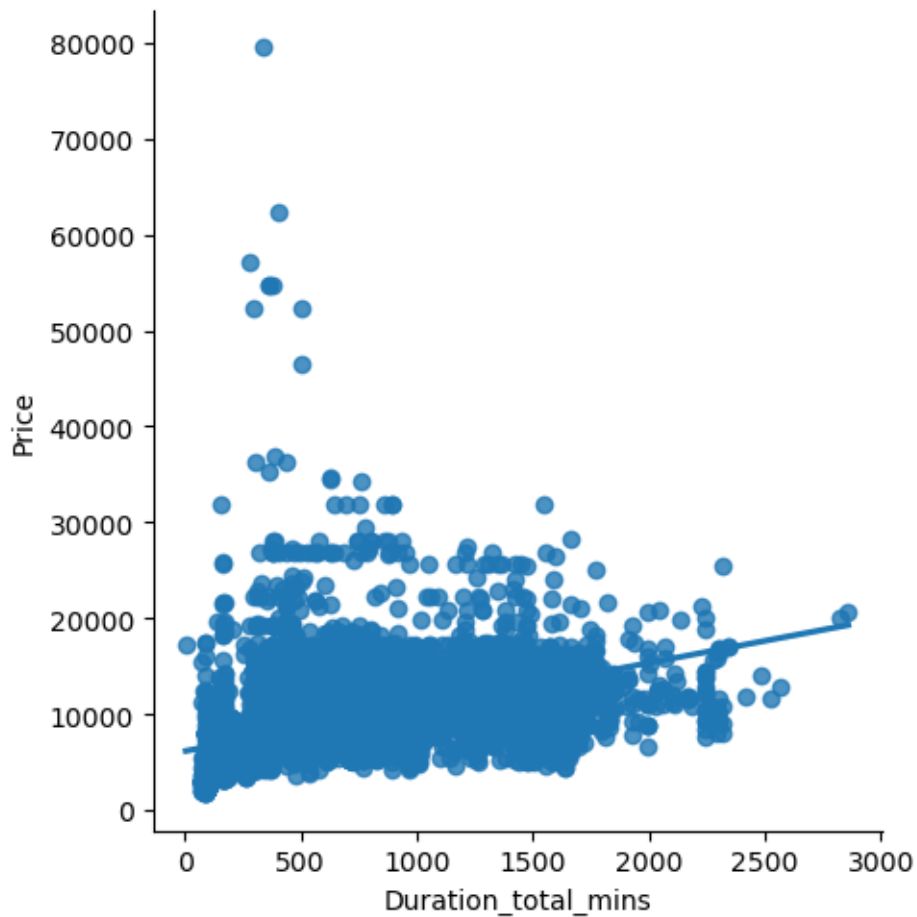
```
[61]: sns.scatterplot(x="Duration_total_mins" , y="Price" , hue="Total_Stops",  
↳ data=data)
```

```
[61]: <Axes: xlabel='Duration_total_mins', ylabel='Price'>
```



```
[62]: sns.lmplot(x="Duration_total_mins" , y="Price" , data=data)
```

```
[62]: <seaborn.axisgrid.FacetGrid at 0x19383ac5430>
```



1.11 Analysis:

1.12 1) On which route Jet Airways is extremely used?

1.13 2) Airline vs Price Analysis

```
[63]: data[data['Airline']=='Jet Airways'].groupby('Route').size().
      ↪sort_values(ascending=False)
```

```
[63]: Route
CCU → BOM → BLR          930
DEL → BOM → COK           875
BLR → BOM → DEL           385
BLR → DEL                 382
CCU → DEL → BLR           300
BOM → HYD                 207
DEL → JAI → BOM → COK     207
DEL → AMD → BOM → COK     141
```

```

DEL → IDR → BOM → COK      86
DEL → NAG → BOM → COK      61
DEL → ATQ → BOM → COK      38
DEL → COK                    34
DEL → BHO → BOM → COK      29
DEL → BDQ → BOM → COK      28
DEL → LKO → BOM → COK      25
DEL → JDH → BOM → COK      23
CCU → GAU → BLR             22
DEL → MAA → BOM → COK      16
DEL → IXC → BOM → COK      13
BLR → MAA → DEL             10
BLR → BDQ → DEL             8
DEL → UDR → BOM → COK       7
BOM → DEL → HYD             5
CCU → BOM → PNQ → BLR       4
BLR → BOM → JDH → DEL       3
DEL → DED → BOM → COK       2
BOM → BDQ → DEL → HYD       2
DEL → CCU → BOM → COK       1
BOM → VNS → DEL → HYD       1
BOM → UDR → DEL → HYD       1
BOM → JDH → DEL → HYD       1
BOM → IDR → DEL → HYD       1
BOM → DED → DEL → HYD       1
dtype: int64

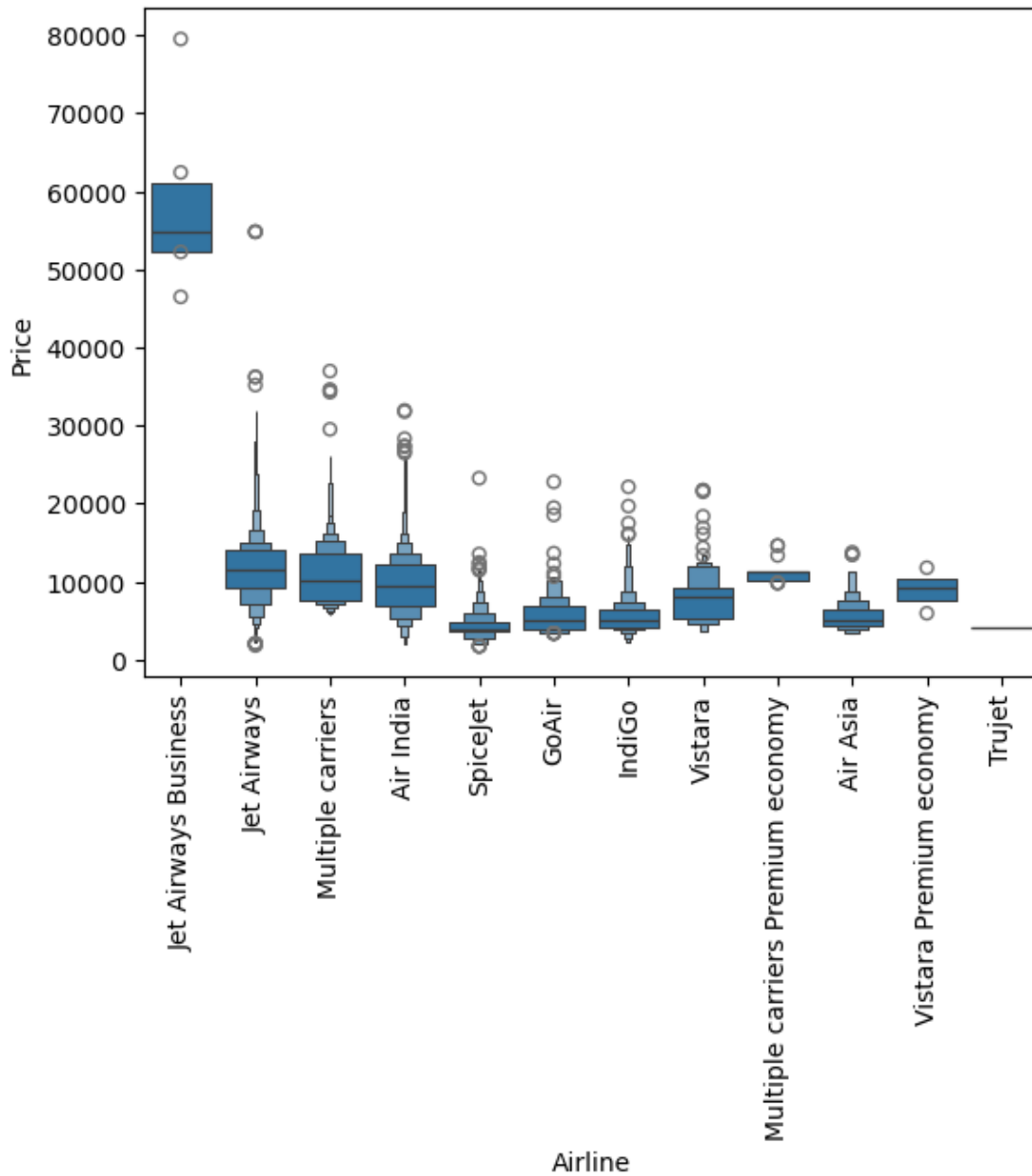
```

```
[64]: data.columns
```

```
[64]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
          'Duration', 'Total_Stops', 'Additional_Info', 'Price', 'Journey_day',
          'Journey_month', 'Journey_year', 'Dep_Time_hour', 'Dep_Time_minute',
          'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours',
          'Duration_mins', 'Duration_total_mins'],
          dtype='object')
```

```
[65]: sns.boxenplot(y='Price' , x='Airline' , data=data.sort_values('Price' ,
↪ascending=False))
plt.xticks(rotation="vertical")
plt.show
```

```
[65]: <function matplotlib.pyplot.show(close=None, block=None)>
```

1.14 Feature Engineering: One Hot Encoding

```
[66]: data.head(2)
```

```
[66]:   Airline Date_of_Journey  Source Destination      Route \
0   IndiGo   2019-03-24  Bangalore   New Delhi      BLR → DEL
1  Air India   2019-05-01   Kolkata   Bangalore  CCU → IXR → BBI → BLR

   Duration Total_Stops Additional_Info  Price  Journey_day  Journey_month \
```

0	2h 50m	non-stop	No info	3897	24	3
1	7h 25m	2 stops	No info	7662	1	5

	Journey_year	Dep_Time_hour	Dep_Time_minute	Arrival_Time_hour	\
0	2019	22	20	1	
1	2019	5	50	13	

	Arrival_Time_minute	Duration_hours	Duration_mins	Duration_total_mins
0	10	2	50	170
1	15	7	25	445

```
[67]: cat_col = [col for col in data.columns if data[col].dtype=="object"]
```

```
[68]: num_col = [col for col in data.columns if data[col].dtype!="object"]
```

```
[69]: cat_col
```

```
[69]: ['Airline',
       'Source',
       'Destination',
       'Route',
       'Duration',
       'Total_Stops',
       'Additional_Info']
```

```
[70]: data['Source'].unique()
```

```
[70]: array(['Bangalore', 'Kolkata', 'Delhi', 'Chennai', 'Mumbai'], dtype=object)
```

```
[71]: data['Source'].apply(lambda x : 1 if x == 'Bangalore' else 0)
```

```
[71]: 0      1
      1      0
      2      0
      3      0
      4      1
      ..
10678    0
10679    0
10680    1
10681    1
10682    0
Name: Source, Length: 10682, dtype: int64
```

```
[72]: for sub_category in data['Source'].unique():
       data['Source_'+sub_category] = data['Source'].apply(lambda x : 1 if x_
↪ ==sub_category else 0)
```

```
[73]: data.head(3)
```

```
[73]:      Airline Date_of_Journey  Source Destination      Route \
0      IndiGo    2019-03-24  Bangalore    New Delhi      BLR → DEL
1      Air India    2019-05-01  Kolkata    Bangalore  CCU → IXR → BBI → BLR
2      Jet Airways    2019-06-09    Delhi    Cochin    DEL → LKO → BOM → COK

      Duration Total_Stops Additional_Info  Price  Journey_day ... \
0      2h 50m    non-stop          No info   3897           24 ...
1      7h 25m      2 stops          No info   7662            1 ...
2     19h 0m      2 stops          No info  13882            9 ...

      Arrival_Time_hour  Arrival_Time_minute  Duration_hours  Duration_mins \
0                      1                    10                2             50
1                     13                    15                7             25
2                      4                    25               19              0

      Duration_total_mins  Source_Banglore  Source_Kolkata  Source_Delhi \
0                      170                1                0             0
1                      445                0                1             0
2                     1140                0                0             1

      Source_Chennai  Source_Mumbai
0                    0              0
1                    0              0
2                    0              0

[3 rows x 24 columns]
```

```
[74]: cat_col
```

```
[74]: ['Airline',
      'Source',
      'Destination',
      'Route',
      'Duration',
      'Total_Stops',
      'Additional_Info']
```

```
[75]: data.head(2)
```

```
[75]:      Airline Date_of_Journey  Source Destination      Route \
0      IndiGo    2019-03-24  Bangalore    New Delhi      BLR → DEL
1      Air India    2019-05-01  Kolkata    Bangalore  CCU → IXR → BBI → BLR

      Duration Total_Stops Additional_Info  Price  Journey_day ... \
0      2h 50m    non-stop          No info   3897           24 ...
```

```

1    7h 25m    2 stops    No info    7662    1 ...

Arrival_Time_hour  Arrival_Time_minute  Duration_hours  Duration_mins  \
0                  1                   10              2         50
1                 13                   15              7         25

Duration_total_mins  Source_Banglore  Source_Kolkata  Source_Delhi  \
0                  170                1                0         0
1                 445                0                1         0

Source_Chennai  Source_Mumbai
0                0             0
1                0             0

[2 rows x 24 columns]

```

```
[76]: data['Airline'].nunique()
```

```
[76]: 12
```

```
[77]: data['Airline'].unique()
```

```
[77]: array(['IndiGo', 'Air India', 'Jet Airways', 'SpiceJet',
        'Multiple carriers', 'GoAir', 'Vistara', 'Air Asia',
        'Vistara Premium economy', 'Jet Airways Business',
        'Multiple carriers Premium economy', 'Trujet'], dtype=object)
```

1.15 Optimized Encoding: Target Guided Encoding

```
[78]: data.groupby(['Airline'])['Price'].mean().sort_values()
```

```
[78]: Airline
Trujet                4140.000000
SpiceJet              4338.284841
Air Asia              5590.260188
IndiGo                5673.682903
GoAir                 5861.056701
Vistara               7796.348643
Vistara Premium economy  8962.333333
Air India             9612.427756
Multiple carriers     10902.678094
Multiple carriers Premium economy 11418.846154
Jet Airways           11643.923357
Jet Airways Business  58358.666667
Name: Price, dtype: float64
```

```
[79]: airlines = data.groupby(['Airline'])['Price'].mean().sort_values().index
```

```
[80]: airlines
```

```
[80]: Index(['Trujet', 'SpiceJet', 'Air Asia', 'IndiGo', 'GoAir', 'Vistara',  
         'Vistara Premium economy', 'Air India', 'Multiple carriers',  
         'Multiple carriers Premium economy', 'Jet Airways',  
         'Jet Airways Business'],  
         dtype='object', name='Airline')
```

1.16 Enumeration

```
[81]: dict_airlines = {key:index for index , key in enumerate(airlines , 0)}
```

```
[82]: dict_airlines
```

```
[82]: {'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}
```

```
[83]: data['Airline'] = data['Airline'].map(dict_airlines)
```

```
[84]: data['Airline']
```

```
[84]: 0      3  
      1      7  
      2     10  
      3      3  
      4      3  
      ..  
    10678     2  
    10679     7  
    10680    10  
    10681     5  
    10682     7  
      Name: Airline, Length: 10682, dtype: int64
```

```
[85]: data.head(3)
```

```
[85]:   Airline Date_of_Journey  Source Destination      Route \  
0      3      2019-03-24  Bangalore   New Delhi  BLR → DEL
```

1	7	2019-05-01	Kolkata	Banglore	CCU → IXR → BBI → BLR
2	10	2019-06-09	Delhi	Cochin	DEL → LKO → BOM → COK

	Duration	Total_Stops	Additional_Info	Price	Journey_day	...	\
0	2h 50m	non-stop	No info	3897	24	...	
1	7h 25m	2 stops	No info	7662	1	...	
2	19h 0m	2 stops	No info	13882	9	...	

	Arrival_Time_hour	Arrival_Time_minute	Duration_hours	Duration_mins	\
0	1	10	2	50	
1	13	15	7	25	
2	4	25	19	0	

	Duration_total_mins	Source_Banglore	Source_Kolkata	Source_Delhi	\
0	170	1	0	0	
1	445	0	1	0	
2	1140	0	0	1	

	Source_Chennai	Source_Mumbai
0	0	0
1	0	0
2	0	0

[3 rows x 24 columns]

```
[86]: data['Destination'].unique()
```

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

```
[87]: data['Destination'].replace('New Delhi' , 'Delhi' , inplace=True)
```

```
[88]: data['Destination'].unique()
```

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

```
[89]: dest = data.groupby(['Destination'])['Price'].mean().sort_values().index
```

```
[90]: dest
```

```
[90]: Index(['Kolkata', 'Hyderabad', 'Delhi', 'Banglore', 'Cochin'], dtype='object',
      name='Destination')
```

```
[91]: dict_dest = {key:index for index , key in enumerate(dest , 0)}
```

```
[92]: dict_dest
```

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

```
[93]: data['Destination'] = data['Destination'].map(dict_dest)
```

```
[94]: data['Destination']
```

```
[94]: 0      2
      1      3
      2      4
      3      3
      4      2
      ..
10678    3
10679    3
10680    2
10681    2
10682    4
Name: Destination, Length: 10682, dtype: int64
```

```
[95]: data.head(3)
```

```
[95]:   Airline Date_of_Journey  Source Destination  Route \
0      3      2019-03-24  Banglore          2      BLR → DEL
1      7      2019-05-01  Kolkata          3  CCU → IXR → BBI → BLR
2     10      2019-06-09    Delhi          4  DEL → LKO → BOM → COK

   Duration Total_Stops Additional_Info  Price  Journey_day  ... \
0    2h 50m    non-stop          No info   3897          24  ...
1    7h 25m     2 stops          No info   7662           1  ...
2   19h 0m     2 stops          No info  13882           9  ...

   Arrival_Time_hour  Arrival_Time_minute  Duration_hours  Duration_mins \
0              1              10              2              50
1             13              15              7              25
2              4              25             19              0

   Duration_total_mins  Source_Banglore  Source_Kolkata  Source_Delhi \
0              170              1              0              0
1              445              0              1              0
2             1140              0              0              1

   Source_Chennai  Source_Mumbai
0              0              0
1              0              0
2              0              0

[3 rows x 24 columns]
```

1.17 Manual Label Encoding

```
[96]: data['Total_Stops']
```

```
[96]: 0      non-stop
      1      2 stops
      2      2 stops
      3      1 stop
      4      1 stop
      ...
     10678    non-stop
     10679    non-stop
     10680    non-stop
     10681    non-stop
     10682     2 stops
      Name: Total_Stops, Length: 10682, dtype: object
```

```
[97]: data['Total_Stops'].unique()
```

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

```
[98]: stop = {'non-stop':0, '2 stops':2, '1 stop':1, '3 stops':3, '4 stops':4}
```

```
[99]: data['Total_Stops'] = data['Total_Stops'].map(stop)
```

```
[100]: data['Total_Stops']
```

```
[100]: 0      0
      1      2
      2      2
      3      1
      4      1
      ..
     10678    0
     10679    0
     10680    0
     10681    0
     10682     2
      Name: Total_Stops, Length: 10682, dtype: int64
```

```
[101]: data.columns
```

```
[101]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
      'Duration', 'Total_Stops', 'Additional_Info', 'Price', 'Journey_day',
      'Journey_month', 'Journey_year', 'Dep_Time_hour', 'Dep_Time_minute',
      'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours',
      'Duration_mins', 'Duration_total_mins', 'Source_Bangalore',
```



```
'Source_Kolkata', 'Source_Delhi', 'Source_Chennai', 'Source_Mumbai'],
dtype='object')
```

```
[102]: data['Additional_Info'].value_counts()/len(data)*100
```

```
[102]: Additional_Info
No info          78.112713
In-flight meal not included  18.554578
No check-in baggage included  2.995694
1 Long layover    0.177869
Change airports   0.065531
Business class    0.037446
No Info           0.028085
1 Short layover   0.009362
Red-eye flight    0.009362
2 Long layover    0.009362
Name: count, dtype: float64
```

```
[103]: data.head(4)
```

```
[103]:   Airline  Date_of_Journey  Source  Destination  Route \
0        3    2019-03-24  Bangalore          2    BLR → DEL
1        7    2019-05-01  Kolkata          3  CCU → IXR → BBI → BLR
2       10    2019-06-09    Delhi          4  DEL → LKO → BOM → COK
3        3    2019-05-12  Kolkata          3    CCU → NAG → BLR
```

```
   Duration  Total_Stops  Additional_Info  Price  Journey_day  ... \
0    2h 50m           0        No info    3897           24  ...
1    7h 25m           2        No info    7662            1  ...
2   19h 0m           2        No info   13882            9  ...
3    5h 25m           1        No info    6218           12  ...
```

```
   Arrival_Time_hour  Arrival_Time_minute  Duration_hours  Duration_mins \
0                1                10                2                50
1               13                15                7                25
2                4                25               19                0
3               23                30                5                25
```

```
   Duration_total_mins  Source_Banglore  Source_Kolkata  Source_Delhi \
0                170                1                0                0
1                445                0                1                0
2               1140                0                0                1
3                325                0                1                0
```

```
   Source_Chennai  Source_Mumbai
0                0                0
1                0                0
```

```

2          0          0
3          0          0

```

[4 rows x 24 columns]

```
[104]: data.columns
```

```
[104]: Index(['Airline', 'Date_of_Journey', 'Source', 'Destination', 'Route',
          'Duration', 'Total_Stops', 'Additional_Info', 'Price', 'Journey_day',
          'Journey_month', 'Journey_year', 'Dep_Time_hour', 'Dep_Time_minute',
          'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours',
          'Duration_mins', 'Duration_total_mins', 'Source_Banglore',
          'Source_Kolkata', 'Source_Delhi', 'Source_Chennai', 'Source_Mumbai'],
          dtype='object')
```

```
[105]: data.drop(columns=['Date_of_Journey' , 'Additional_Info' , 'Source' ,
          ↪ 'Journey_year'] , axis=1 , inplace=True)
```

```
[106]: data.drop(columns=['Route'] , axis=1 , inplace=True)
```

```
[107]: data.columns
```

```
[107]: Index(['Airline', 'Destination', 'Duration', 'Total_Stops', 'Price',
          'Journey_day', 'Journey_month', 'Dep_Time_hour', 'Dep_Time_minute',
          'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours',
          'Duration_mins', 'Duration_total_mins', 'Source_Banglore',
          'Source_Kolkata', 'Source_Delhi', 'Source_Chennai', 'Source_Mumbai'],
          dtype='object')
```

```
[108]: data.columns
```

```
[108]: Index(['Airline', 'Destination', 'Duration', 'Total_Stops', 'Price',
          'Journey_day', 'Journey_month', 'Dep_Time_hour', 'Dep_Time_minute',
          'Arrival_Time_hour', 'Arrival_Time_minute', 'Duration_hours',
          'Duration_mins', 'Duration_total_mins', 'Source_Banglore',
          'Source_Kolkata', 'Source_Delhi', 'Source_Chennai', 'Source_Mumbai'],
          dtype='object')
```

```
[109]: data.head(3)
```

```
[109]:
```

	Airline	Destination	Duration	Total_Stops	Price	Journey_day \
0	3	2	2h 50m	0	3897	24
1	7	3	7h 25m	2	7662	1
2	10	4	19h 0m	2	13882	9

	Journey_month	Dep_Time_hour	Dep_Time_minute	Arrival_Time_hour \
0	3	22	20	1
1	5	5	50	13

	2	6	9	25	4
--	---	---	---	----	---

	Arrival_Time_minute	Duration_hours	Duration_mins	Duration_total_mins	\
0	10	2	50	170	
1	15	7	25	445	
2	25	19	0	1140	

	Source_Banglore	Source_Kolkata	Source_Delhi	Source_Chennai	\
0	1	0	0	0	
1	0	1	0	0	
2	0	0	1	0	

	Source_Mumbai
0	0
1	0
2	0

```
[110]: data.drop(columns=['Duration_total_mins'] , axis=1 , inplace=True)
```

```
[111]: data.drop(columns=['Duration'] , axis=1 , inplace=True)
```

```
[112]: data.head(3)
```

```
[112]:
```

	Airline	Destination	Total_Stops	Price	Journey_day	Journey_month	\
0	3	2	0	3897	24	3	
1	7	3	2	7662	1	5	
2	10	4	2	13882	9	6	

	Dep_Time_hour	Dep_Time_minute	Arrival_Time_hour	Arrival_Time_minute	\
0	22	20	1	10	
1	5	50	13	15	
2	9	25	4	25	

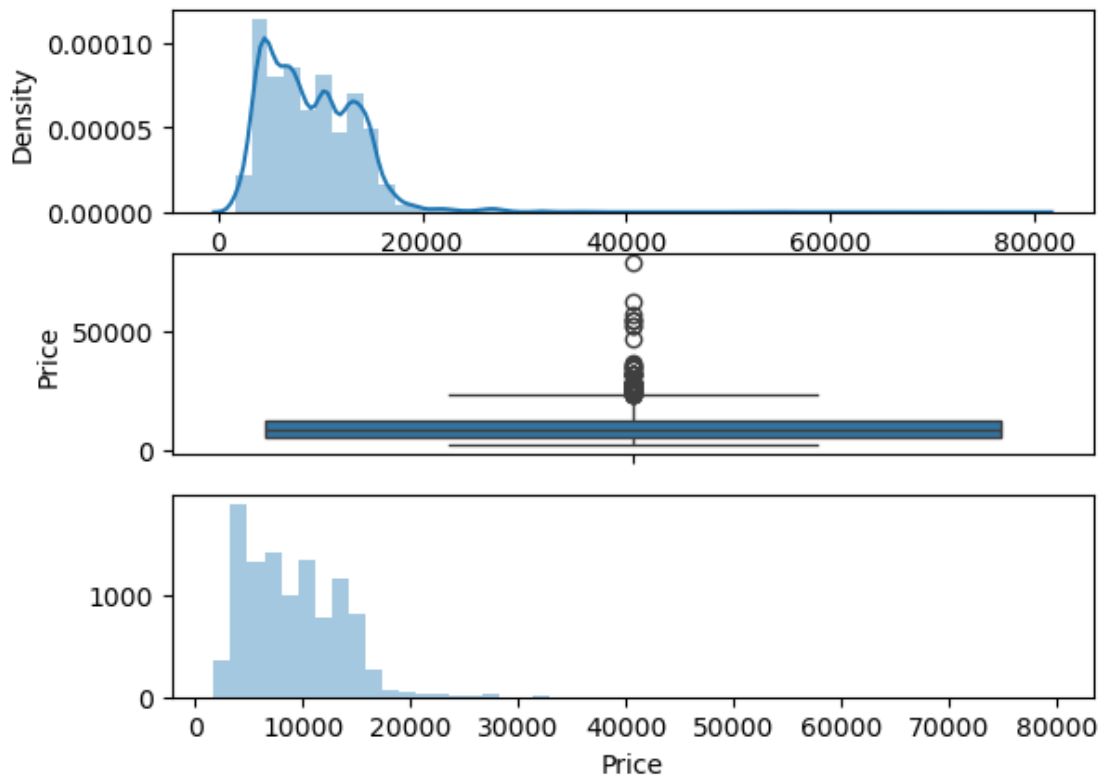
	Duration_hours	Duration_mins	Source_Banglore	Source_Kolkata	\
0	2	50	1	0	
1	7	25	0	1	
2	19	0	0	0	

	Source_Delhi	Source_Chennai	Source_Mumbai
0	0	0	0
1	0	0	0
2	1	0	0

1.18 Outlier Detection

```
[113]: def plot(df , col):  
        fig , (ax1 , ax2 , ax3) = plt.subplots(3,1)  
  
        sns.distplot(df[col] , ax=ax1)  
        sns.boxplot(df[col] , ax=ax2)  
        sns.distplot(df[col] , ax=ax3 , kde=False)
```

```
[114]: plot(data , 'Price')
```



```
[115]: q1 = data['Price'].quantile(0.25)  
        q3 = data['Price'].quantile(0.75)  
  
        iqr = q3 - q1  
  
        maximum = q3 + 1.5*iqr  
        minimum = q1 - 1.5*iqr
```

```
[116]: print(maximum)
```

23017.0

```
[117]: print(minimum)
```

-5367.0

```
[118]: print([price for price in data['Price'] if price>maximum or price<minimum])
```

[27430, 36983, 26890, 26890, 25139, 27210, 52229, 26743, 26890, 25735, 27992, 26890, 26890, 23583, 26890, 23533, 24115, 25735, 54826, 31783, 27992, 26890, 26890, 25430, 36235, 27210, 26890, 25735, 54826, 26890, 35185, 79512, 28097, 27992, 26890, 25735, 26092, 31825, 25913, 25735, 27992, 31825, 23267, 62427, 54826, 31825, 25430, 26890, 36235, 23843, 26890, 25735, 28322, 25735, 25735, 31825, 26890, 27992, 34273, 46490, 29528, 26890, 26890, 26890, 34503, 26890, 27992, 26890, 26890, 23170, 24528, 26890, 27992, 25735, 34608, 25703, 26890, 23528, 31825, 27282, 25735, 27992, 52285, 24017, 31945, 26890, 24318, 23677, 27992, 24210, 57209, 26890, 31825, 26480]

```
[119]: len([price for price in data['Price'] if price>maximum or price<minimum])
```

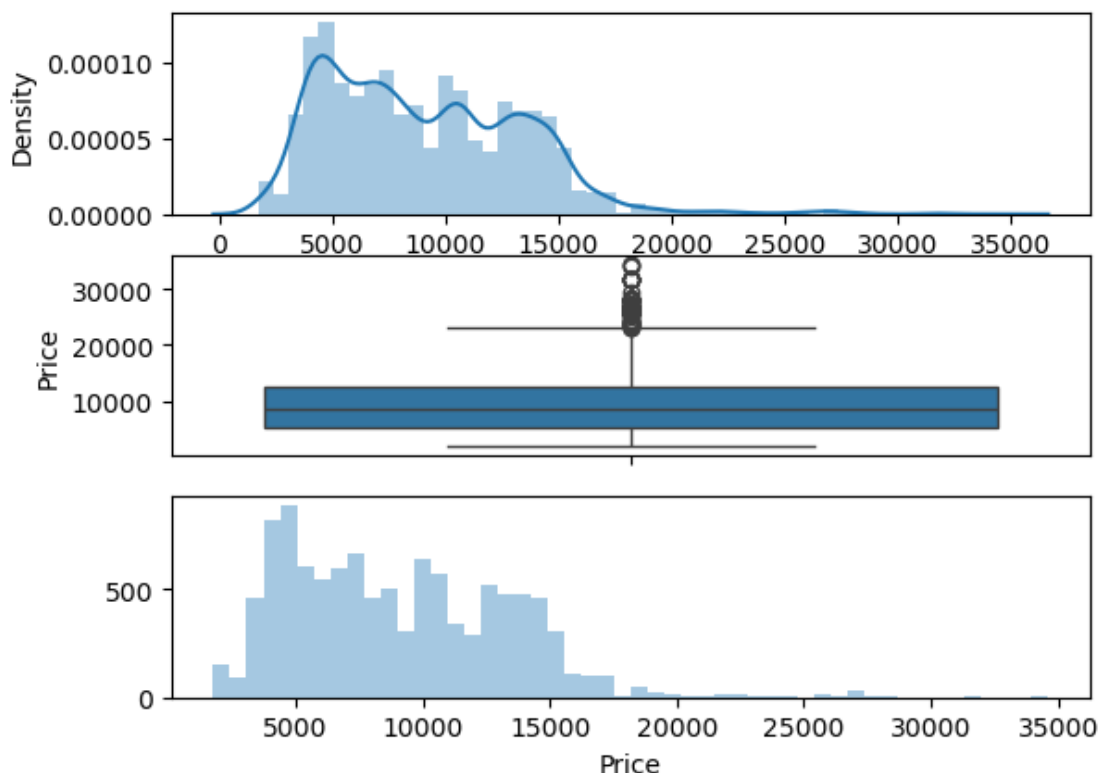
```
[119]: 94
```

```
[120]: data['Price'] = np.where(data['Price']>=35000 , data['Price'].median() ,  
    ↪data['Price'] )
```

```
[121]: data['Price']
```

```
[121]: 0      3897.0  
      1      7662.0  
      2     13882.0  
      3      6218.0  
      4     13302.0  
      ...  
    10678     4107.0  
    10679     4145.0  
    10680     7229.0  
    10681    12648.0  
    10682    11753.0  
      Name: Price, Length: 10682, dtype: float64
```

```
[122]: plot(data , 'Price')
```



1.19 Feature Selection

```
[123]: X = data.drop(['Price'] , axis=1)
```

```
[124]: y = data['Price']
```

```
[125]: from sklearn.feature_selection import mutual_info_regression
```

```
[126]: imp = mutual_info_regression(X , y)
```

```
[127]: imp
```

```
[127]: array([1.32242588, 1.06873428, 0.78609461, 0.38237418, 0.61839087,
          0.93453414, 0.75535656, 1.13792216, 0.89756616, 1.12525252,
          0.67404378, 0.39688153, 0.45552395, 0.51221784, 0.13437008,
          0.19534493])
```

```
[128]: imp_df = pd.DataFrame(imp , index=X.columns)
```

```
[129]: imp_df.columns = ['importance']
```

```
[130]: imp_df
```

```
[130]:
```

	importance
Airline	1.322426
Destination	1.068734
Total_Stops	0.786095
Journey_day	0.382374
Journey_month	0.618391
Dep_Time_hour	0.934534
Dep_Time_minute	0.755357
Arrival_Time_hour	1.137922
Arrival_Time_minute	0.897566
Duration_hours	1.125253
Duration_mins	0.674044
Source_Banglore	0.396882
Source_Kolkata	0.455524
Source_Delhi	0.512218
Source_Chennai	0.134370
Source_Mumbai	0.195345

```
[131]: imp_df.sort_values(by='importance' , ascending=False)
```

```
[131]:
```

	importance
Airline	1.322426
Arrival_Time_hour	1.137922
Duration_hours	1.125253
Destination	1.068734
Dep_Time_hour	0.934534
Arrival_Time_minute	0.897566
Total_Stops	0.786095
Dep_Time_minute	0.755357
Duration_mins	0.674044
Journey_month	0.618391
Source_Delhi	0.512218
Source_Kolkata	0.455524
Source_Banglore	0.396882
Journey_day	0.382374
Source_Mumbai	0.195345
Source_Chennai	0.134370

1.20 Machine Learning Model Building and Saving (Regression, Classification, Clustering)

```
[132]: from sklearn.model_selection import train_test_split
```

```
[133]: X_train, X_test, y_train, y_test = train_test_split(
        X, y, test_size=0.33, random_state=42)
```

```
[134]: from sklearn.ensemble import RandomForestRegressor
```

```
[135]: ml_model = RandomForestRegressor()

[136]: ml_model.fit(X_train , y_train)

[136]: RandomForestRegressor()

[137]: y_pred = ml_model.predict(X_test)

[138]: y_pred

[138]: array([16723.03 , 5339.6 , 8998.62 , ..., 8166.02 , 9107.56 ,
          11679.555])

[139]: from sklearn import metrics

[140]: metrics.r2_score(y_test , y_pred)

[140]: 0.8172808853427979
```

1.21 Saving Model

```
[141]: import pickle

[142]: file = open(r'E:\DS Projects\1. Predict Fare of Airlines Tickets using Machine_
↳ Learning\rf_random.pkl' , 'wb')

[143]: pickle.dump(ml_model , file)

[144]: model = open(r'E:\DS Projects\1. Predict Fare of Airlines Tickets using Machine_
↳ Learning\rf_random.pkl' , 'rb')

[145]: forest = pickle.load(model)

[146]: y_pred2 = forest.predict(X_test)

[147]: y_pred2

[147]: array([16723.03 , 5339.6 , 8998.62 , ..., 8166.02 , 9107.56 ,
          11679.555])

[148]: metrics.r2_score(y_test , y_pred2)

[148]: 0.8172808853427979
```


1.22 Define Evaluation Metric and Automate Machine Learning Pipeline

```
[149]: def mape(y_true , y_pred):  
        y_true , y_pred = np.array(y_true) , np.array(y_pred)  
        return np.mean(np.abs((y_true - y_pred) / y_true)) * 100
```

```
[150]: mape(y_test , y_pred)
```

```
[150]: 13.124320812886314
```

```
[152]: from sklearn import metrics
```

```
[156]: def predict(ml_model):  
        model = ml_model.fit(X_train, y_train) # Fix indentation here  
        print('Training score : {}'.format(model.score(X_train, y_train)))  
        y_prediction = model.predict(X_test)  
        print('Predictions are: {}'.format(y_prediction))  
        print('\n')  
        r2_score = metrics.r2_score(y_test, y_prediction) # Fix indentation and  
        ↪ parentheses here  
        print('R2 score : {}'.format(r2_score))  
        print('MAE : {}'.format(metrics.mean_absolute_error(y_test, y_prediction)))  
        print('MSE : {}'.format(metrics.mean_squared_error(y_test, y_prediction)))  
        print('RMSE : {}'.format(np.sqrt(metrics.mean_squared_error(y_test, ↪  
        ↪ y_prediction))))  
        print('MAPE : {}'.format(mape(y_test, y_prediction)))  
        sns.displot(y_test - y_prediction)
```

```
[157]: predict(RandomForestRegressor())
```

Training score : 0.9543029543825003

Predictions are: [16671.58 5337.8 8830.04 ... 8037.09
 9220.47 12307.16866667]

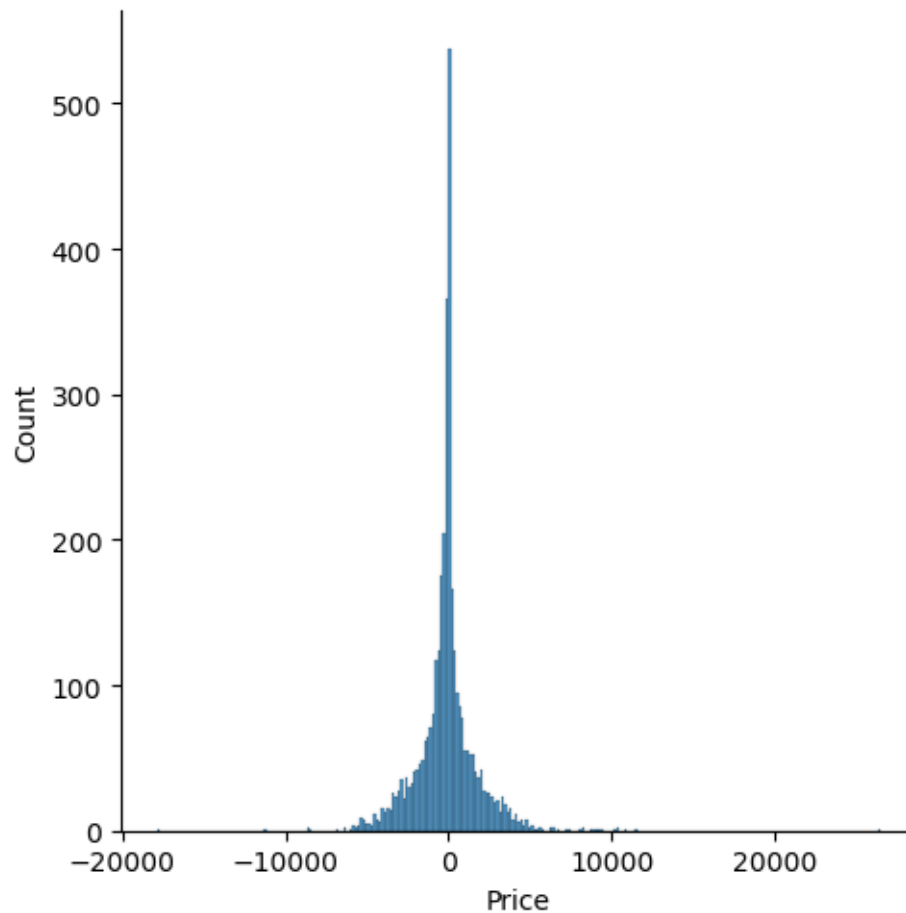
R2 score : 0.8172578862267197

MAE : 1164.9011671313656

MSE : 3509328.433285369

RMSE : 1873.3201630488497

MAPE : 13.114686931362417



```
[158]: from sklearn.tree import DecisionTreeRegressor
```

```
[159]: predict(DecisionTreeRegressor())
```

Training score : 0.9696998040730191

Predictions are: [16840. 4959. 8085. ... 6442. 10141. 11652.5]

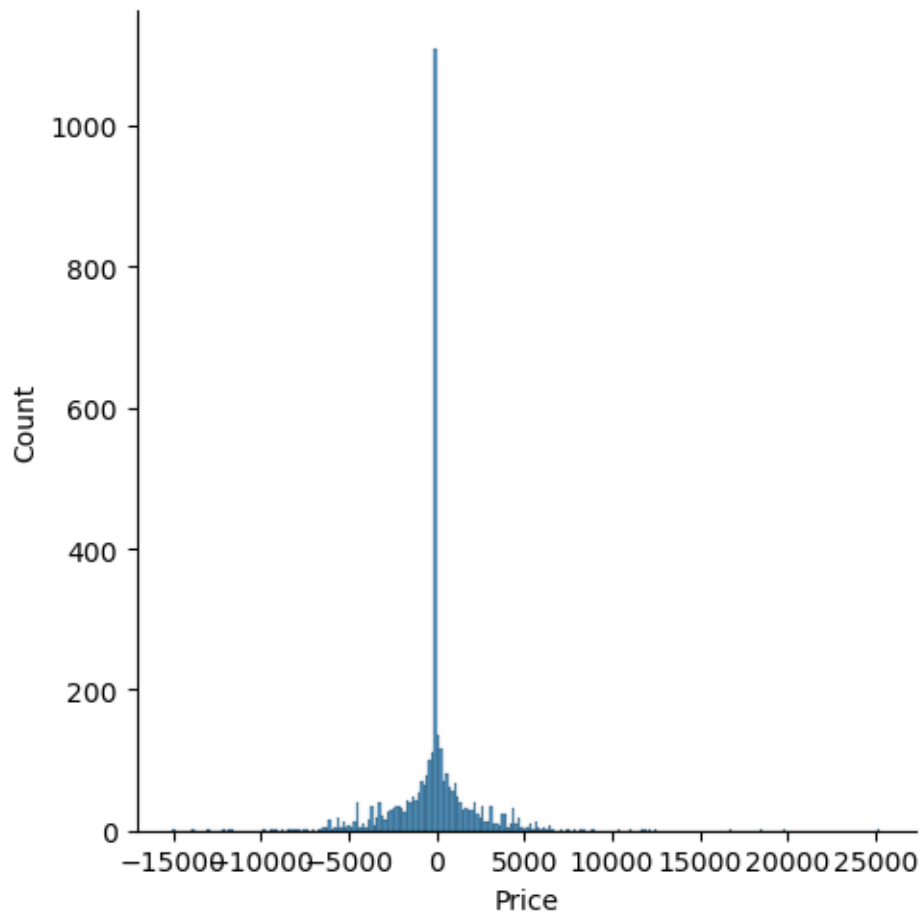
R2 score : 0.6904468907030217

MAE : 1397.1093590470787

MSE : 5944571.317674103

RMSE : 2438.1491582087638

MAPE : 15.538578369246133



1.23 Hypertune ML Model (Hyperparameter Optimization)

```
[194]: from sklearn.ensemble import RandomForestRegressor
from sklearn.model_selection import RandomizedSearchCV
import numpy as np

# Create an instance of RandomForestRegressor
reg_rf = RandomForestRegressor()

# Define parameter grid
n_estimators = [int(x) for x in np.linspace(start=100, stop=1200, num=6)]
max_features = ["auto", "sqrt"]
max_depth = [int(x) for x in np.linspace(start=5, stop=30, num=4)]
min_samples_split = [5, 10, 15, 100]

random_grid = {
    'n_estimators': n_estimators,
    'max_features': max_features,
```

```

    'max_depth': max_depth,
    'min_samples_split': min_samples_split
}

# Create the RandomizedSearchCV instance with the reg_rf instance
rf_random = RandomizedSearchCV(estimator=reg_rf,
    ↪param_distributions=random_grid, cv=3, n_jobs=-1, verbose=2)

# Fit the model
rf_random.fit(X_train, y_train)

```

Fitting 3 folds for each of 10 candidates, totalling 30 fits

```

[194]: RandomizedSearchCV(cv=3, estimator=RandomForestRegressor(), n_jobs=-1,
    param_distributions={'max_depth': [5, 13, 21, 30],
    'max_features': ['auto', 'sqrt'],
    'min_samples_split': [5, 10, 15, 100],
    'n_estimators': [100, 320, 540, 760,
    980, 1200]},
    verbose=2)

```

```

[195]: rf_random.best_params_

```

```

[195]: {'n_estimators': 980,
    'min_samples_split': 5,
    'max_features': 'sqrt',
    'max_depth': 13}

```

```

[197]: rf_random.best_estimator_

```

```

[197]: RandomForestRegressor(max_depth=13, max_features='sqrt', min_samples_split=5,
    n_estimators=980)

```

```

[198]: rf_random.best_score_

```

```

[198]: 0.7978165042231741

```