```
# Title = 1.Predict the price of the Uber ride from a given pickup
point to the agreed drop-off location.
# Name = Tanmay Shrikrishna Badhe
# Div = B
# Roll No. 01
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
df = pd.read csv('uber.csv')
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200000 entries, 0 to 199999
Data columns (total 9 columns):
#
     Column
                        Non-Null Count
                                          Dtype
_ _ _
 0
                        200000 non-null
                                          int64
     Unnamed: 0
1
     key
                        200000 non-null
                                          object
 2
     fare amount
                        200000 non-null
                                          float64
 3
     pickup datetime
                        200000 non-null
                                          obiect
 4
     pickup longitude
                        200000 non-null
                                          float64
 5
     pickup_latitude
                        200000 non-null
                                          float64
 6
     dropoff longitude 199999 non-null
                                          float64
 7
     dropoff latitude
                        199999 non-null
                                          float64
8
     passenger count
                        200000 non-null
                                          int64
dtypes: float64(5), int64(2), object(2)
memory usage: 13.7+ MB
df.shape
(200000, 9)
df.head()
                                               fare amount \
   Unnamed: 0
                                          key
0
                 2015-05-07 19:52:06.0000003
     24238194
                                                       7.5
                 2009-07-17 20:04:56.0000002
                                                       7.7
1
     27835199
2
     44984355
                2009-08-24 21:45:00.00000061
                                                      12.9
3
                 2009-06-26 08:22:21.0000001
                                                       5.3
     25894730
4
               2014-08-28 17:47:00.000000188
     17610152
                                                      16.0
                            pickup longitude
                                               pickup latitude \
           pickup datetime
  2015-05-07 19:52:06 UTC
                                   -73.999817
                                                     40.738354
   2009-07-17 20:04:56 UTC
                                                     40.728225
1
                                   -73.994355
2
  2009-08-24 21:45:00 UTC
                                   -74.005043
                                                     40.740770
  2009-06-26 08:22:21 UTC
                                   -73.976124
                                                     40.790844
   2014-08-28 17:47:00 UTC
                                   -73.925023
                                                     40.744085
```

1 -7 2 -7 3 -7	longitud 73.99951 73.99471 73.96256 73.96531 73.97308	2 0 5 6	off_latitude 40.723217 40.750325 40.772647 40.803349 40.761247	passenger_count 1 1 1 3 5		
df.isnull()						
pickup_longs			fare_amount False	<pre>pickup_datetime False</pre>		
False	False	False	False	False		
False 2 False	False	False	False	False		
3 False	False	False	False	False		
4 False	False	False	False	False		
199995 False	False	False	False	False		
199996 False	False	False	False	False		
199997 False	False	False	False	False		
199998 False	False	False	False	False		
199999 False	False	False	False	False		
	kun lati	tude d	ropoff_longit	ude dropoff_lati	tude	
passenger_co	ount	alse	_	· —	alse	
False		alse			alse	
False 2					False	
False	False False				False	
False						
4 False	Г	alse	Га	lse F	alse	
100005	_	-1	_	1		
199995	F	alse	Fa	lse F	alse	

```
False
199996
                  False
                                      False
                                                         False
False
199997
                  False
                                      False
                                                         False
False
199998
                  False
                                      False
                                                         False
False
199999
                  False
                                      False
                                                         False
False
[200000 rows x 9 columns]
df.drop(columns=["Unnamed: 0", "key"], inplace=True)
df.head()
   fare amount
                        pickup_datetime
                                          pickup longitude
pickup latitude \
           7.5 2015-05-07 19:52:06 UTC
                                                -73.999817
40.738354
           7.7
                2009-07-17 20:04:56 UTC
                                                -73.994355
1
40.728225
          12.9
                2009-08-24 21:45:00 UTC
                                                -74,005043
40.740770
3
           5.3
                2009-06-26 08:22:21 UTC
                                                -73.976124
40.790844
                2014-08-28 17:47:00 UTC
                                                -73.925023
          16.0
40.744085
   dropoff longitude
                      dropoff latitude
                                         passenger count
0
          -73.999512
                              40.723217
                                                        1
1
          -73.994710
                              40.750325
                                                        1
2
          -73.962565
                              40.772647
                                                        1
3
                                                        3
          -73.965316
                              40.803349
                                                        5
4
          -73.973082
                              40.761247
df.isnull().sum()
fare amount
                     0
                     0
pickup datetime
pickup_longitude
                     0
pickup latitude
                     0
dropoff longitude
                     1
dropoff latitude
                     1
passenger_count
                     0
dtype: int64
df['dropoff latitude'].fillna(value=df['dropoff latitude'].mean(),inpl
ace = True)
df['dropoff longitude'].fillna(value=df['dropoff longitude'].median(),
inplace = True)
```

```
df.dtypes
fare amount
                      float64
pickup datetime
                       object
pickup longitude
                      float64
pickup latitude
                      float64
dropoff_longitude
                      float64
dropoff latitude
                      float64
passenger count
                        int64
dtype: object
df.pickup datetime = pd.to datetime(df.pickup datetime)
df.dtypes
fare amount
                                  float64
pickup datetime
                      datetime64[ns, UTC]
pickup_longitude
                                  float64
pickup latitude
                                  float64
dropoff_longitude
                                  float64
dropoff latitude
                                  float64
passenger count
                                    int64
dtype: object
df = df.assign(hour = df.pickup datetime.dt.hour,
               day = df.pickup datetime.dt.day,
               month = df.pickup datetime.dt.month,
               year = df.pickup datetime.dt.year,
               dayofweek = df.pickup datetime.dt.dayofweek)
df
        fare amount
                               pickup datetime
                                                 pickup longitude \
0
                7.5 2015-05-07 19:52:06+00:00
                                                       -73.999817
1
                7.7 2009-07-17 20:04:56+00:00
                                                       -73.994355
2
               12.9 2009-08-24 21:45:00+00:00
                                                       -74.005043
3
                5.3 2009-06-26 08:22:21+00:00
                                                       -73.976124
4
               16.0 2014-08-28 17:47:00+00:00
                                                       -73.925023
                3.0 2012-10-28 10:49:00+00:00
                                                       -73.987042
199995
199996
                7.5 2014-03-14 01:09:00+00:00
                                                       -73.984722
               30.9 2009-06-29 00:42:00+00:00
                                                       -73.986017
199997
               14.5 2015-05-20 14:56:25+00:00
                                                       -73.997124
199998
199999
               14.1 2010-05-15 04:08:00+00:00
                                                       -73.984395
        pickup latitude dropoff longitude dropoff latitude
passenger count \
              40.738354
                                 -73.999512
                                                     40.723217
0
1
1
              40.728225
                                 -73.994710
                                                     40.750325
1
2
              40.740770
                                 -73.962565
                                                     40.772647
```

```
1
3
               40.790844
                                   -73.965316
                                                        40.803349
3
4
               40.744085
                                   -73.973082
                                                        40.761247
5
. . .
199995
               40.739367
                                   -73.986525
                                                        40.740297
1
199996
               40.736837
                                   -74.006672
                                                        40.739620
199997
               40.756487
                                   -73.858957
                                                        40.692588
199998
                                   -73.983215
               40.725452
                                                        40.695415
199999
               40.720077
                                   -73.985508
                                                        40.768793
        hour
                    month
                                   dayofweek
               day
                            year
0
           19
                            2015
                 7
                         5
                         7
                                            4
1
           20
                17
                            2009
2
                                            0
           21
                24
                         8
                            2009
3
            8
                26
                         6
                            2009
                                            4
4
                         8
                                            3
           17
                            2014
                28
199995
           10
                28
                        10
                            2012
                                            6
                         3
                            2014
                                            4
199996
            1
                14
199997
            0
                29
                         6
                            2009
                                            0
                         5
                                            2
199998
                            2015
           14
                20
199999
            4
                15
                         5
                            2010
[200000 rows x 12 columns]
df = df.drop(["pickup datetime"], axis =1)
df
        fare_amount pickup_longitude pickup_latitude
dropoff_longitude \
                 7.5
                             -73.999817
                                                 40.738354
73.999512
                 7.7
                             -73.994355
                                                 40.728225
1
73.994710
                12.9
                             -74.005043
                                                 40.740770
73.962565
                 5.3
                             -73.976124
                                                 40.790844
73.965316
                             -73.925023
                                                 40.744085
                16.0
73.973082
. . .
. . .
```

```
199995
                 3.0
                              -73.987042
                                                  40.739367
73.986525
199996
                 7.5
                              -73.984722
                                                  40.736837
74.006672
199997
                30.9
                              -73.986017
                                                  40.756487
73.858957
                14.5
                              -73.997124
199998
                                                  40.725452
73.983215
199999
                14.1
                              -73.984395
                                                  40.720077
73.985508
         dropoff latitude passenger count hour day month year
dayofweek
                40.723217
0
                                                  19
                                                        7
                                                                5
                                                                   2015
3
1
                40.750325
                                                  20
                                                       17
                                                                7
                                                                   2009
4
2
                40.772647
                                            1
                                                 21
                                                       24
                                                                8
                                                                   2009
0
3
                40.803349
                                            3
                                                   8
                                                       26
                                                                6
                                                                   2009
4
4
                40.761247
                                                  17
                                                       28
                                                                   2014
3
. . .
199995
                40.740297
                                                 10
                                                       28
                                                               10
                                                                   2012
199996
                40.739620
                                                   1
                                                       14
                                                                3
                                                                   2014
                                            2
                                                   0
                                                       29
199997
                40.692588
                                                                6
                                                                   2009
                40.695415
199998
                                                  14
                                                       20
                                                                5
                                                                   2015
199999
                40.768793
                                              4
                                                       15
                                                                5
                                                                   2010
[200000 rows x 11 columns]
from math import *
def distance formula(longitude1, latitude1, longitude2, latitude2):
    travel dist = []
    for pos in range (len(longitudel)):
         lon1, lan1, lon2, lan2 = map(radians, [longitude1[pos],
latitude1[pos], longitude2[pos], latitude2[pos]])
         dist lon = lon2 - lon1
         dist lan = lan2 - lan1
         a = \sin(\operatorname{dist \, lan/2})^{**2} + \cos(\operatorname{lan1}) * \cos(\operatorname{lan2}) *
```

```
sin(dist_lon/2)**2

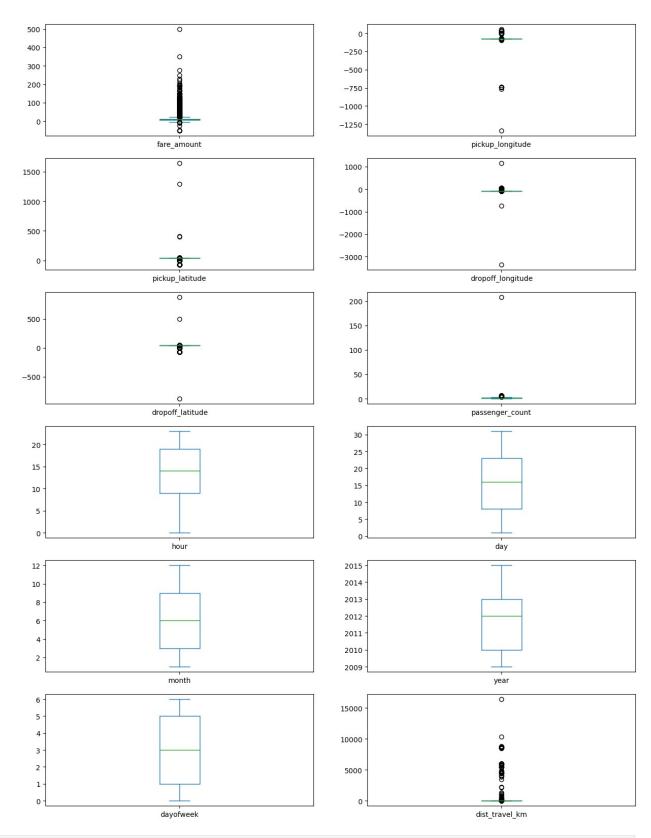
    #radius of earth = 6371
    c = 2 * asin(sqrt(a)) * 6371
    travel_dist.append(c)

return travel_dist

df['dist_travel_km'] =
    distance_formula(df.pickup_longitude.to_numpy(),
    df.pickup_latitude.to_numpy(),    df.dropoff_longitude.to_numpy(),
    df.dropoff_latitude.to_numpy())
```

2. Identify outliers.

```
df.plot(kind = "box", subplots = True, layout = (6,2), figsize=(15,20))
#Boxplot to check the outliers
plt.show()
```



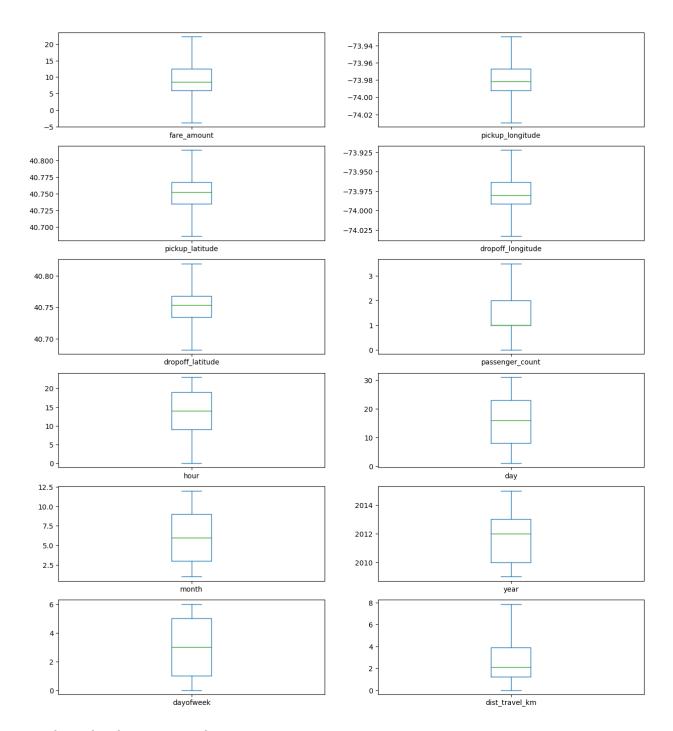
#Using the InterQuartile Range to fill the values
def remove_outlier(df1 , col):

```
Q1 = df1[col].quantile(0.25)
Q3 = df1[col].quantile(0.75)
IQR = Q3 - Q1
lower_whisker = Q1-1.5*IQR
upper_whisker = Q3+1.5*IQR
df[col] = np.clip(df1[col] , lower_whisker , upper_whisker)
return df1

def treat_outliers_all(df1 , col_list):
    for c in col_list:
        df1 = remove_outlier(df , c)
    return df1

df = treat_outliers_all(df , df.iloc[: , 0::])

#Boxplot shows that dataset is free from outliers
df.plot(kind = "box", subplots = True, layout = (7,2), figsize=(15,20))
plt.show()
```



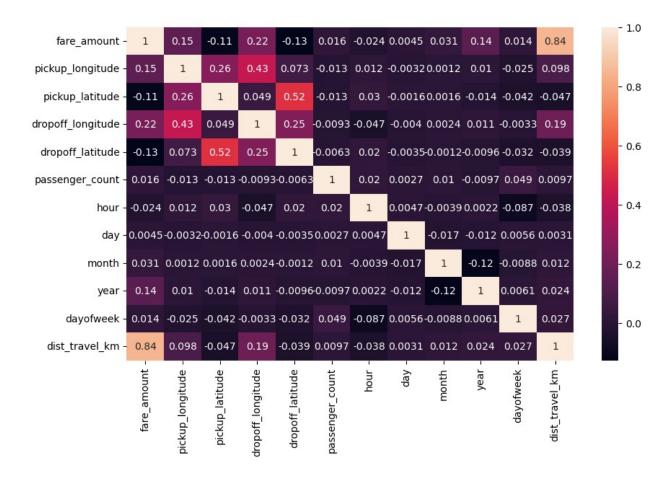
3. Check the correlation.

```
corr = df.corr()
corr
                                 pickup_longitude
                                                    pickup_latitude \
                    fare_amount
                                                           -0.110842
fare_amount
                       1.000000
                                          0.154069
pickup_longitude
                       0.154069
                                          1.000000
                                                            0.259497
pickup_latitude
                                          0.259497
                      -0.110842
                                                            1.000000
```

dropoff_longitude dropoff_latitude passenger_count hour day month year dayofweek dist_travel_km	0.218675 -0.125898 0.015778 -0.023623 0.004534 0.030817 0.141277 0.013652 0.844374	0.425619 0.073290 -0.013213 0.011579 -0.003204 0.001169 0.010198 -0.024652 0.098094	0.048889 0.515714 -0.012889 0.029681 -0.001553 0.001562 -0.014243 -0.042310 -0.046812
	dropoff longitude	dropoff latitude	
passenger_count \frac{1}{2} fare_amount 0.015778	· -	-0.125898	
pickup_longitude	0.425619	0.073290	-
0.013213 pickup_latitude 0.012889	0.048889	0.515714	-
dropoff_longitude	1.000000	0.245667	-
0.009303 dropoff_latitude 0.006308	0.245667	1.000000	-
passenger_count	-0.009303	-0.006308	
1.000000 hour 0.020274	-0.046558	0.019783	
day	-0.004007	-0.003479	
0.002712 month 0.010351	0.002391	-0.001193	
year	0.011346	-0.009603	-
0.009749 dayofweek	-0.003336	-0.031919	
0.048550			
dist_travel_km 0.009709	0.186531	-0.038900	
days Const.	hour day	month ye	ear
<pre>dayofweek \ fare_amount</pre>	-0.023623 0.004534	0.030817 0.1412	277 0.013652
pickup_longitude	0.011579 -0.003204	0.001169 0.0101	.98 -0.024652
pickup_latitude	0.029681 -0.001553	0.001562 -0.0142	243 -0.042310
dropoff_longitude	-0.046558 -0.004007	0.002391 0.0113	346 -0.003336
dropoff_latitude	0.019783 -0.003479	-0.001193 -0.0096	603 -0.031919
passenger_count	0.020274 0.002712	0.010351 -0.0097	49 0.048550

```
1.000000
                             0.004677 -0.003926 0.002156
hour
                                                             -0.086947
day
                   0.004677 1.000000 -0.017360 -0.012170
                                                              0.005617
                  -0.003926 -0.017360 1.000000 -0.115859
                                                             -0.008786
month
year
                   0.002156 -0.012170 -0.115859 1.000000
                                                              0.006113
dayofweek
                  -0.086947  0.005617  -0.008786  0.006113
                                                              1.000000
dist travel km
                  -0.038366 0.003062 0.011628 0.024278
                                                              0.027053
                   dist travel km
fare amount
                          0.844374
pickup_longitude
                          0.098094
pickup latitude
                         -0.046812
dropoff_longitude
dropoff_latitude
                         0.186531
                         -0.038900
passenger count
                          0.009709
                         -0.038366
hour
                          0.003062
day
                          0.011628
month
                          0.024278
year
dayofweek
                          0.027053
dist travel km
                         1.000000
fig,axis = plt.subplots(figsize = (10,6))
sns.heatmap(df.corr(),annot = True) #Correlation Heatmap (Light values
means highly correlated)
```

<Axes: >



4. Implement linear regression and random forest regression models.

```
df x =
df[['pickup longitude','pickup latitude','dropoff longitude','dropoff
latitude', 'passenger_count', 'hour', 'day', 'month', 'year', 'dayofweek', 'd
ist travel km']]
df \overline{y} = df[\overline{fare amount'}]
# Dividing the dataset into training and testing dataset
x_train, x_test, y_train, y_test = train_test_split(df_x, df_y,
test size=0.2, random state=1)
df
        fare amount
                      pickup longitude pickup latitude
dropoff longitude \
                7.50
                             -73.999817
                                                40.738354
73.999512
                7.70
                             -73.994355
                                                40.728225
73.994710
               12.90
                             -74.005043
                                                40.740770
73.962565
```

3 73.965316	5.30	-73.976124	40.	79084	4	-
4	16.00	-73.929786	40.	74408	5	_
73.973082	_0.30		.01		_	
199995	3.00	-73.987042	40	73936	7	
73.986525	3.00	-73.907042	40.	73930	/	-
199996	7.50	-73.984722	40.	73683	7	-
74.006672						
199997	22.25	-73.986017	40.	75648	7	-
73.922036	14 50	72 007124	40	72545	2	
199998 73.983215	14.50	-73.997124	40.	72545	2	-
199999	14.10	-73.984395	40.	72007	7	_
73.985508				_,_,,		
لــ	onoff letitude	naaaanaa	herre	devi	mar-t-	
dr dayofweek	<pre>copoff_latitude \</pre>	passenger_count	hour	day	month	year
0	40.723217	1.0	19	7	5	2015
3 1	.0.,23217	110		,		_010
	40.750325	1.0	20	17	7	2009
4	40 772647	1.0	2.1	2.4	0	2000
2 0	40.772647	1.0	21	24	8	2009
3	40.803349	3.0	8	26	6	2009
4		310	J		J	_500
4	40.761247	3.5	17	28	8	2014
3						
		• • • •				
199995	40.740297	1.0	10	28	10	2012
6		2.0	_3			
199996	40.739620	1.0	1	14	3	2014
4	40 002500	2.0	0	20	_	2000
199997 0	40.692588	2.0	0	29	6	2009
199998	40.695415	1.0	14	20	5	2015
2	101033113	110	± r	20	3	2013
199999	40.768793	1.0	4	15	5	2010
5						
di	st travel km					
0	1.683323					
	2.457590					
2	5.036377					
1 2 3 4	1.661683					
	4.475450					

```
199995
              0.112210
199996
              1.875050
199997
              7.865286
              3.539715
199998
199999
              5.417783
[200000 rows x 12 columns]
from sklearn.linear model import LinearRegression
reg = LinearRegression()
reg.fit(x train, y train)
LinearRegression()
y pred lin = reg.predict(x test)
print(y pred lin)
[ 6.27615184 5.09986098 9.43641238 ... 11.07663949 12.15392248
11.414960751
from sklearn.ensemble import RandomForestRegressor
rf = RandomForestRegressor(n estimators=100)
rf.fit(x_train,y_train)
RandomForestRegressor()
y_pred_rf = rf.predict(x_test)
print(y_pred_rf)
                                     11.268 13.325 ]
[ 4.929 6.491
                  9.3825 ... 11.39
```

5. Evaluate the models and compare their respective scores like R2, RMSE, etc

```
cols = ['Model', 'RMSE', 'R-Squared']
result_tabulation = pd.DataFrame(columns = cols)
from sklearn import metrics
from sklearn.metrics import r2_score
reg_RMSE = np.sqrt(metrics.mean_squared_error(y_test, y_pred_lin))
reg_squared = r2_score(y_test, y_pred_lin)
full_metrics = pd.Series({'Model': "Linear Regression", 'RMSE':
reg_RMSE, 'R-Squared': reg_squared})
result_tabulation = result_tabulation.append(full_metrics,
```

```
ignore index = True)
result tabulation
C:\Users\tanma\AppData\Local\Temp\ipykernel 22232\3572740849.py:9:
FutureWarning: The frame.append method is deprecated and will be
removed from pandas in a future version. Use pandas.concat instead.
  result tabulation = result tabulation.append(full metrics,
ignore index = True)
               Model
                          RMSE R-Squared
   Linear Regression 2.703957
                                 0.753906
rf_RMSE = np.sqrt(metrics.mean_squared_error(y_test, y_pred_rf))
rf squared = r2 score(y test, y pred rf)
full metrics = pd.Series({'Model': "Random Forest ", 'RMSE':rf RMSE,
'R-Squared': rf squared})
result tabulation = result tabulation.append(full metrics,
ignore index = True)
result tabulation
C:\Users\tanma\AppData\Local\Temp\ipykernel 22232\1194553861.py:6:
FutureWarning: The frame.append method is deprecated and will be
removed from pandas in a future version. Use pandas.concat instead.
  result tabulation = result tabulation.append(full metrics,
ignore index = True)
               Model
                          RMSE R-Squared
   Linear Regression 2.703957
                                 0.753906
      Random Forest
1
                      2.359689
                                 0.812582
```