```
import numpy as np
In [1]:
         import pandas as pd
In [2]: data = pd.read csv("C:/Users/Admin/Desktop/Aditya K/uber.csv")
In [3]:
         data.head
         <bound method NDFrame.head of</pre>
                                                 Unnamed: 0
                                                                                         key fare
Out[3]:
         _amount \
         0
                   24238194
                                2015-05-07 19:52:06.0000003
                                                                       7.5
         1
                   27835199
                                2009-07-17 20:04:56.0000002
                                                                       7.7
         2
                   44984355
                               2009-08-24 21:45:00.00000061
                                                                      12.9
         3
                   25894730
                                2009-06-26 08:22:21.0000001
                                                                      5.3
                   17610152 2014-08-28 17:47:00.000000188
         4
                                                                      16.0
                         . . .
                                                                       . . .
         199995
                   42598914
                               2012-10-28 10:49:00.00000053
                                                                       3.0
         199996
                   16382965
                                2014-03-14 01:09:00.0000008
                                                                      7.5
         199997
                   27804658
                              2009-06-29 00:42:00.00000078
                                                                    30.9
         199998
                   20259894
                                2015-05-20 14:56:25.0000004
                                                                      14.5
         199999
                               2010-05-15 04:08:00.00000076
                                                                      14.1
                   11951496
                         pickup_datetime pickup_longitude
                                                              pickup_latitude
         0
                 2015-05-07 19:52:06 UTC
                                                  -73.999817
                                                                    40.738354
         1
                 2009-07-17 20:04:56 UTC
                                                  -73.994355
                                                                    40.728225
         2
                 2009-08-24 21:45:00 UTC
                                                  -74.005043
                                                                    40.740770
         3
                 2009-06-26 08:22:21 UTC
                                                                    40.790844
                                                  -73.976124
         4
                 2014-08-28 17:47:00 UTC
                                                  -73.925023
                                                                    40.744085
                 2012-10-28 10:49:00 UTC
                                                  -73.987042
                                                                    40.739367
         199995
                 2014-03-14 01:09:00 UTC
                                                  -73.984722
                                                                    40.736837
         199996
                 2009-06-29 00:42:00 UTC
         199997
                                                  -73.986017
                                                                    40.756487
         199998
                 2015-05-20 14:56:25 UTC
                                                  -73.997124
                                                                    40.725452
         199999
                 2010-05-15 04:08:00 UTC
                                                  -73.984395
                                                                    40.720077
                 dropoff_longitude dropoff_latitude passenger_count
         0
                         -73.999512
                                            40.723217
                                                                       1
         1
                        -73.994710
                                            40.750325
                                                                      1
         2
                                            40.772647
                                                                      1
                        -73.962565
         3
                        -73.965316
                                            40.803349
                                                                      3
                        -73.973082
                                           40.761247
                                                                     5
         4
         . . .
                                                  . . .
         199995
                        -73.986525
                                            40.740297
                                                                      1
                                            40.739620
                        -74.006672
                                                                      1
         199996
         199997
                        -73.858957
                                            40.692588
                                                                      2
         199998
                        -73.983215
                                            40.695415
                                                                      1
         199999
                        -73.985508
                                            40.768793
                                                                      1
         [200000 rows x 9 columns]>
In [4]:
         data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
        RangeIndex: 200000 entries, 0 to 199999
        Data columns (total 9 columns):
             Column
                                Non-Null Count
                                                Dtype
         0
             Unnamed: 0
                                200000 non-null int64
         1
             key
                                200000 non-null object
         2
             fare amount
                                200000 non-null
                                                float64
         3
             pickup_datetime
                                200000 non-null object
         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
             passenger_count
                                200000 non-null int64
        dtypes: float64(5), int64(2), object(2)
        memory usage: 13.7+ MB
In [5]:
        data["pickup datetime"]= pd.to datetime(data["pickup datetime"])
In [6]: data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 200000 entries, 0 to 199999
        Data columns (total 9 columns):
             Column
                                Non-Null Count
                                                Dtype
             -----
                                -----
                                                ----
         0
             Unnamed: 0
                                200000 non-null int64
                                200000 non-null object
         1
             kev
         2
             fare_amount
                                200000 non-null float64
                                200000 non-null datetime64[ns, UTC]
         3
             pickup_datetime
                                200000 non-null float64
         4
             pickup_longitude
         5
             pickup latitude
                                200000 non-null float64
         6
             dropoff_longitude 199999 non-null float64
             dropoff_latitude
                                199999 non-null float64
         7
         8
                                200000 non-null int64
             passenger_count
        dtypes: datetime64[ns, UTC](1), float64(5), int64(2), object(1)
        memory usage: 13.7+ MB
In [7]: #for finding missing values
        data.isnull().sum()
                             0
        Unnamed: 0
Out[7]:
        kev
                             0
        fare amount
                             0
        pickup_datetime
                             0
        pickup longitude
                             0
        pickup_latitude
                             0
        dropoff_longitude
                             1
        dropoff latitude
                             1
        passenger_count
        dtype: int64
In [8]: # 0 means false & 1 means True
        #if True means null or missing values in dataset or in row
        #drop the row if it has missing values
        data.dropna(inplace = True)
In [9]:
        data.isnull().sum()
```

```
Unnamed: 0
                              0
Out[9]:
         key
                              0
         fare amount
                              0
         pickup_datetime
         pickup_longitude
                              0
         pickup latitude
                              0
         dropoff_longitude
                              0
         dropoff_latitude
                              0
         passenger_count
         dtype: int64
In [10]: # now create the machine Learning
In [11]: from sklearn.linear_model import LinearRegression
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import mean squared error
In [12]: # X IS PREDICTOR VARIABLE
         x =data.drop("fare amount",axis = 1)
         #y is target variable
         y = data["fare amount"]
In [13]: #to apply model
         x['pickup_datetime'] = pd.to_numeric(pd.to_datetime(x['pickup_datetime']))
         x = x.loc[:, x.columns.str.contains('^Unnamed')]
In [14]: x_train , x_test ,y_train ,y_test =train_test_split(x,y,test_size =0.2,)
         #testing dataset is 20%
         #traing dataset is 80% ,allocated to model
In [15]: # creating linear regression model
         lrmodel =LinearRegression()
         lrmodel.fit(x_train, y_train)
Out[15]: ▼ LinearRegression
         LinearRegression()
In [16]: #model is created
         pred = lrmodel.predict(x test)
In [17]: #calculating RMSEroot mean squared error
         lrmodelrmse = np.sqrt(mean squared error(pred ,y test))
         print("RMSE error is : ",lrmodelrmse)
         RMSE error is : 9.678125467724962
In [18]: #Random forest Regression
         from sklearn.ensemble import RandomForestRegressor
         # create RFR
         rfrmodel = RandomForestRegressor(n_estimators = 100 , random_state = 101)
In [26]: # fit the forest
```

```
rfrmodel.fit(x_train , y_train)
          rfrmodel pred = rfrmodel.predict(x_test)
In [25]:
         #calculate RMSE for RFR
         rfrmodel_rmse = np.sqrt(mean_squared_error(rfrmodel_pred ,y_test))
         print("RFR RMSE error is : ",rfrmodel_rmse)
         RFR RMSE error is : 11.83010330673784
In [24]:
         #pridiction
         pred = lrmodel.predict(x_test)
         print("hh",pred)
         lrmodel.predict(x_test)
         hh [11.37276195 11.35858736 11.35280229 ... 11.37269616 11.38204769
          11.36220594]
         array([11.37276195, 11.35858736, 11.35280229, ..., 11.37269616,
Out[24]:
                11.38204769, 11.36220594])
In [28]: from sklearn import metrics
         #R2 Score
         metrics.r2_score(y_test, rfrmodel_pred)
         -0.49417200840639075
Out[28]:
In [29]:
         from sklearn import metrics
         #R2 Score
         #R2 score Linear Regression
         metrics.r2_score(y_test , pred)
         -1.3856878633866643e-05
Out[29]:
In [30]:
         #R2 score Linear Regression is : 52%
         #R2 score RF Model is : 52%
         #Random Forest Model Best fit for Ths dataset, is perfect
In [ ]:
```