```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from scipy.stats import chi2_contingency
import seaborn as sns
from sklearn.neighbors import KNeighborsClassifier
from random import randrange, uniform
Train_Data = pd.read_csv(r'trainn.csv')
Train_Data.head(1)
₹
         Unnamed:
                               key fare_amount pickup_datetime pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude pass@
Train_Data.drop(labels='Unnamed: 0',axis=1,inplace=True)
Train_Data.drop(labels='key',axis=1,inplace=True)
Train Data.shape
→ (200000, 7)
test = pd.read_csv(r'testt.csv')
test.head(1)
₹
         Unnamed: Unnamed: Unnamed:
                                                     key pickup_datetime pickup_longitude pickup_latitude dropoff_longitude dropoff_latitu
                        0.1
                                0.1.1
test.shape, Train_Data.shape

→ ((50000, 10), (200000, 7))
Train_Data.head()
₹
         fare_amount
                            pickup_datetime pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude passenger_count
      0
                 7.5 2015-05-07 19:52:06 UTC
                                                     -73.999817
                                                                       40.738354
                                                                                          -73.999512
                                                                                                             40.723217
                 7.7 2009-07-17 20:04:56 UTC
                                                     -73.994355
                                                                       40.728225
                                                                                          -73.994710
                                                                                                             40.750325
                                                                                                                                      1
      2
                 12.9
                      2009-08-24 21:45:00 UTC
                                                     -74.005043
                                                                       40.740770
                                                                                          -73.962565
                                                                                                             40.772647
      3
                 5.3 2009-06-26 08:22:21 UTC
                                                     -73.976124
                                                                       40.790844
                                                                                          -73.965316
                                                                                                             40.803349
                                                                                                                                      3
                 16.0 2014-08-28 17:47:00 UTC
                                                     -73.925023
                                                                       40.744085
                                                                                          -73.973082
                                                                                                             40.761247
test.head()
\overline{\Sigma}
```

<u> </u>	Unnamed: 0	Unnamed: 0.1	Unnamed: 0.1.1	key	pickup_datetime	pickup_longitude	pickup_latitude	dropoff_longitude	dropoff_latitu
C	0	37338	31401407	2011-02-10 19:06:00.000000169	2011-02-10 19:06:00 UTC	-73.951662	40.790710	-73.947570	40.7562
1	1	160901	33158465	2011-06-23 09:24:00.000000157	2011-06-23 09:24:00 UTC	-73.951007	40.771508	-73.974075	40.7635
2	2 2	40428	10638355	2012-07-14 10:37:00.000000149	2012-07-14 10:37:00 UTC	-73.996473	40.747930	-73.990298	40.7561
3	3	63353	3836845	2014-10-19 22:27:05.0000002	2014-10-19 22:27:05 UTC	-73.997934	40.716890	-73.952617	40.7271
4									>

Y

As this is Taxi fare data and we know there are many factors which affect the price of taxi like

- 1. Travelled distance
- 2. Time of Travel
- 3. Demand and Availability of Taxi
- 4. Some special places are more costlier like Airport or other places where there might be toll

```
print(Train_Data.info())
print(test.info())
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200000 entries, 0 to 199999
    Data columns (total 7 columns):
                         200000 non-null float64
    pickup_datetime
pickup_longitude
                         200000 non-null object
                         200000 non-null float64
    pickup_latitude
                         200000 non-null float64
                         199999 non-null float64
    dropoff_longitude
    dropoff_latitude
                         199999 non-null float64
    passenger_count
                         200000 non-null int64
    dtypes: float64(5), int64(1), object(1)
    memory usage: 10.7+ MB
    None
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 50000 entries, 0 to 49999
    Data columns (total 10 columns):
    Unnamed: 0
                         50000 non-null int64
    Unnamed: 0.1
                         50000 non-null int64
    Unnamed: 0.1.1
                         50000 non-null int64
                         50000 non-null object
    key
    pickup_datetime
                         50000 non-null object
    pickup longitude
                         50000 non-null float64
    pickup_latitude
                         50000 non-null float64
    dropoff_longitude
                         50000 non-null float64
    dropoff_latitude
                         50000 non-null float64
    passenger_count
                         50000 non-null int64
    dtypes: float64(4), int64(4), object(2)
    memory usage: 3.8+ MB
    None
```

here we can see there are 8columns in which 6 numerics and 2 are object.

Lets change the type of pickup_datetime from object to DateTime

```
Train_Data["pickup_datetime"] = pd.to_datetime(Train_Data["pickup_datetime"])
print(Train_Data.info())
→ <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200000 entries, 0 to 199999
    Data columns (total 7 columns):
    fare_amount
                         200000 non-null float64
    pickup_datetime
                        200000 non-null datetime64[ns, UTC]
    pickup_longitude
                         200000 non-null float64
    pickup_latitude
                        200000 non-null float64
    dropoff_longitude 199999 non-null float64
    dropoff_latitude
                         199999 non-null float64
                        200000 non-null int64
    passenger count
    dtypes: datetime64[ns, UTC](1), float64(5), int64(1)
    memory usage: 10.7 MB
    None
Train_Data.describe()
```

$\overline{\Rightarrow}$		fare_amount	pickup_longitude	pickup_latitude	dropoff_longitude	dropoff_latitude	passenger_count
	count	200000.000000	200000.000000	200000.000000	199999.000000	199999.000000	200000.000000
	mean	11.359955	-72.527638	39.935885	-72.525292	39.923890	1.684535
	std	9.901776	11.437787	7.720539	13.117408	6.794829	1.385997
	min	-52.000000	-1340.648410	-74.015515	-3356.666300	-881.985513	0.000000
	25%	6.000000	-73.992065	40.734796	-73.991407	40.733823	1.000000
	50%	8.500000	-73.981823	40.752592	-73.980093	40.753042	1.000000
	75%	12.500000	-73.967154	40.767158	-73.963658	40.768001	2.000000
	max	499.000000	57.418457	1644.421482	1153.572603	872.697628	208.000000

- 1. Here first thing which we can see is minimum value of fare is negative which is -52 which is not the valid value, so we need to remove the fare which are negative values.
- 2.Secondly, passenger_count minimum value is 0 and maximum value is 208 which impossible, so we need to remove them as well, for safer side we can think that a taxi can have maximum 7 people.

Unsupported Cell Type. Double-Click to inspect/edit the content.

```
Train_Data.isnull().sum()
```

 fare_amount	0
pickup_datetime	0
pickup_longitude	0
pickup_latitude	0
dropoff_longitude	1
dropoff_latitude	1
passenger_count	0
dtype: int64	

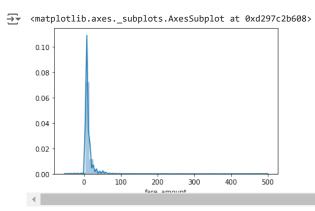
Train_Data.dropna(axis = 0, inplace= True)

print(Train_Data.isnull().sum())

fare_amount	0
pickup_datetime	0
pickup_longitude	0
pickup_latitude	0
dropoff_longitude	0
dropoff_latitude	0
passenger_count	0
dtype: int64	
	pickup_datetime pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude passenger_count

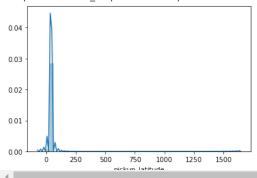
Lets see the statistics of our data

sns.distplot(Train_Data['fare_amount'])

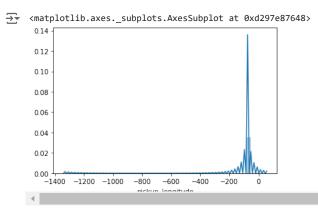


sns.distplot(Train_Data['pickup_latitude'])

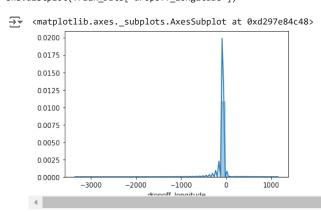
<matplotlib.axes._subplots.AxesSubplot at 0xd297d1b648>



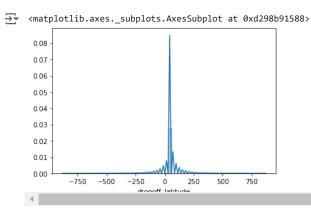
sns.distplot(Train_Data['pickup_longitude'])



sns.distplot(Train_Data['dropoff_longitude'])



sns.distplot(Train_Data['dropoff_latitude'])



print("drop_off latitude min value",Train_Data["dropoff_latitude"].min())
print("drop_off latitude max value",Train_Data["dropoff_latitude"].max())
print("drop_off longitude min value", Train_Data["dropoff_longitude"].min())
print("drop_off longitude max value",Train_Data["dropoff_longitude"].max())
print("pickup latitude min value",Train_Data["pickup_latitude"].min())

```
print("pickup latitude max value",Train_Data["pickup_latitude"].max())
print("pickup longitude min value", Train_Data["pickup_longitude"].min())
print("pickup longitude max value", Train_Data["pickup_longitude"].max())
→ drop_off latitude min value -881.9855130000002
     drop_off latitude max value 872.6976279999999
     drop_off longitude min value -3356.6663
     drop_off longitude max value 1153.5726029999996
     pickup latitude min value -74.01551500000002
     pickup latitude max value 1644.421482
     pickup longitude min value -1340.64841
     pickup longitude max value 57.418457
print("drop_off latitude min value",test["dropoff_latitude"].min())
print("drop_off latitude max value",test["dropoff_latitude"].max())
print("drop_off longitude min value", test["dropoff_longitude"].min())
print("drop_off longitude max value",test["dropoff_longitude"].max())
print("pickup latitude min value",test["pickup_latitude"].min())
print("pickup latitude max value",test["pickup_latitude"].max())
print("pickup longitude min value",test["pickup_longitude"].min())
print("pickup longitude max value",test["pickup_longitude"].max())
→ drop_off latitude min value -74.00110699999998
     drop_off latitude max value 47.433332
     drop_off longitude min value -1491.194073
     drop_off longitude max value 40.796262
     pickup latitude min value -74.001047
     pickup latitude max value 42.46842
     pickup longitude min value -80.734728
     pickup longitude max value 40.812005
min_longitude=-1491.194073,
min_latitude=-74.001047,
max_longitude=40.812005,
max_latitude=41.709555
min_longitude=-1491.194073,
min_latitude=-74.001047,
max longitude=40.812005,
max_latitude=41.709555
tempdf=Train_Data[(Train_Data["dropoff_latitude"]<min_latitude) |</pre>
                (Train_Data["pickup_latitude"]<min_latitude) |
                (Train_Data["dropoff_longitude"]<min_longitude) |
                (Train_Data["pickup_longitude"]<min_longitude) |</pre>
                (Train_Data["dropoff_latitude"]>max_latitude) |
                (Train_Data["pickup_latitude"]>max_latitude) |
                (Train_Data["dropoff_longitude"]>max_longitude) |
                (Train_Data["pickup_longitude"]>max_longitude) ]
print("before droping",Train_Data.shape)
Train_Data.drop(tempdf.index,inplace=True)
print("after droping", Train_Data.shape)
    before droping (199999, 7)
     after droping (199961, 7)
import calendar
Train_Data['day']=Train_Data['pickup_datetime'].apply(lambda x:x.day)
Train_Data['hour']=Train_Data['pickup_datetime'].apply(lambda x:x.hour)
Train_Data['month']=Train_Data['pickup_datetime'].apply(lambda x:x.month)
Train_Data['year']=Train_Data['pickup_datetime'].apply(lambda x:x.year)
Train_Data['weekday']=Train_Data['pickup_datetime'].apply(lambda x: calendar.day_name[x.weekday()])
Train_Data.weekday = Train_Data.weekday.map({'Sunday':0,'Monday':1,'Tuesday':2,'Wednesday':3,'Thursday':4,'Friday':5,'Saturday':6})
Train_Data.drop(labels = 'pickup_datetime',axis=1,inplace=True)
Train_Data.head(1)
Train_Data.info()
<class 'pandas.core.frame.DataFrame'>
     Int64Index: 199961 entries, 0 to 199999
     Data columns (total 11 columns):
     fare_amount
                          199961 non-null float64
```

```
pickup_longitude
                     199961 non-null float64
pickup_latitude
                     199961 non-null float64
{\tt dropoff\_longitude}
                     199961 non-null float64
                     199961 non-null float64
dropoff_latitude
passenger_count
                     199961 non-null int64
                     199961 non-null int64
hour
                     199961 non-null int64
                     199961 non-null int64
month
                     199961 non-null int64
year
weekday
                     199961 non-null int64
```

dtypes: float64(5), int64(6)

memory usage: 18.3 MB

Model Building

from sklearn.model_selection import train_test_split

```
x=Train_Data.drop("fare_amount", axis=1)
```

	pickup_longitude	pickup_latitude	${\tt dropoff_longitude}$	dropoff_latitude	passenger_d	count	day	hour	month	year	weekday
0	-73.999817	40.738354	-73.999512	40.723217		1	7	19	5	2015	4
1	-73.994355	40.728225	-73.994710	40.750325		1	17	20	7	2009	5
2	-74.005043	40.740770	-73.962565	40.772647		1	24	21	8	2009	1
3	-73.976124	40.790844	-73.965316	40.803349		3	26	8	6	2009	5
4	-73.925023	40.744085	-73.973082	40.761247		5	28	17	8	2014	4
199995	-73.987042	40.739367	-73.986525	40.740297		1	28	10	10	2012	0
199996	-73.984722	40.736837	-74.006672	40.739620		1	14	1	3	2014	5
199997	-73.986017	40.756487	-73.858957	40.692588		2	29	0	6	2009	1
199998	-73.997124	40.725452	-73.983215	40.695415		1	20	14	5	2015	3
199999	-73.984395	40.720077	-73.985508	40.768793		1	15	4	5	2010	6
199961 ro	ws × 10 columns										
4											

y=Train_Data["fare_amount"]

x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=101)

x_train.head()

_		pickup_longitude	pickup_latitude	dropoff_longitude	dropoff_latitude	passenger_count	day	hour	month	year	weekday
	36449	-73.963597	40.761797	-73.970199	40.762533	1	4	22	9	2014	4
1	77679	-74.013143	40.705700	-73.867012	40.768862	4	2	17	1	2013	3
	36877	-73.993683	40.702455	-73.917713	40.684747	2	31	3	10	2010	0
	20428	-73.954686	40.780613	-73.971005	40.758253	1	29	10	8	2012	3
4	18927	-73.978887	40.777162	-73.993860	40.746392	1	14	16	7	2013	0

x_test.head()

→ *		pickup_longitude	pickup_latitude	dropoff_longitude	dropoff_latitude	passenger	_count	day	hour	month	year	weekday
	35707	-73.979422	40.743943	-73.969162	40.758608		1	26	16	6	2012	2
	37333	-74.000873	40.747298	-73.991410	40.764548		2	14	17	4	2011	4
	131999	-74.007640	40.732222	-73.988398	40.748832		5	29	8	11	2010	1
	106818	-73.960133	40.719825	-73.942702	40.717567		4	6	21	9	2013	5
	52881	-73.995711	40.764551	-73.991177	40.750312		2	1	8	5	2009	5

```
y_train.head()
→ 36449
                4.5
     177679
               47.3
     36877
               20.9
     20428
                8.9
     18927
               11.0
     Name: fare_amount, dtype: float64
y_test.head()
<del>→</del> 35707
               7.3
     37333
               6.5
     131999
               8.1
     106818
     52881
               9.7
     Name: fare_amount, dtype: float64
print(x_train.shape)
print(x_test.shape)
print(y_test.shape)
print(y_train.shape)

→ (159968, 10)
     (39993, 10)
     (39993,)
     (159968,)
```

10/10/24, 7:16 PM

Linear Regression

```
from sklearn.linear_model import LinearRegression

lrmodel=LinearRegression()
lrmodel.fit(x_train, y_train)

    LinearRegression()

predictedvalues = lrmodel.predict(x_test)

from sklearn.metrics import mean_squared_error

lrmodelrmse = np.sqrt(mean_squared_error(predictedvalues, y_test))
print("RMSE value for Linear regression is", lrmodelrmse)

    RMSE value for Linear regression is 10.05963609768883
```

Random Forest

```
from sklearn.ensemble import RandomForestRegressor
rfrmodel = RandomForestRegressor(n_estimators=100, random_state=101)

rfrmodel.fit(x_train,y_train)
rfrmodel_pred= rfrmodel.predict(x_test)

rfrmodel_rmse=np.sqrt(mean_squared_error(rfrmodel_pred, y_test))
print("RMSE value for Random forest regression is ",rfrmodel_rmse)

→ RMSE value for Random forest regression is 4.838358623297524

rfrmodel_pred.shape

→ (39993,)
```

Working on Test Data

```
test = pd.read_csv(r'testt.csv')
test.drop(test[['Unnamed: 0','Unnamed: 0.1','Unnamed: 0.1.1','key']],axis=1,inplace=True)
test.isnull().sum()
→ pickup_datetime
                          a
     pickup_longitude
     pickup_latitude
     {\tt dropoff\_longitude}
                          0
     dropoff_latitude
                          0
     passenger_count
     dtype: int64
test["pickup_datetime"] = pd.to_datetime(test["pickup_datetime"])
test['day']=test['pickup_datetime'].apply(lambda x:x.day)
test['hour']=test['pickup_datetime'].apply(lambda x:x.hour)
test['month']=test['pickup_datetime'].apply(lambda x:x.month)
test['year']=test['pickup_datetime'].apply(lambda x:x.year)
test['weekday']=test['pickup_datetime'].apply(lambda x: calendar.day_name[x.weekday()])
test.head(5)
\overline{2}
         pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude passenger_count day hour
                                                                                                              month year weekday
      0
               -73.951662
                                  40.790710
                                                    -73.947570
                                                                        40.756220
                                                                                                    10
                                                                                                           19
                                                                                                                   2 2011
      1
                -73.951007
                                  40.771508
                                                     -73.974075
                                                                        40.763553
                                                                                                                     2011
                                                                                                 1
                                                                                                    23
                                                                                                            9
                                                                                                                   6
      2
                -73.996473
                                  40.747930
                                                    -73.990298
                                                                        40.756152
                                                                                                 6
                                                                                                           10
                                                                                                                   7 2012
                                                                                                                                  6
                                                                                                     14
                -73.997934
                                  40.716890
                                                    -73.952617
                                                                        40.727149
                                                                                                                                  0
                                                                                                     19
                                                                                                           22
                                                                                                                  10 2014
                -73.952583
                                  40.714039
                                                     -73.906128
                                                                        40.711281
                                                                                                           22
                                                                                                                   5 2015
                                                                                                    25
test.drop(['pickup_datetime'], axis = 1, inplace = True)
test.weekday = test.weekday.map({'Sunday':0,'Monday':1,'Tuesday':2,'Wednesday':3,'Thursday':4,'Friday':5,'Saturday':6})
rfrmodel_pred= rfrmodel.predict(test)
df = pd.DataFrame(rfrmodel_pred)
df
₹
          0 11.2380
              9.1880
          2
              5.9320
          3 11.8570
          4 11.9080
      49995
             6.7250
      49996 27.4532
      49997
             7.4750
      49998
            7.7190
      49999 10.8940
     50000 rows × 1 columns
```

Start coding or generate with AI.

Start coding or $\underline{\text{generate}}$ with AI.