

Model1LR

June 12, 2022

```
[ ]: #import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
[ ]: #read data
Data = pd.read_csv('C:/Users/Ahmed Atef/Desktop/data.csv')
print('Data = \n',Data.head(10))
```

Data =

	X1	Y1	X2	Y2	X3	Y3	X4	\
0	0.860019	-3.096488	0.753413	-2.530522	0.629331	-1.962419	0.361692	
1	0.456454	-2.852605	0.565661	-2.188466	0.568863	-1.593943	0.363601	
2	-0.462493	-2.031486	-0.125549	-1.528304	0.123850	-1.026633	0.238231	
3	0.106369	-2.186939	0.116188	-1.520864	0.142727	-0.703652	0.068778	
4	0.381094	-3.316631	0.390072	-2.754067	0.427469	-2.131407	0.336888	
5	-0.262968	-3.256483	-0.094474	-2.594541	0.073604	-1.954448	0.088474	
6	-0.970940	-2.595038	-0.683583	-2.042165	-0.376025	-1.472786	-0.190104	
7	-0.304093	-1.997585	-0.215963	-1.307632	-0.116897	-0.454085	-0.069022	
8	0.055419	-3.592592	0.131205	-2.973669	0.252867	-2.286043	0.242098	
9	-0.344932	-3.766955	-0.137432	-2.997226	0.053303	-2.269871	0.066792	

	Y4	Angle	X5	Y5	X6	Y6	X7	\
0	-1.227911	15.958005	0.456454	-2.852605	0.565661	-2.188466	0.568863	
1	-0.907820	16.424084	-0.462493	-2.031486	-0.125549	-1.528304	0.123850	
2	-0.510583	15.635178	-0.915264	-0.867298	-0.556912	-0.619065	-0.347119	
3	0.170585	26.299685	0.381094	-3.316631	0.390072	-2.754067	0.427469	
4	-1.266161	15.050765	-0.262968	-3.256483	-0.094474	-2.594541	0.073604	
5	-1.111491	13.807231	-0.970940	-2.595038	-0.683583	-2.042165	-0.376025	
6	-0.800848	14.902200	-1.493574	-1.457450	-1.151122	-1.159289	-0.929195	
7	0.416620	15.051214	0.055419	-3.592592	0.131205	-2.973669	0.252867	
8	-1.329521	12.915006	-0.344932	-3.766955	-0.137432	-2.997226	0.053303	
9	-1.308215	7.824200	-0.905729	-3.185208	-0.604510	-2.520129	-0.291183	

	Y7	X8	Y8	Angle number	RightorLeft
0	-1.593943	0.363601	-0.907820	2	0
1	-1.026633	0.238231	-0.510583	3	0
2	-0.385286	-0.006106	-0.007940	4	0

3	-2.131407	0.336888	-1.266161	1	0
4	-1.954448	0.088474	-1.111491	2	0
5	-1.472786	-0.190104	-0.800848	3	0
6	-0.890322	-0.549895	-0.352399	4	0
7	-2.286043	0.242098	-1.329521	1	0
8	-2.269871	0.066792	-1.308215	2	0
9	-1.861470	-0.161199	-1.063399	3	0

```
[ ]: #show data details
print('Data.describe = \n',Data.describe())
```

```
Data.describe =
```

	X1	Y1	X2	Y2	X3 \
count	4382.000000	4382.000000	4382.000000	4382.000000	4382.000000
mean	-0.252894	-2.350716	-0.187523	-1.889514	-0.088361
std	0.948101	0.848965	0.822127	0.751828	0.653238
min	-2.702862	-4.136758	-2.085325	-3.225857	-1.476798
25%	-0.831879	-2.977523	-0.673609	-2.433257	-0.496952
50%	-0.146311	-2.435614	-0.075717	-1.984988	0.004254
75%	0.420229	-1.936213	0.418198	-1.596749	0.424851
max	2.221333	0.793691	1.774284	0.824333	1.342360

	Y3	X4	Y4	Angle	X5 \
count	4382.000000	4382.000000	4382.000000	4382.000000	4382.000000
mean	-1.375213	-0.047605	-0.671134	16.798604	0.055876
std	0.696510	0.392837	0.572180	9.504037	0.965893
min	-2.409056	-0.941107	-1.434023	2.763818	-2.157534
25%	-1.882652	-0.400934	-1.053675	10.121567	-0.711105
50%	-1.562080	0.000243	-0.886340	14.579348	0.023831
75%	-1.056059	0.291479	-0.433275	20.187453	0.762173
max	0.771254	0.805917	0.790989	61.767698	2.469773

	Y5	X6	Y6	X7	Y7 \
count	4382.000000	4382.000000	4382.000000	4382.000000	4382.000000
mean	-2.276064	0.114580	-1.893112	0.157911	-1.477540
std	0.825647	0.804275	0.661706	0.665437	0.508988
min	-4.128808	-1.736918	-3.225857	-1.427917	-2.409056
25%	-2.947262	-0.536102	-2.412434	-0.373223	-1.866497
50%	-2.297369	0.093548	-1.953126	0.161692	-1.548207
75%	-1.663655	0.698498	-1.451524	0.622315	-1.156084
max	0.532019	2.074761	0.591465	1.746115	0.604385

	X8	Y8	Angle number	RightorLeft
count	4382.000000	4382.000000	4382.000000	4382.000000
mean	0.101443	-0.804559	2.518028	0.500228
std	0.440341	0.348770	1.106318	0.500057
min	-0.941107	-1.434023	1.000000	0.000000
25%	-0.228869	-1.044162	2.000000	0.000000

50%	0.107943	-0.879848	3.000000	1.000000
75%	0.408215	-0.577973	3.750000	1.000000
max	1.148080	0.564301	4.000000	1.000000

```
[ ]: # Quick look at the data structure
Data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4382 entries, 0 to 4381
Data columns (total 19 columns):
#   Column          Non-Null Count  Dtype
---  -
0   X1               4382 non-null   float64
1   Y1               4382 non-null   float64
2   X2               4382 non-null   float64
3   Y2               4382 non-null   float64
4   X3               4382 non-null   float64
5   Y3               4382 non-null   float64
6   X4               4382 non-null   float64
7   Y4               4382 non-null   float64
8   Angle            4382 non-null   float64
9   X5               4382 non-null   float64
10  Y5               4382 non-null   float64
11  X6               4382 non-null   float64
12  Y6               4382 non-null   float64
13  X7               4382 non-null   float64
14  Y7               4382 non-null   float64
15  X8               4382 non-null   float64
16  Y8               4382 non-null   float64
17  Angle number     4382 non-null   int64
18  RightorLeft      4382 non-null   int64
dtypes: float64(17), int64(2)
memory usage: 650.6 KB
```

```
[ ]: Data['RightorLeft'].value_counts()
```

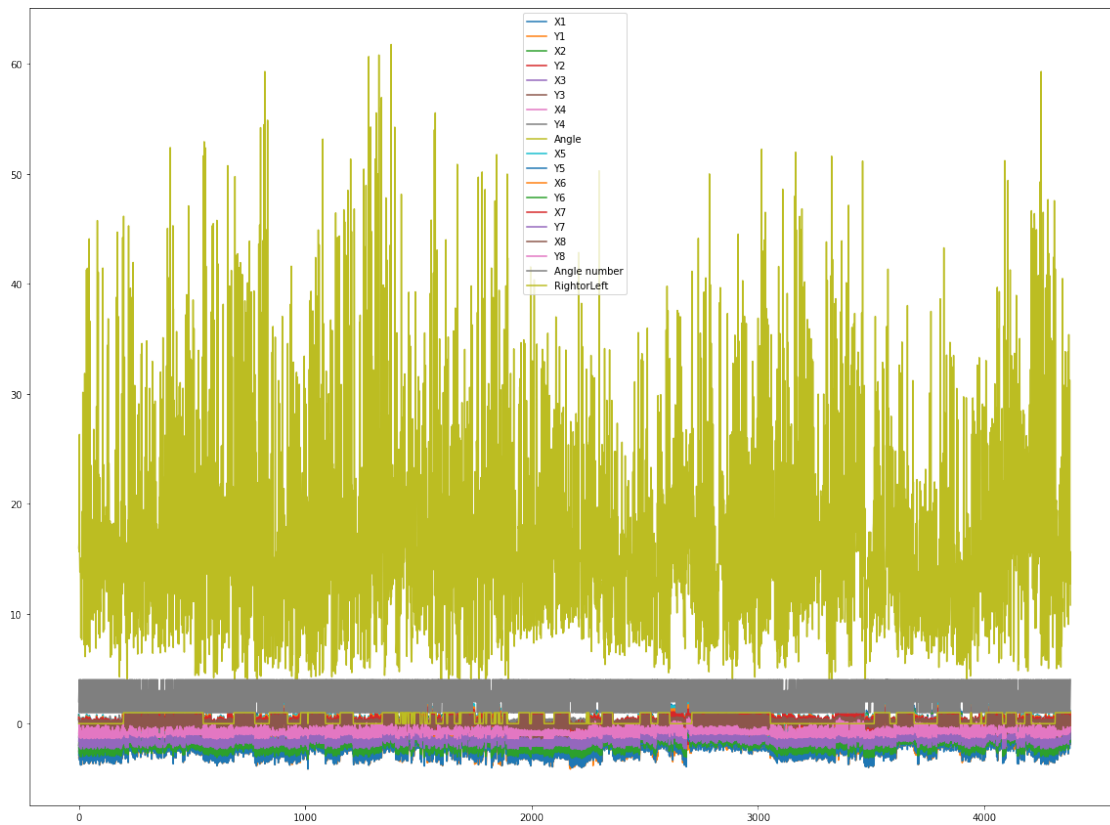
```
[ ]: 1    2192
     0    2190
     Name: RightorLeft, dtype: int64
```

```
[ ]: Data['Angle number'].value_counts()
```

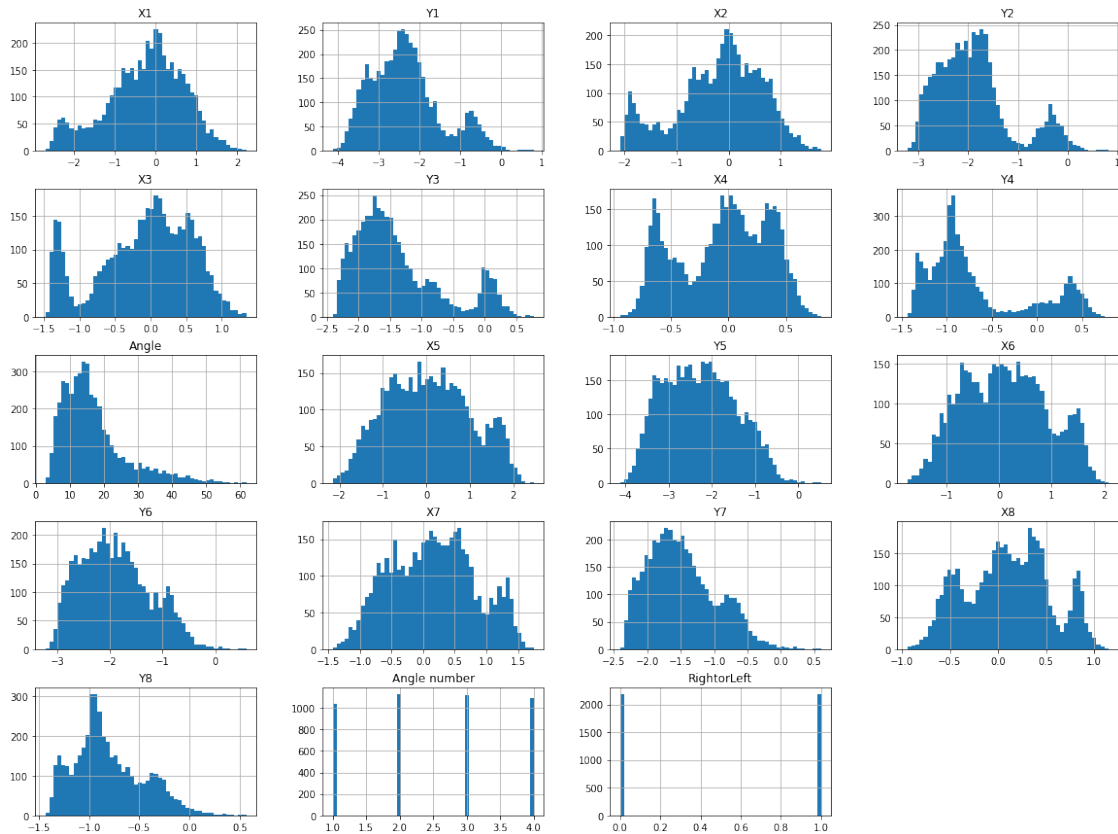
```
[ ]: 2    1132
     3    1116
     4    1096
     1    1038
     Name: Angle number, dtype: int64
```

```
[ ]: Data.plot(figsize=(20,15))
```

```
[ ]: <AxesSubplot:>
```



```
[ ]: %matplotlib inline
import matplotlib.pyplot as plt
Data.hist(bins=50, figsize=(20,15))
plt.show()
```



```
[ ]:
```

```
[ ]: # split the data
from sklearn.model_selection import train_test_split

train_set, test_set = train_test_split(Data, test_size=0.2, random_state=42)
```

```
[ ]: test_set.head()
```

```
[ ]:
```

	X1	Y1	X2	Y2	X3	Y3	X4	\
670	-0.036958	-2.439708	-0.151748	-1.751869	-0.203871	-0.937441	-0.286370	
2417	-0.496443	-2.212595	-0.277505	-1.715926	-0.049095	-1.199572	0.013035	
596	-0.284349	-3.169149	-0.103853	-2.481122	0.073029	-1.829867	0.053820	
2629	1.558989	-2.525897	1.278059	-2.063646	0.987350	-1.607824	0.559566	
1395	0.921632	-1.839371	0.870753	-1.643973	0.769634	-1.350292	0.511160	

	Y4	Angle	X5	Y5	X6	Y6	X7	\
670	-0.047744	39.606503	1.560510	-3.120231	1.263452	-2.564677	0.977038	
2417	-0.613426	15.629651	-1.010371	-1.038258	-0.697754	-0.783644	-0.524011	
596	-1.061236	4.180429	-0.885141	-2.500813	-0.577390	-1.839822	-0.304928	
2629	-1.053691	25.326243	1.193229	-2.051963	1.046683	-1.562685	0.855178	

```
1395 -0.733900 13.204648 1.370833 -0.738958 1.302034 -0.746930 1.207433
```

	Y7	X8	Y8	Angle number	RightorLeft
670	-1.978325	0.514782	-1.211725	1	0
2417	-0.556461	-0.271823	-0.118667	4	0
596	-1.263751	-0.188361	-0.663147	3	0
2629	-1.099170	0.454378	-0.651524	2	0
1395	-0.688365	0.869758	-0.309678	4	1

```
[ ]: Data = train_set.copy()
```

```
[ ]: len(test_set)
```

```
[ ]: 877
```

```
[ ]: len(train_set)
```

```
[ ]: 3505
```

```
[ ]: corr_matrix = Data.corr()
```

```
[ ]: corr_matrix["Angle"].sort_values(ascending=False)
```

```
[ ]: Angle          1.000000
     Y4             0.673085
     Y3             0.633347
     Y2             0.551183
     Y1             0.448352
     RightorLeft    -0.007250
     X5             -0.046684
     X6             -0.083856
     Y5             -0.103842
     X7             -0.117606
     Y6             -0.129496
     Y7             -0.149414
     X8             -0.171393
     Y8             -0.207860
     X4             -0.324361
     X1             -0.355447
     X3             -0.373661
     X2             -0.378034
     Angle number   -0.491408
     Name: Angle, dtype: float64
```

```
[ ]: # Prepare the data for Machine Learning algorithms
```

```
Data = train_set.drop("Angle", axis=1) # drop labels for training set
Data_label = train_set["Angle"].copy()
```

```
[ ]: Data_label
```

```
[ ]: 647      8.964206
      2194    18.802308
      2244    18.284985
      4256    15.296240
      1164    16.180668
      ...
      3444    15.691533
      466     6.504340
      3092    41.578601
      3772    14.907241
      860     5.382245
      Name: Angle, Length: 3505, dtype: float64
```

```
[ ]: # Select and train a model

      from sklearn.linear_model import LinearRegression

      lin_reg = LinearRegression()
      lin_reg.fit(Data, Data_label)
```

```
[ ]: LinearRegression()
```

```
[ ]: some_data = Data.iloc[:5]
      some_labels = Data_label.iloc[:5]
      print("Predictions:", lin_reg.predict(some_data))
```

```
Predictions: [ 9.51326826 11.45098353 11.91521907 14.17160773 13.56828679]
```

```
[ ]: # Compare against the actual values:

      print("Labels:", list(some_labels))
```

```
Labels: [8.964205768, 18.80230809, 18.28498482, 15.29624034, 16.1806679]
```

```
[ ]: from sklearn.metrics import mean_squared_error

      Data_predictions = lin_reg.predict(Data)
      lin_mse = mean_squared_error(Data_label, Data_predictions)
      lin_rmse = np.sqrt(lin_mse)
      lin_rmse
```

```
[ ]: 5.675163465130474
```

```
[ ]: # Cross Validation

      from sklearn.model_selection import cross_val_score
```

```
lin_scores = cross_val_score(lin_reg, Data, Data_label,
                             scoring="neg_mean_squared_error", cv=10)
lin_rmse_scores = np.sqrt(-lin_scores)
```

```
[ ]: def display_scores(scores):
      print("Scores:", scores)
      print("Mean:", scores.mean())
      print("Standard deviation:", scores.std())

      display_scores(lin_rmse_scores)
```

```
Scores: [5.5514259  5.70120708 5.636896   5.81338988 5.43271082 5.9824288
         5.82029774 5.59668624 5.9919363  5.58201444]
Mean: 5.710899320110563
Standard deviation: 0.17698286810221756
```

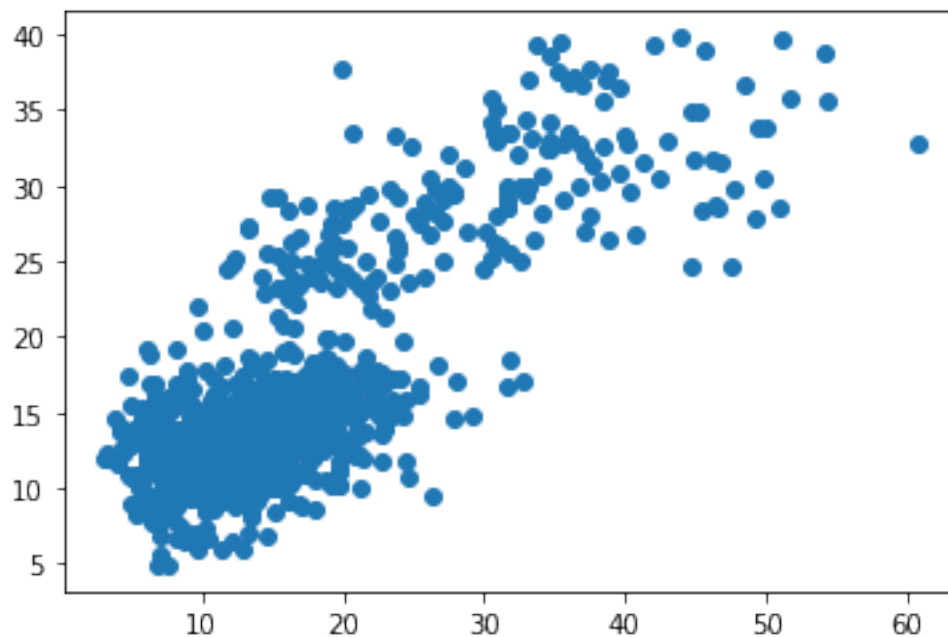
```
[ ]: X_test = test_set.drop("Angle", axis=1)
      y_test = test_set["Angle"].copy()

      final_predictions = lin_reg.predict(X_test)

      final_mse = mean_squared_error(y_test, final_predictions)
      final_rmse = np.sqrt(final_mse)
```

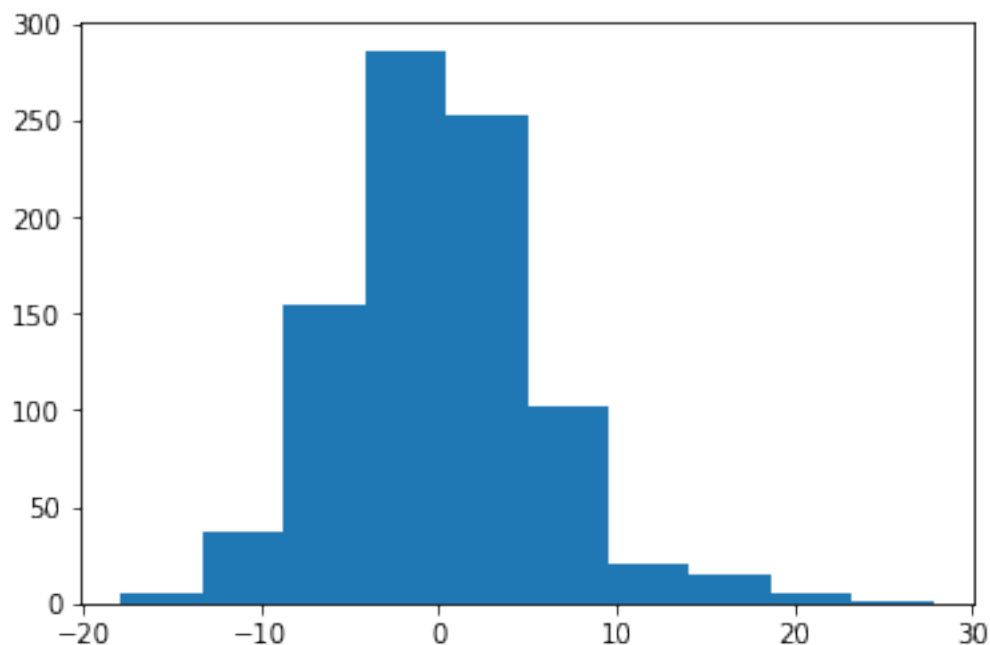
```
[ ]: plt.scatter(y_test, final_predictions)
```

```
[ ]: <matplotlib.collections.PathCollection at 0x1ce4413ba30>
```




```
[ ]: plt.hist(y_test - final_predictions)
```

```
[ ]: (array([ 5., 37., 154., 286., 252., 102., 20., 15., 5., 1.]),  
      array([-17.84574147, -13.27659286, -8.70744425, -4.13829565,  
            0.43085296,  5.00000157,  9.56915017, 14.13829878,  
            18.70744739, 23.27659599, 27.8457446 ]),  
      <BarContainer object of 10 artists>)
```



```
[ ]: final_rmse
```

```
[ ]: 5.806596073732241
```

```
[ ]: from scipy import stats  
  
confidence = 0.95  
squared_errors = (final_predictions - y_test) ** 2  
np.sqrt(stats.t.interval(confidence, len(squared_errors) - 1,  
                          loc=squared_errors.mean(),  
                          scale=stats.sem(squared_errors)))
```

```
[ ]: array([5.43405513, 6.15663551])
```

```
[ ]: import sklearn.metrics as sm
```

```

print("Regressor model performance:")
print("Mean absolute error(MAE) =", round(sm.mean_absolute_error(y_test,
    ↪final_predictions), 2))
print("Mean squared error(MSE) =", round(sm.mean_squared_error(y_test,
    ↪final_predictions), 2))
mean_test=round(sm.mean_squared_error(y_test, final_predictions), 2)
print(np.sqrt(mean_test))
print("Median absolute error =", round(sm.median_absolute_error(y_test,
    ↪final_predictions), 2))
print("Explain variance score =", round(sm.explained_variance_score(y_test,
    ↪final_predictions), 2))
print("R2 score =", round(sm.r2_score(y_test, final_predictions), 2))

```

Regressor model performance:
 Mean absolute error(MAE) = 4.4
 Mean squared error(MSE) = 33.72
 5.80689245638319
 Median absolute error = 3.49
 Explain variance score = 0.62
 R2 score = 0.62

```

[ ]: def mean_absolute_percentage_error(y_test, final_predictions):
    y_test, final_predictions = np.array(y_test), np.array(final_predictions)
    return np.mean(np.abs((y_test - final_predictions) / final_predictions)) *
    ↪100

```

```

[ ]: print(mean_absolute_percentage_error(y_test, final_predictions))

```

```

[ ]:

```