

servo Prediction using Linear Regression

In control engineering a servomechanism, usually shortened to servo, is an automatic device that uses error- sensing negative feedback to correct the action of a mechanism.

from previous projects data

```
In [1]: import pandas as pd
import numpy as np
```

```
In [131]: df = pd.read_csv("/home/student/Downloads/finance.csv")
```

```
In [132]: df.head()
```

Out[132]:

	motor	screw	Pgain	Vgain	class
0	1	2	5	4	4
1	3	5	6	5	11
2	2	3	4	3	6
3	2	4	3	2	48
4	5	3	6	5	6

```
In [133]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5 entries, 0 to 4
Data columns (total 5 columns):
#   Column   Non-Null Count  Dtype
---  -
0    motor    5 non-null      int64
1    screw    5 non-null      int64
2    Pgain    5 non-null      int64
3    Vgain    5 non-null      int64
4    class    5 non-null      int64
dtypes: int64(5)
memory usage: 328.0 bytes
```

```
In [134]: df.describe()
```

Out[134]:

	motor	screw	Pgain	Vgain	class
count	5.000000	5.000000	5.000000	5.000000	5.000000
mean	2.600000	3.400000	4.800000	3.800000	15.000000
std	1.516575	1.140175	1.30384	1.30384	18.627936
min	1.000000	2.000000	3.000000	2.000000	4.000000
25%	2.000000	3.000000	4.000000	3.000000	6.000000
50%	2.000000	3.000000	5.000000	4.000000	6.000000
75%	3.000000	4.000000	6.000000	5.000000	11.000000

motor	screw	Pgain	Vgain	class
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```
In [135]: df[['motor']] value_counts()
```

```
Out[135]: motor
2         2
1         1
3         1
5         1
dtype: int64
```

```
In [136]: df[['screw']] value_counts()
```

```
Out[136]: screw
3         2
2         1
4         1
5         1
dtype: int64
```

```
In [137]: v = df['class']
```

```
In [139]: v.shape
```

```
Out[139]: (5,)
```

```
In [138]: v
```

```
Out[138]: 0      4
1     11
2      6
3     48
4      6
Name: class, dtype: int64
```

```
In [140]: x = df[['Pgain', 'Vgain']]
```

```
In [141]: x = df.drop('class', axis=1)
```

```
In [142]: x.shape
```

```
Out[142]: (5, 4)
```

```
In [143]: x
```

```
Out[143]:
```

	motor	screw	Pgain	Vgain
0	1	2	5	4
1	3	5	6	5
2	2	3	4	3
3	2	4	3	2
4	5	3	6	5

```
In [110]: from sklearn.model_selection import train_test_split
```

```
In [111]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.3, r:

In [112]: x_train.shape, x_test.shape, y_train.shape, y_test.shape

Out[112]: ((3, 2), (2, 2), (3,), (2,))

In [115]: from sklearn.linear_model import LinearRegression

In [114]: lr = LinearRegression()

In [151]: lr.fit(x_train, y_train)

Out[151]: LinearRegression()

In [118]: y_pred = lr.predict(x_test)

In [79]: y_pred.shape

Out[79]: (2,)

In [119]: y_pred

Out[119]: array([5.33333333, 5.33333333])

In [81]: from sklearn.metrics import mean_squared_error, mean_absolute_error, r

In [120]: mean_squared_error(y_test, y_pred)

Out[120]: 926.2777777777777

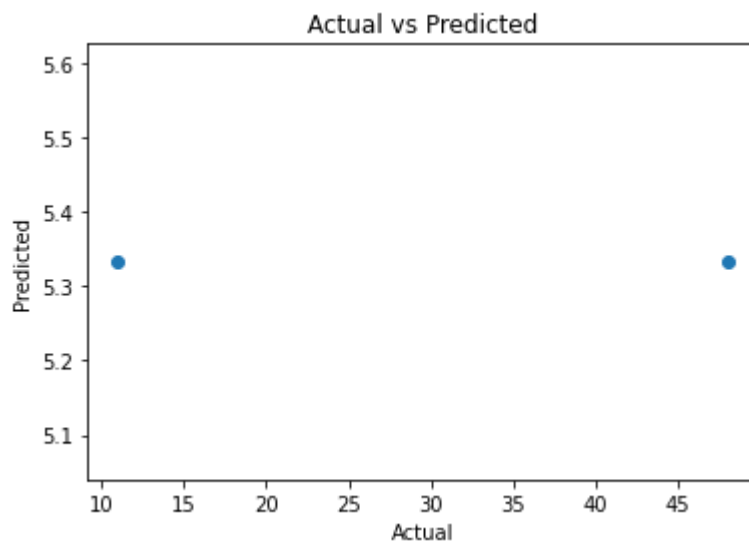
In [121]: mean_absolute_error(y_test, y_pred)

Out[121]: 24.166666666666664

In [122]: r2_score(y_test, y_pred)

Out[122]: -1.7064361658956253
```

```
In [144]: import matplotlib.pyplot as plt
plt.scatter(y_test,y_pred)
plt.xlabel("Actual")
plt.ylabel("Predicted")
plt.title("Actual vs Predicted")
plt.show()
```



```
In [145]: x_new = df.sample(1)
```

```
In [146]: x_new
```

```
Out[146]:
```

	motor	screw	Pgain	Vgain	class
3	2	4	3	2	48

```
In [147]: x_new.shape
```

```
Out[147]: (1, 5)
```

```
In [ ]: y_pred_new = lr.predict(x_new)
```

```
In [ ]: y_pred_new
```

```
In [ ]: A data frame with some observations on variables, 4 nominals and 1 as
of a servo system involving a servo amplifier, a motor, a lead screw/
```

1. Motor A,B,C,D,E
2. Screw A,B,C,D,E
3. Pgain 3,4,5,6
4. Vgain 1,2,3,4,5
5. Class 0.13 to 7.10

```
In [ ]:
```