Simple Linear Regression

```
# from google.colab import files
# up = files.upload()

v import dataset
```

```
import pandas as pd
df = pd.read_csv('dataset.csv')
df.head()
```

```
A B C T

0 2.0 4 8.5 196

1 2.4 4 9.6 221

2 1.5 4 5.9 136

3 3.5 6 11.1 255

4 3.5 6 10.6 244
```

```
A T

0 2.0 196

1 2.4 221

2 1.5 136

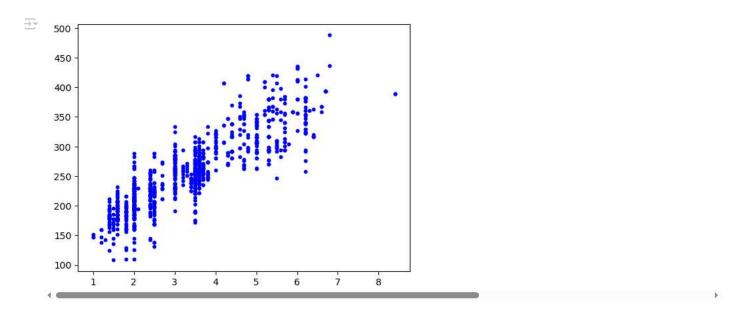
3 3.5 255

4 3.5 244
```

df.size
df.shape
df.info()
df.describe()

 $\overline{2}$ **count** 1067.000000 1067.000000 3.346298 256.228679 mean std 1.415895 63.372304 1.000000 108.000000 min 25% 2.000000 207.000000 50% 3.400000 251.000000 294.000000 4.300000 75% 488.000000 8.400000

```
import matplotlib.pyplot as plt
plt.scatter(df['A'], df['T'],s=10, color='blue')
plt.show()
```



cleaning

```
# clean the data here
```

encoding

do not need for encoding in simple linear regression

y_test = np.array(test[['CO2EMISSIONS']])

from sklearn.metrics import r2_score

from sklearn.linear_model import LinearRegression

finding best random state

1st = []

define x , y

```
import numpy as np
x = np.array(df[['A']])
y = np.array(df['T'])
# x = df[['ENGINESIZE']].values
\# y = df['CO2EMISSIONS'].values
y[:5]
→ array([196, 221, 136, 255, 244], dtype=int64)
spliting
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=42)
### spliting without sklearn
# msk = np.random.rand(len(df)) < 0.8</pre>
# train = df[msk]
\# test = df[\simmsk]
# x_train = np.array(train[['ENGINESIZE']])
# x_test = np.array(test[['ENGINESIZE']])
# y_train = np.array(train[['CO2EMISSIONS']])
```

```
# for i in range(1,10):
     x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=i)
#
      slr = LinearRegression()
#
      slr.fit(x_train, y_train)
#
#
      yhat_test = slr.predict(x_test)
      r2 = r2_score(y_test, yhat_test)
#
#
      lst.append(r2)
# print(f"r2_score: {round(max(lst), 2)}")
\# rs = np.argmax(lst) + 1
# print(f"random state: {rs}")
plt.scatter(x_train, y_train, s=10, c='b')
plt.scatter(x_test, y_test, s=10, c='r')
plt.show()
     500
     450
     400
     350
     300
     250
     200
     150
     100
          1
                       3
                                                       8
```

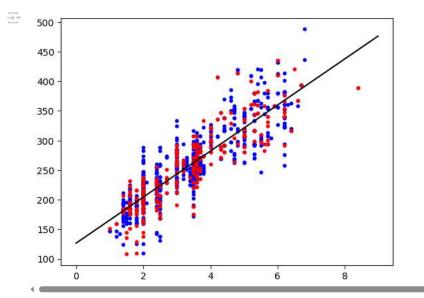
scaling

do not need for scaling in simple linear regression

fit train data

```
500 - 450 - 400 - 350 - 300 - 250 - 200 - 150 - 100 - 0 2 4 6 8
```

```
xx = np.arange(0,9,0.01)
plt.scatter(x_train, y_train, s=10, c='b')
plt.plot(xx, slr.intercept_ + slr.coef_[0] * xx, c='black')
plt.scatter(x_test, y_test, c='r', s=10)
plt.show()
```



K-fold cross validation

```
# from sklearn.linear_model import LinearRegression
# from sklearn.model_selection import GridSearchCV
# parameters = {
      'fit_intercept': [True, False],
#
      'copy_X': [True, False],
#
      'n_jobs': [None],
#
#
      'positive': [True, False]
# lr = LinearRegression()
# gs = GridSearchCV(estimator=lr, param_grid=parameters, cv=5)
# gs.fit(x_train, y_train)
# best_params = gs.best_params_
# print(best_params)
```

predict test data

```
yhat test = slr.predict(x test)
```

v evaluate the model

```
from sklearn.metrics import r2_score
 print("r2-score: %0.2f" % r2_score(y_test, yhat_test))
→ r2-score: 0.77
from sklearn.metrics import mean_squared_error
from sklearn.metrics import mean absolute error
print(f"MSE: {mean_squared_error(y_test, yhat_test)}")
print(f"MAE: {mean_absolute_error(y_test, yhat_test)}")
→ MSE: 972.1181539625659
    MAE: 23.80428191911295
### evaluate without sklearn
# print("MAE: %0.2f" % np.mean(np.absolute(y_test - yhat_test)))
# print("MSE: %0.2f" % np.mean((y_test - yhat_test) ** 2))
predict new data
slr.predict([[0]])
→ array([126.62361301])
save the model
# import joblib
# joblib.dump(slr, 'slr_model.pkl')
load the model
# import joblib
# slr = joblib.load('slr_model.pkl')
```