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

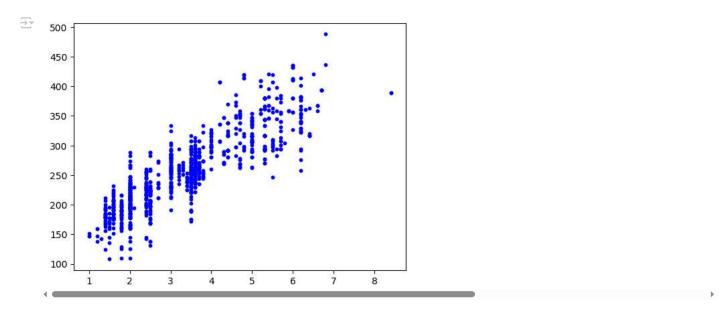
3 3.5 255

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
```

encoding

encode the data

define x , y

```
import numpy as np
x = np.array(df[['A']])
y = np.array(df['T'])

# x = df[['A']].values
# y = df['T'].values

y[:5]
    array([196, 221, 136, 255, 244])

    spliting
```

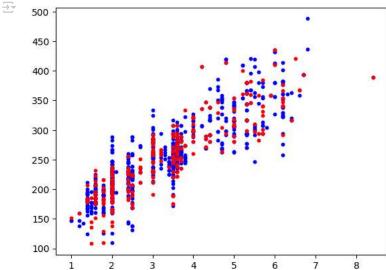
from sklearn.model_selection import train_test_split

```
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
# train = df[msk]
# test = df[~msk]

# x_train = np.array(train[['ENGINESIZE']])
# x_test = np.array(test[['ENGINESIZE']])
# y_train = np.array(train[['CO2EMISSIONS']])
# y_test = np.array(test[['CO2EMISSIONS']])

### finding best random state
# from sklearn.model_selection import train_test_split
# from sklearn.linear_model import LinearRegression
# from sklearn.metrics import r2_score</pre>
```

```
# lst = []
# 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()
```



scaling

do not need for scaling in simple linear regression

fit train data

```
500 -

450 -

400 -

350 -

300 -

250 -

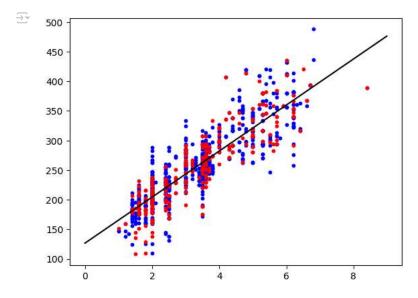
200 -

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: 972.1181539625654
    MAE: 23.804281919112945
### evaluate without sklearn
# print("MSE: %0.2f" % np.mean((y_test - yhat_test) ** 2))
# print("MAE: %0.2f" % np.mean(np.absolute(y_test - yhat_test)))
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')
```