```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
```

## Creating a Dataset

```
X=np.random.randint(1,100,100)
X=X.reshape(-1,1)
      array([[43],
             [86],
             [59],
             [62],
             [98],
             [87],
             [71],
             [69],
              [21],
              [30],
             [73],
             [70],
             [44],
             [58],
             [43],
             [96],
             [15],
             [26],
             [35],
             [69],
             [8],
             [60],
             [56],
             [94],
             [54],
             [55],
             [30],
             [46],
             [51],
             [73],
             [29],
             [85],
             [73],
             [51],
             [47],
             [67],
             [77],
             [13],
             [12],
              [6],
             [16],
             [15],
             [12],
              [21],
             [36],
             [20],
             [90],
             [76],
             [21],
             [28],
             [41],
             [52],
             [85],
             [42],
             [53],
             [92],
             [60],
             [61],
```

## Create Errors and add to y

[0.92143322], [0.10682987],

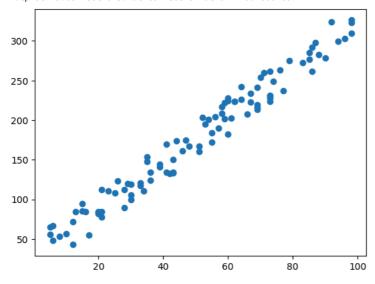
```
[0.96528252],
            [0.56597248],
             [0.23142304],
            [0.85187746],
            [0.81636823],
            [0.66350869],
            [0.0605795],
            [0.26933085],
            [0.78377649],
            [0.87890487],
            [0.84196591],
             [0.66219466],
             [0.56087308],
            [0.03219164],
            [0.70439104],
            [0.33056653],
            [0.75385337],
            [0.1236444 ],
            [0.2872219],
            [0.44533002],
            [0.25732365],
            [0.16141242],
             [0.64603882],
            [0.41596631],
            [0.82746091],
            [0.12040676],
            [0.65528035],
            [0.63516401],
            [0.09469582],
            [0.89392777],
            [0.70314239],
            [0.96119862],
            [0.71745855],
            [0.97998839],
            [0.11660079],
            [0.26869413],
            [0.3006044],
            [0.45511645],
            [0.15454835],
            [0.68529609],
            [0.41087612],
            [0.0910359],
            [0.91144231],
            [0.92977831],
             [0.58725193],
            [0.11645415],
             [0.69403651],
            [0.94447958],
            [0.93836521],
v=3*X+1+err*50
y=y.reshape(-1,1)
     array([[134.23701478],
            [291.93060922],
            [201.86427508],
            [223.71533429],
            [323.22524203],
            [297.8735401],
            [260.07166086],
            [213.34149333],
            [112.26412604],
            [119.29862387],
            [231.57115184],
            [253.59387311],
            [173.81841165],
             [208.17543429],
             [133.02897487],
            [302.46654275],
            [ 85.18882458],
            [122.94524342],
            [148.09829571],
            [241.10973309],
            [ 53.04365394],
            [182.60958177],
            [204.21955192],
            [299.52832647],
            [200.69266863],
            [172.18221988],
             [105.36109478],
            [161.26650106],
             [166.86618233],
            [228.07062108],
            [120.30194123],
```

```
[276.79831574],
[261.37304537],
[160.02033777],
[174.76401764],
[233.7582004],
[236.73479123],
[ 84.69638825],
[ 72.15711972],
  67.05993079],
 84.87292766],
 94.99941959],
[ 42.83003957],
77.43470653],
[124.03022016],
 83.75582237],
[278.72741733],
[263.26480457],
[ 84.54380601],
[ 89.55179478],
[169.57211531],
[203.48891536],
[285.36259635],
[132.82270741],
[194.70182527],
[324.22397882],
[227.91826025],
[202 7065/500]
```

## Visualizing the data

plt.scatter(X,y)

<matplotlib.collections.PathCollection at 0x7f7ad18b07c0>



## Split the dataset

#### scale the features

```
X_{train\_scaled=sc.transform(X_{train})}
X_train_scaled[:5]
     array([[ 1.33401317],
             [ 1.25770208],
             [ 0.87614666],
             [-0.76454169]
             [ 0.34196905]])
X_test_scaled=sc.transform(X_test)
X_test_scaled[:5]
     array([[ 0.53274677],
             [ 0.03672471],
             [-1.41318592],
             [ 0.87614666]
             [-0.53560843]])

    Add column X0

X_train_scaled=np.insert(X_train_scaled,0,1,axis=1)
X_test_scaled=np.insert(X_test_scaled,0,1,axis=1)
X train scaled.shape
     (80, 2)
X_test_scaled.shape
     (20, 2)
X_train_scaled.min(),X_train_scaled.max()
     (-1.7184302659385815, 1.8300352290886308)

    Initialize the Parameters and Hyperpara-meters

alpha=0.05
n=len(X_train)
theta=np.array([0.5,0.5])
theta
     array([0.5, 0.5])

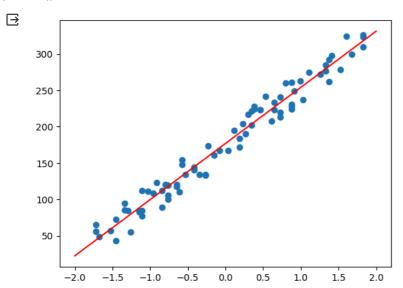
✓ Algorithm

# For a single column
# y cap=np.dot(X train scaled[0],theta)
 \begin{tabular}{ll} # theta-alpha*(y_cap-y[0]*X_train_scaled[0]) \\ \end{tabular} 
for epoch in range(5000):
    gradient=0
    for i in range(n):
        y_cap=np.dot(X_train_scaled[i],theta)
        gradient+=(y_cap-y_train[i])*X_train_scaled[i]
    gradient=(gradient)/n
    theta=theta-alpha*gradient
     array([176.9481513 , 77.24699918])
```

Visualizing the new graph with the line plotted using the update theta

```
plt.scatter(X_train_scaled[:,1],y_train)
X_plot=np.linspace(-2,2,100)
y_plot=(theta[1]*X_plot)+theta[0]
plt.plot(X_plot,y_plot,'r-')
```

plt.show()



### → R<sup>2</sup> evaluation

```
y_pred=np.dot(X_test_scaled,theta)
y_pred
       array([218.1012405 , 179.78502495, 67.78377952, 244.62785126,
               135.57400702, 167.99542017, 188.62722854, 288.83886919,
               232.83824647, 156.20581539, 150.311013 , 200.41683332, 153.25841419, 209.25903691, 47.15197115, 182.73242615,
               235.78564767, 206.31163571, 53.04677355, 312.41807875])
y_test=y_test.flatten()
y_test
       array([226.49423741, 160.02033777, 84.69638825, 231.57115184,
               124.03022016, 174.76401764, 200.69266863, 282.34438437, 216.07768966, 150.45118274, 169.57211531, 208.17543429,
               132.82270741, 202.70654509, 67.05993079, 203.48891536, 253.59387311, 182.60958177, 53.04365394, 302.46654275])
sstot=(y_test-np.mean(y_test))**2
ssres=(y_test-y_pred)**2
r2=1-(np.sum(ssres)/np.sum(sstot))
       0.9488622687901136
a=y_pred-y_test
(np.sqrt(np.abs(a)))
       array([2.89706695, 4.44574934, 4.11249422, 3.61340552, 3.39761488,
               2.60165283,\ 3.47353424,\ 2.54842791,\ 4.0939659\ ,\ 2.39888154,
               4.38874724, 2.78542653, 4.520587 , 2.55978355, 4.46183366, 4.55592902, 4.21997932, 4.86847553, 0.05585344, 3.15460552])
```

# → Calculating R<sup>2</sup> using Metrics

```
from sklearn.metrics import r2_score
r2_score(y_pred,y_test)
    0.9540856405763084
```

#### Calculating r^2 using Scikit Model

from sklearn.linear\_model import LinearRegression
model=LinearRegression()

```
{\tt model.fit(X\_train\_scaled,y\_train)}
      ▼ LinearRegression
     LinearRegression()
y_pred=model.predict(X_test_scaled)
y_pred=y_pred.reshape(-1,1)
y_pred
     array([[218.1012405],
            [179.78502495],
            [ 67.78377952],
             [244.62785126],
             [135.57400702],
             [167.99542017],
             [188.62722854],
            [288.83886919],
            [232.83824647],
            [156.20581539],
            [150.311013 ],
[200.41683332],
            [153.25841419],
             [209.25903691],
             [ 47.15197115],
             [182.73242615],
             [235.78564767],
             [206.31163571],
             [ 53.04677355],
             [312.41807875]])
y_test=y_test.reshape(-1,1)
r_squared = model.score(X_test_scaled, y_test)
r_squared
     0.9488622687901133
model.coef_
     array([[ 0. , 77.24699918]])
```