

## Diabetes dataset using sklearn

```
%matplotlib inline
import numpy as np
import matplotlib
import matplotlib.pyplot as plt
from sklearn import datasets, linear_model, metrics
```

```
diabetes = datasets.load_diabetes()
X = diabetes.data
y = diabetes.target

samp, natt = X.shape

for i in range(natt):
    attr_mean = np.mean(X[:,i])
    attr_std = np.std(X[:,i])
    X[:,i] = (X[:,i] - attr_mean)/attr_std
```

```
print(np.mean(X, axis = 0))
print(np.std(X, axis = 0))
```

```
↳ [-9.54490383e-18 -4.21985222e-17 -5.52599696e-17 -4.82268825e-17
    5.52599696e-18 -1.35638107e-17 -2.81323481e-17 -1.48448373e-16
    -1.84932512e-17 -2.67508489e-17]
    [1. 1. 1. 1. 1. 1. 1. 1. 1.]
```

```
samp, natt = X.shape
```

```
samp, natt
```

```
↳ (442, 10)
```

```
regressor = linear_model.LinearRegression(fit_intercept=False, normalize=False)
regressor.fit(X, y)
```

```
↳ LinearRegression(copy_X=True, fit_intercept=False, n_jobs=None, normalize=False)
```

```
regressor.intercept_
```

```
↳ 0.0
```

```
regressor.coef_
```

```
↳ array([ -0.47623169, -11.40703082,  24.72625713,  15.42967916,
    -37.68035801,  22.67648701,   4.80620008,   8.422084  ,
    35.73471316,   3.21661161])
```

```
y_pred = regressor.predict(X)
RSS = np.mean((y_pred-y)**2)/(np.std(y)**2)
```

```
Rsq = 1 - RSS
print(Rsq)
mse = np.mean((y_pred-y)**2)
print(mse)
```

```
↳ -3.385293788056333
    26004.28740231017
```

## Diabetes dataset without using sklearn

```
def MSE(X,y,w):
    y_pred = X.dot(w)
    return (1*np.linalg.norm(y_pred-y)**2)/442

def MSE_gradient(X,y,w):
    return (X.T.dot(X.dot(w)-y))/442

def gradient_descent(init, steps, grad):
    xs = [init]
    for step in steps:
        xs.append(xs[-1] - step * grad(X,y,xs[-1]))
    return xs

n,d = samp,natt
w0 = np.random.normal(0,1,d)
ws = gradient_descent(w0,[0.1]*6000,MSE_gradient)

all_mse = []
for w in ws:
    mse = MSE(X,y,w)
    all_mse.append(mse)

for i in range(len(all_mse)):
    if i%500==0:
        print(f'MSE at iteration {i} = {all_mse[i]}')

↳ MSE at iteration 0 = 29139.86111223589
    MSE at iteration 500 = 26013.848159460198
    MSE at iteration 1000 = 26008.347652765853
    MSE at iteration 1500 = 26006.011708990412
    MSE at iteration 2000 = 26005.01968066765
    MSE at iteration 2500 = 26004.598386196605
    MSE at iteration 3000 = 26004.419470910365
    MSE at iteration 3500 = 26004.34348918772
    MSE at iteration 4000 = 26004.3112212787
    MSE at iteration 4500 = 26004.297517746927
    MSE at iteration 5000 = 26004.291698132663
    MSE at iteration 5500 = 26004.289226659563
    MSE at iteration 6000 = 26004.288177074657

y_prediction = X.dot(ws[-1])
RSS_manual = np.mean((y_prediction-y)**2)/(np.std(y)**2)
Rsq_manual = 1-RSS;Rsq
```



-3.385293788056333