Optimization Problem

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¹EE16BTECH11043

November 14, 2019

Problem

► Find w such that,

$$L = ||y - Xw||^2 \tag{1}$$

is minimized. where,

$$X = \begin{pmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{pmatrix} \tag{2}$$

$$y = \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix} \tag{3}$$

Least Squares Method

$$\widehat{w} = \min_{X} ||y - Xw||^{2}$$

$$= (X^{T}X)^{-1}X^{T}y$$
(5)

$$= \left(X^T X\right)^{-1} X^T y \tag{5}$$

Gradient Descent

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$$w_{n+1} = w_n - \mu \frac{\partial L}{\partial w} \tag{6}$$

Till,

$$w_{n+1} \approx w_n \tag{7}$$

Python code

```
#Gradient Descent
#Least Squares
import numpy as np
c = 2
cur_{-}w = np. array([[1.0], [1.0]])
# The algorithm starts at w = [1,1]
gamma = 0.01 # step size multiplier
precision = 1e-6
previous_step_size = 1
\max iters = 10000
# maximum number of iterations
iters = 0 #iteration counter
X = np. array([[1.0,0.0],[1.0,1.0],[1.0,2.0]])
y = np.array([[6.0],[0.0],[0.0])
```

Python code

```
df = lambda \times, y, w:
        (-2.0*np.matmul(x.T,(y-np.matmul(x,w))))
while (previous_step_size > precision)
                 & (iters < max_iters):
    prev_w = cur_w
    cur_w = gamma * df(X,y,prev_w)
    previous_step_size =
                np.linalg.norm(cur_w - prev_w)
    iters+=1
```

Python code

```
Gradient_descent = curr_w

# equation for least squares
least_squares = np.matmul(
np.linalg.inv(np.matmul(X.T,X)),(np.matmul(X.T,y)))
```

solution from Gradient Descent

Answer

Answer:

► The minimum occurs at(gradient Descent)

$$w = \begin{pmatrix} 1.0 \\ 0.84 \end{pmatrix} \tag{8}$$

The minimum occurs at(least squares)

$$w = \begin{pmatrix} 1.0 \\ 0.84 \end{pmatrix} \tag{9}$$