

Section D - Ridge Regression

Q5 If $E(w)$ is the objective, which is differentiable in the neighbourhood of point a , in order to decrease the error function we move in the direction -gradient i.e. $-\nabla E(a)$

$$\therefore b = a - \eta \nabla E(a) \quad (1)$$

where η - learning rate
 $\nabla E(a)$ - gradient of error

Steps

(1) Initialize $w(0)$ and $t=1$.

(2) while $\eta' > \epsilon$

$$w = w^{(t-1)} - \eta' \nabla E(w^{(t-1)})$$

If $E(w) < E(w^{(t-1)})$, break

$$\eta = \frac{\eta'}{L}$$

(3) $w^{(t)} = w$

(4) $t = t + 1$

Let the original loss function be L_0 , to add L_2 regularization term,

$$\bar{L}(w) = L_0(w) + \lambda \|w\|_2^2$$

or

$$\nabla E(w) = - \sum_{n=1}^N (t_n - w \cdot \phi(x_n)) \phi(x_n) + 2 \lambda w$$

where λ - regularization parameter.

SGD - We don't require whole data set in one go, we update the model parameters after accessing each data pt.

Error function for SGD

$$E_n(w) = \frac{1}{2} (t_n - w^T \phi(x_n))^2 + \lambda \|w\|_2^2$$

Steps -

1) Initialize the parameter for $w^{(0)} = 1$ $z = 1$

2) while !

→ select random pt (x_n, y_n)

$$\rightarrow w^{(z)} = w^{(z-1)} - \eta^{(z)} \nabla E_n(w^{(z-1)}) - \lambda \|w\|^2$$

→ $z = z + 1$

where $z \rightarrow$ Iteration

$\lambda \rightarrow$ Regularization parameter.