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Question 32212

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Q1:

(a) No.

(b) We assume N is normally distributed with mean 0 and variance σ^2 .

(c) Given cost function: $J(\mathbf{w}) := \frac{1}{n} \sum_{j=1}^n (\mathbf{w} \cdot \mathbf{x}^{(j)} - y^{(j)})^2 + \alpha \|\mathbf{w}\|_2$, where $\|\mathbf{w}\|_2 = \sqrt{\mathbf{w} \cdot \mathbf{w}}$

is the Euclidean norm, we have:

$$\begin{aligned} \nabla J(\mathbf{w}) &= \frac{2}{n} \sum_{j=1}^n (x^{(j)} \mathbf{w} - y^{(j)}) \cdot x^{(j)} + \alpha \cdot 2\mathbf{w} \cdot \frac{1}{2\sqrt{\mathbf{w} \cdot \mathbf{w}}} \\ &= \frac{2}{n} \sum_{j=1}^n [(x^{(j)})^2 \cdot \mathbf{w} - x^{(j)} y^{(j)}] + \alpha \frac{\mathbf{w}}{\|\mathbf{w}\|_2} \\ &= \left[\frac{2}{n} \sum_{j=1}^n (x^{(j)})^2 + \frac{\alpha}{\|\mathbf{w}\|_2} \right] \mathbf{w} - \frac{2}{n} \sum_{j=1}^n x^{(j)} y^{(j)} \end{aligned}$$

This equation is not linear, thus cannot be solved directly.

One can set the model parameters using gradient descent:

Initiate with parameter \mathbf{w} , learning rate $e > 0$.

Repeat:

$$\mathbf{w} := \mathbf{w} - e \cdot \nabla J(\mathbf{w})$$

(d) We update $\nabla J(\mathbf{w})$ using the given conditions that $y^{(j)} - \mathbf{w} \cdot \mathbf{x}^{(j)} = 0$ except $j = 1$:

$$\begin{aligned}\nabla J(\mathbf{w}) &= \frac{2}{n} \sum_{j=1}^n (x^{(j)} \mathbf{w} - y^{(j)}) \cdot x^{(j)} + \alpha \frac{\mathbf{w}}{\|\mathbf{w}\|_2} \\ &= \frac{2}{n} (x^{(1)} \mathbf{w} - y^{(1)}) \cdot x^{(1)} + \alpha \frac{\mathbf{w}}{\|\mathbf{w}\|_2} \\ &= \frac{2}{n} \left(-\delta \cdot x^{(1)} + \frac{\alpha n}{2\|\mathbf{w}\|_2} \right)\end{aligned}$$

According to the updating rule of gradient descent:

$$w := w - e \cdot \nabla J(w)$$

To increase parameter w_1 , given learning rate > 0 , we need $\nabla J(w_1) < 0$. Thus:

$$\begin{aligned}\nabla J(w_1) &= \frac{2}{n} (-\delta \cdot x^{(1)} + \frac{\alpha n}{2\|\mathbf{w}\|} w_1) < 0 \\ \frac{\alpha n}{2\|\mathbf{w}\|} w_1 &< \delta \cdot x^{(1)}\end{aligned}$$

Given $w_1 > 0$:

$$\alpha < \frac{2\delta \cdot x^{(1)} \cdot \|\mathbf{w}\|_2}{nw_1}$$

Since $\delta > 0$, $x^{(1)} > 0$, $n \in \mathbb{N}^+$, we verify that $\alpha > 0$. Thus:

For $\alpha \in (0, \frac{2\delta \cdot x^{(1)} \cdot \|\mathbf{w}\|_2}{nw_1})$, the model parameter w_1 will increase.

Q2:

(a) The underlying idea of using “dropout” is to approximate the ensemble methods on a large number of neural networks to reduce generalisation error.

(b) During training of the neural network. It implicitly averages outputs from all subnetworks to learn the weights for the full network. When testing or using the network, the full neural network should be used, otherwise will return stochastic results from some subnetworks.

(c) Given dropout rate p unchanged, define a noise term $\epsilon \sim \mathcal{N}(1, \sigma^2)$ on the probability of the Bernoulli distribution $d_j^l \sim \text{Ber}(1 - p)$, redefine the dropout activation unit as:

$$\tilde{a}_j^l = \frac{1}{1 - p \cdot \epsilon} \cdot \text{Ber}(1 - p \cdot \epsilon) \cdot \phi(z_j^l)$$

One can show that the expectation of the activation unit stays:

$$\begin{aligned}\mathbb{E}[\tilde{a}_j^l] &= p\epsilon \cdot \frac{1}{1 - p\epsilon} \cdot 0 \cdot \phi(z_j^l) + (1 - p\epsilon) \cdot \frac{1}{1 - p\epsilon} \cdot 1 \cdot \phi(z_j^l) \\ &= 0 + \phi(z_j^l) = a_j^l\end{aligned}$$

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