

Solⁿ 1

$$J(w) = \frac{1}{2n} \sum_{i=1}^n (\hat{y}^{(i)} - y^{(i)})^2$$

$$\hat{y} = X^T w$$

$$\nabla_w J(w) = \frac{1}{2n} \sum_{i=1}^n X (X^T w - y) \quad \text{--- (1)}$$

$$\text{Hessian for } J(w), \text{ (w.r.t. } w) \quad \text{--- (2)}$$
$$\frac{1}{n} X X^T$$

According to Newton's method -

$$w^* = w^{(0)} - H(f)(w^{(0)})^{-1} \nabla f(w^{(0)}) \quad \text{--- (3)}$$

Putting (1) & (2) in (3) ~~$f(w) = J(w)$~~

$$w^* = w^{(0)} - \frac{1}{n} (X X^T)^{-1} \left[\frac{1}{n} (X X^T w^{(0)} - X y) \right]$$

$$= w^{(0)} - (X X^T)^{-1} (X X^T w^{(0)} - X y)$$

$$w^* = w^{(0)} - \cancel{X X^T} w^{(0)} + (X X^T)^{-1} X y$$

$$(\because (X X^T)^{-1} (X X^T) = I)$$

Hence, $w^* = (X X^T)^{-1} X y$

Solⁿ 2

Given that,

$$\hat{y}_K = \frac{e^{z_K}}{\sum_{k=1}^C e^{z_K}}$$

$$z_K = x^T w^{(K)} + b$$

$$\nabla_{w^{(l)}} \text{loss}(w, b) = -\frac{1}{n} \sum_{i=1}^n \sum_{K=1}^C y_K^{(i)} \nabla_{w^{(l)}} \log \hat{y}_K^{(i)}$$

⇒ To find, $\nabla_{w^{(l)}} \hat{y}_K^{(i)}$ when $l=K$

$$\hookrightarrow \nabla_{w^{(K)}} \hat{y}_K^{(i)} = \frac{\left(\sum_{k=1}^C e^{z_K} \right) \nabla_w z_K - e^{z_K} \nabla_w \sum_{k=1}^C e^{z_K}}{\left(\sum_{k=1}^C e^{z_K} \right)^2}$$

∴ $K=1$

$$\nabla_{w^{(1)}} \hat{y}_K^{(i)} = \frac{\left(\sum_{k=1}^C e^{z_K} \right) \nabla_w z_K - (e^{z_K} e^{z_K})}{\left(\sum_{k=1}^C e^{z_K} \right)^2} \nabla_w z_K$$

$$\boxed{\nabla_w z^1 = x}$$

$$= \frac{x e^{z_K}}{\left(\sum_{k=1}^C e^{z_K} \right)} - \frac{x e^{z_K^2}}{\left(\sum_{k=1}^C e^{z_K} \right)^2}$$

$$= x \hat{y}_K - x \hat{y}_K^2$$

$$\nabla_{w^{(K)}} \hat{y}_K^{(i)} = x^{(i)} \hat{y}_K^{(i)} (1 - \hat{y}_K^{(i)})$$

⇒ To find $\nabla_{w^{(1)}} \hat{y}_K^{(i)}$ when $l \neq K$.

$$= \frac{\left(\sum_{k=1}^L e^{z_k} \right) \nabla_{w^{(1)}} e^{z_k} - \left(e^{z_l} \nabla_{w^{(1)}} \sum_{k=1}^L e^{z_k} \right) (\nabla z_k)}{\left(\sum_{k=1}^L e^{z_k} \right)^2} \quad (\nabla z^l = X)$$

∵ $k \neq l$ $\nabla_{w^{(1)}} e^{z_k} = 0$ ∴ $\left(\nabla_{w^{(1)}} \sum_{k=1}^L e^{z_k} = e^{z_l} \right)$

Hence

above equation becomes -

$$= \frac{0 - X e^{z_k} e^{z_l}}{\left(\sum_{k=1}^L e^{z_k} \right)^2}$$

$$= - \frac{X^{(i)} \hat{y}_K^{(i)} \hat{y}_l^{(i)}}{\hat{y}_K \hat{y}_l}$$

⇒ $\nabla_{w^{(1)}} f_{CE}(w, b) = - \frac{1}{n} \sum_i \sum_{k \neq l} y_k^{(i)} \nabla_{w^{(1)}} \log \hat{y}_K^{(i)}$

From above 2nd solⁿs we know $\nabla_{w^{(1)}} \hat{y}_K^{(i)}$ at $K=l$ & $K \neq l$.

Hence,

$$= \frac{-1}{n} \sum_i \left[\sum_{k=1}^L \frac{y_k}{\hat{y}_k} (X \hat{y}_k - X \hat{y}_k \hat{y}_k) + \sum_{k \neq l} \frac{y_k X \hat{y}_k \hat{y}_l}{\hat{y}_k \hat{y}_l} \right]$$

$$= \frac{-1}{n} \sum_i \left(\sum_{k=1}^L y_k X - \sum_{k=1}^L X \hat{y}_k y_k \right) - \left(\sum_{k \neq l} y_k X \hat{y}_l \right)$$

$$= \frac{-1}{n} \left[\sum_{k \neq l} y_k X y_l - \sum_{k=1}^L X \hat{y}_k y_l - \sum_{k \neq l} y_k X \hat{y}_l \right]$$

On combining $K=l \neq K' \neq l$

$$-\frac{1}{n} \left[\sum_{l=K}^C x y_l - \sum_{K=1}^C x \hat{y}_l y_{lk} \right]$$

$\because \sum_{K=1}^C y_K = 1$ because of softmax.

Hence

$$-\frac{1}{n} \left[\sum_{l=K}^C x^{(i)} y_l^{(i)} - x^{(i)} \hat{y}_l^{(i)} y_{lk}^{(i)} \right]$$

$$= -\frac{1}{n} \sum_{i=1}^n x^{(i)} (y_l^{(i)} - \hat{y}_l^{(i)})$$

\Rightarrow Using similar approach as previous parts, on differentiating \hat{y}_l with respect to b , $\nabla_{w^l} z^k$ becomes 1

Hence,

$$\nabla_{w^l} \hat{y}_K = \frac{\left[\sum_{K'=1}^C e^{z_{K'}} (e^{z_l} - e^{z_l^2}) \right] \nabla_{w^l} z^l}{\left(\sum_{K'=1}^C e^{z_{K'}} \right)^2}$$

$$= (\hat{y}_l - \hat{y}_l^2)$$

$\nabla_{w^l} z^l = 1$

For $l \neq K$

$$\nabla_{w^l} \hat{y}_K = 0 - \frac{e^{z_K} e^{z_l}}{\left(\sum_{K'=1}^C e^{z_{K'}} \right)^2}$$

For $l \neq k$

$$= -\hat{y}_k \hat{y}_l$$

Combining the 2 cases, we get -

$$\nabla_{w(k)} f_{CE}(w, b) = \frac{-1}{n} \sum_{i=1}^n \left[\sum_{k \neq l} y_{ik} (\hat{y}_l - \hat{y}_k) - \sum_{l \neq k} y_{kl} (\hat{y}_k \hat{y}_l) \right]$$

Since $k \neq l$

$$= \frac{-1}{n} \sum_{i=1}^n \left[\sum_{l \neq k} y_{il} - \sum_{l \neq k} \hat{y}_k y_{il} - \sum_{l \neq k} y_{il} \hat{y}_l \right]$$

$\hat{y}_k = \hat{y}_l$

On combining $k = l$ & $k \neq l$,

$$= \frac{-1}{n} \sum_{i=1}^n \left[\sum_{l=1}^n y_{il} - \sum_{k=1}^n y_{ik} \hat{y}_k \right]$$

$$\sum_{k=1}^n \hat{y}_k = 1$$

$$= \frac{-1}{n} \sum_{i=1}^n \left(y_{il} - \hat{y}_l \right)$$

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