

Faiyaz Sundrani fs1459
Ilyas Habeeb mih278

H.W 3

Part-I (Theory)

$$1. h(x) = \frac{1}{1 + e^{-(w^T x + w_0)}}$$

if $h(x) > \frac{1}{2}$, then classify as $C_1(1)$

else as $C_2(0)$

$w^T x + w_0 \geq 0$, classify as $C_1(1)$

else $C_2(0)$

	x_1	x_2	y
1	-5	3	1
2	2	3	0

Predict $w_2 x_2 + w_1 x_1 + w_0$

At start, $w_0, w_1, w_2 = 0.01$

$$\eta = 0.005$$

(a)

$$P(C_1 | x^{(1)}) = \frac{1}{1 + e^{-(3w_0 + 5w_1 x_1 + w_2)}}$$

$$\because w_0, w_1, w_2 = 0.01$$

$$= \frac{1}{1 + e^{-(0.03 - 0.05 + 0.01)}}$$

$$= \frac{1}{1 + e^{0.01}} = \underline{\underline{0.497}}$$

$$\therefore P(C_1 | x^{(1)}) = 0.497$$

(b)

Since $P(C_1 | x^{(1)}) < \frac{1}{2}$, the predicted label
of $x^{(1)}$ is $\boxed{C_2 \text{ (i.e } \gamma = 0)}$

$$(c) P(C_1 | x^{(2)}) = \frac{1}{1 + e^{-(3w_2 + 2w_1 + w_0)}}$$

$$= \frac{1}{1 + e^{-(0.03 + 0.02 + 0.01)}}$$

$$\therefore P(C_1 | x^{(2)}) = 0.514$$

Since $P(C_1 | x^{(2)}) > \frac{1}{2}$, the predicted label

of $x^{(2)}$ is G (ie $\gamma = 1$)

	x_1	x_2	γ (Output)	Predicted
1	-5	3	1	0
2	2	3	0	1

d) Training Error = % of labels predicted inaccurately

$$= \frac{\# \text{predicted inaccurately}}{\# \text{total}} \times 100$$

$$= \frac{2}{2} \times 100 = 100\%$$

(e)

$$\text{Err}(\omega, \omega_0 | X) = - \sum_t (\gamma^t \log y^t + (1-\gamma^t) \log (1-y^t))$$

Cross Entropy Error

Since $y^{(1)} = 0.497$, $\gamma^{(1)} = 1$
 $y^{(2)} = 0.514$, $\gamma^{(2)} = 0$

$$\begin{aligned}\therefore \text{Err}(\omega, \omega_0 | X) &= -[(\gamma^{(1)} \log y^{(1)} + (1-\gamma^{(1)}) \log (1-y^{(1)})) \\ &\quad + \gamma^{(2)} \log y^{(2)} + (1-\gamma^{(2)}) \log (1-y^{(2)})] \\ &= -[(1 \cdot \log(0.497) + (1-1) \dots) + (0 \cdot \dots) + (1-0) \\ &\quad \log(1-0.514))] \\ &= -[\log(0.497) + \log(0.486)] \\ &= -[-1.4207] = 1.4207\end{aligned}$$

$$\therefore \text{Err}(\omega, \omega_0 | X) = 1.4207$$

$$P) w_0 \leftarrow w_0 + \eta \sum_t (y^t - \hat{y}^t)$$

$$w_1 \leftarrow w_1 + \eta \sum_t (y^t - \hat{y}^t) x_1$$

$$w_2 \leftarrow w_2 + \eta \sum_t (y^t - \hat{y}^t) x_2$$

$$\text{Since } \eta = 0.005$$

$$\text{Initial } w_0, w_1, w_2 = 0.01$$

$$w_0 \leftarrow 0.01 + 0.005 [(1 - 0.497) + (0 - 0.514)]$$

$$\Rightarrow w_0 \leftarrow 0.009945$$

$$w_1 \leftarrow 0.01 + 0.005 [(1 - 0.497) + (0 - 0.514)(2)]^{(-5)}$$

$$\Rightarrow w_1 \leftarrow -0.0077$$

$$w_2 \leftarrow 0.01 + 0.005 [(1 - 0.497)(3) + (0 - 0.514)(3)]$$

$$\Rightarrow w_2 \leftarrow 0.009805$$

g)

$$y^{(1)} = \frac{1}{1 + e^{-(0.009805 \cdot 3 + 0.0077 \cdot 5 + 0.0009935)}}$$
$$= \frac{1}{1 + e^{-0.02395}}$$

$$\therefore y^{(1)} = 0.519$$

$$y^{(2)} = \frac{1}{1 + e^{-(0.009805 \cdot 3 - 0.0077 \cdot 2 + 0.0009935)}}$$
$$= \frac{1}{1 + e^{-0.02395}}$$

$$\therefore y^{(2)} = 0.506$$

$$E_{\text{RR}}(\omega, \omega_0 | X) = -\log(0.519) - \log(1-0.506)$$

$$\therefore E_{\text{xx}}(\omega, \omega_0 | \pi) = 1.36$$

(h) The cross entropy error ~~were~~ went down after one iteration from 1.4207 to 1.36.

I expected this, since we are optimizing the weights in Gr. D to reduce our cross-entropy error.

(i) The training error of classifier has gone down from $\sqrt{100\%}$ to 50% , since it is classifying $y^{(1)}$ correctly (as $C_1, \gamma=1$)

$\leftarrow T$

Q.

a)

Why it happens: Learning rate is too high

Therefore, it oscillates to and fro from the minimum and never converges.

How to fix : Reduce Learning Rate

b)

Why it happens : Learning rate too slow

Thus, will need a lot of iterations to converge

How to fix : Increase Learning Rate

c)

Why it happens : Converges to local minimum instead of global minimum

How to fix : Randomly choose starting points

& pick the one with lowest minimum

3.

Ans Given :

$$\frac{1}{1 + e^{-(\omega^T x + w_0)}} > 0.30$$

$$1 > 0.30 + 0.30 e^{-(\omega^T x + w_0)}$$

$$0.70 > 0.30 e^{-(\omega^T x + w_0)}$$

$$\frac{0.70}{0.30} > e^{-(\omega^T x + w_0)}$$

$$\frac{7}{3} > e^{-(\omega^T x + w_0)}$$

Apply log on both sides

$$\log e^{-(\omega^T x + w_0)} < \log(7) - \log(3)$$

$$-(\omega^T x + w_0) < 1.9459101491 - 1.098612287$$

$$\omega^T x + w_0 > 0.8472978604$$

∴ When $w^T x + w_0 \geq 0.847$, classify as C_1
 else, as C_2

Ans -

Given:

$$Err_{\text{reg}}(w, w_0) = \left[-\sum_t (\gamma^t \log y^t + (1-\gamma^t) \log (1-y^t)) \right. \\ \left. + \frac{\lambda}{2} \sum_{j=1}^d w_j^2 \right]$$

When $j = 0$,

$$\frac{\partial Err_{\text{reg}}}{\partial w_j} = -\sum_t (\gamma^t - y^t)$$

When $j > 0$

$$\frac{\partial Err_{\text{reg}}}{\partial w_j} = -\sum_t (\gamma^t - y^t)x_j + \lambda w_j$$

w.k.t

$$w_j \leftarrow w_j - \eta \frac{\partial E_{\text{xx}}}{\partial w_j} \text{ for } j=0, 1, \dots, d$$

$$\therefore w_j \leftarrow w_j - \eta \left(-\sum_t (x^t - y^t) \right)$$

$$\Rightarrow \boxed{w_j \leftarrow w_j + \eta \sum_t (x^t - y^t) \text{ for } j=0}$$

$$\therefore w_j \leftarrow w_j - \eta \left(-\sum_t (x^t - y^t)x_j + \lambda w_j \right)$$

$$\Rightarrow \boxed{w_j \leftarrow w_j + \eta \left(\sum_t (x^t - y^t)x_j - \lambda w_j \right) \text{ for } j > 0}$$

----- X ----- X ----- X -----