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Section: 18

Date: 15.05.2025

Assignment

CSE422

Problem - 01

a) Total instances = 14

$$\text{Yes} = 9$$

$$\text{No} = 5$$

$$\text{Entropy} \rightarrow H(D) = - \left(\frac{9}{14} \log_2 \frac{9}{14} + \frac{5}{14} \log_2 \frac{5}{14} \right)$$

$$\approx 0.9402$$

b)

i) Sunny (5 instances):

$$\text{Yes} = 2$$

$$\text{No} = 3$$

$$H(D_{\text{sunny}}) = - \left(\frac{2}{5} \log_2 \frac{2}{5} + \frac{3}{5} \log_2 \frac{3}{5} \right)$$

$$\approx 0.971$$

ii) Overcast (4 instances):

$$\text{Yes} = 4$$

$$\text{No} = 0$$

$$H = 0$$

iii) Rainy (5 instances):

$$\text{Yes} = 3$$

$$\text{No} = 2$$

$$H(D_{\text{rainy}}) = - \left(\frac{3}{5} \log_2 \frac{3}{5} + \frac{2}{5} \log_2 \frac{2}{5} \right) \\ \approx 0.971$$

$$\therefore H(D|\text{outlook}) = \frac{5}{14} \times 0.971 + \frac{4}{14} \times 0 + \frac{5}{14} \times 0.971 \\ \approx 0.694$$

Information gain for outlook,

$$\begin{aligned} IG(D, \text{outlook}) &= H(D) - H(D|\text{outlook}) \\ &= 0.990 - 0.694 \\ &= 0.296 \end{aligned}$$

c) Resulting entropy for each branch after splitting on Outlook:

$$\text{Sunny} \approx 0.971$$

$$\text{Overcast} = 0$$

$$\text{Rainy} \approx 0.971$$

Problem - 02

a) Given, $\hat{y} = w_0 + w_1 x$

Data points = (1, 2), (2, 3), (3, 5)

Initial weights, $w_0 = 0, w_1 = 0$

$\alpha = 0.1$

Mean squared error (MSE) Loss function,

$$L(w_0, w_1) = \frac{1}{m} \sum_{i=1}^m (y_i - \hat{y}_i)^2$$

$$= \frac{1}{3} \sum_{i=1}^3 (y_i - (w_0 + w_1 x))^2$$

b) The gradients of MSE are,

$$\frac{\partial L}{\partial w_0} = -\frac{2}{3} \sum_{i=1}^3 (y - (w_0 + w_1 x))$$

$$\frac{\partial L}{\partial w_1} = -\frac{2}{3} \sum_{i=1}^3 (y - (w_0 + w_1 x)) x_i$$

Now,

For,

$$(1, 2) : 2 - (0 + 0.1) = 2$$

$$(2, 3) : 3 - (0 + 0.2) = 3$$

$$(3, 5) : 5 - (0 + 0.3) = 5$$

Gradients,

$$\frac{\partial L}{\partial w_0} = -\frac{2}{3} (2+3+5) \approx -6.6667$$

$$\begin{aligned}\frac{\partial L}{\partial w_1} &= -\frac{2}{3} ((2 \times 1) + (3 \times 2) + (5 \times 3)) \\ &= -\frac{2}{3} (2+6+15) \\ &\approx -15.33\end{aligned}$$

c) Perform one gradient descent step,

$$w_0^{\text{new}} = w_0 - \alpha \frac{\partial L}{\partial w_0}$$

$$\begin{aligned}&= 0 - 0.1 \times (-6.6667) \\ &= 0.6667\end{aligned}$$

$$w_1^{\text{new}} = w_1 - \alpha \frac{\partial L}{\partial w_1}$$

$$\begin{aligned}&= 0 - 0.1 \times (-15.33) \\ &= 1.533\end{aligned}$$

(Ans)

Problem - 03

a) Given, $\hat{y} = \sigma(\omega_0 + \omega_1 x_1 + \omega_2 x_2)$

$$\sigma(z) = \frac{1}{1 + e^{-z}}, \quad x_1 = 1, \quad x_2 = 2, \quad y = 1$$

$$\omega_0 = 0, \quad \omega_1 = 0.5, \quad \omega_2 = -0.5$$

$$z = \omega_0 + \omega_1 x_1 + \omega_2 x_2 = 0 + 0.5 \times 1 + (-0.5) \times 2 \\ = 0.5 - 1.0 = -0.5$$

$$\hat{y} = \sigma(z) = \frac{1}{1 + e^{-(-0.5)}} \approx 0.3775$$

b) binary cross-entropy loss,

$$L = - [y \log(\hat{y}) + (1-y) \log(1-\hat{y})]$$

$$\therefore L = - [1 \cdot \log(0.3775) + 0 \cdot \log(1 - 0.3775)]$$

$$= - \log(0.3775)$$

$$= 0.9791$$

c) compute the gradient of the loss,

$$\frac{\partial L}{\partial \omega_i} = (\hat{y} - y)x_i$$

i) for ω_0 : $\frac{\partial L}{\partial \omega_0} = \hat{y} - y = 0.3775 - 1$

$$= -0.6225$$

ii) for ω_1 : $\frac{\partial L}{\partial \omega_1} = (\hat{y} - y)x_1 = -0.6225 \times 1$

$$= -0.6225$$

iii) for ω_2 : $\frac{\partial L}{\partial \omega_2} = (\hat{y} - y)x_2 = -0.6225 \times 2$

$$= -1.245$$

(Ans)

d) weight update rule, $\omega_i^{\text{new}} = \omega_i - \alpha \frac{\partial L}{\partial \omega_i}$

update ω_0 ,

$$\omega_0^{\text{new}} = 0 - 0.1 \times (-0.6225)$$

$$= 0.06225$$

update w_1 ;

$$\begin{aligned}w_1^{\text{new}} &= 0.5 - 0.1 \times (-0.6225) \\&= 0.5 + 0.06225 \\&= 0.56225\end{aligned}$$

update w_2 ;

$$\begin{aligned}w_2^{\text{new}} &= -0.5 - 0.1 \times (-1.295) \\&= -0.5 + 0.1295 \\&= -0.3755\end{aligned}$$

updated weights are,

$$w_0 \approx 0.06225$$

$$w_1 \approx 0.56225$$

$$w_2 \approx -0.3755$$

Problem - 01

a) The derivative of binary cross-entropy loss,

$$\frac{\partial L}{\partial a} = -\left(\frac{y}{\hat{y}} - \frac{1-y}{1-\hat{y}}\right)$$

$$= -\frac{1}{0.6} + \frac{0}{0.4}$$

$$\approx -1.6667$$

b) Using chain rule computing the gradient of the loss with respect to w_1 and w_2 ,

we know,

$$\frac{dL}{dw_1} = \frac{dL}{d\hat{y}} \cdot \frac{d\hat{y}}{da} \cdot \frac{da}{dz} \cdot \frac{dz}{dw_1}$$

$$\textcircled{1} \quad \frac{d\hat{y}}{da} = \hat{y}(1-\hat{y}) \cdot w_3 = 0.6(1-0.6) \times 0.4$$
$$= 0.096$$

$$\textcircled{ii} \quad a = \sigma(z) = \sigma(0.3) \Rightarrow \frac{da}{dz} = a(1-a)$$

$$\sigma(0.3) = \frac{1}{1+e^{-0.3}} \approx 0.5744$$

$$\frac{da}{dz} = 0.5744(1 - 0.5744)$$

$$\approx 0.2445$$

$$\frac{dz}{d\omega_1} = x_1 = 1, \quad \frac{dz}{d\omega_2} = x_2 = 2$$

Now,

$$\frac{dL}{d\omega_1} = (-1.6667) \times (0.096) \times (0.2445) \times 1$$

$$\approx -0.0391$$

$$\frac{dL}{d\omega_2} = (-1.6667) \times (0.096) \times (0.2445) \times 2$$

$$\approx -0.0782$$

(Ans)

c)

$$\alpha = 0.1$$

weight update rule,

$$\omega_i^{\text{new}} = \omega_i - \alpha \frac{\partial L}{\partial \omega_i}$$

(i) ω_1 ,

$$\begin{aligned}\omega_1^{\text{new}} &= 0.1 - 0.1 \times (-0.391) \\ &\approx 0.1039\end{aligned}$$

(ii) ω_2 ,

$$\begin{aligned}\omega_2^{\text{new}} &= 0.2 - 0.1 \times (-0.0782) \\ &\approx 0.2078\end{aligned}$$

updated weights,

$$\omega_1 \approx 0.1039$$

$$\omega_2 \approx 0.2078$$

(Ans)