

COMP 9417 - Machine Learning

Tutorial - Classification

Question 2

(C) Email	a?	b?	c?	class
e_1	0	1	0	+
e_2	0	1	1	+
e_3	1	0	0	+
e_4	1	1	0	+
e_5	1	1	0	-
e_6	1	0	1	-
e_7	1	0	0	-
e_8	0	0	0	-

Then the bit vector z is $(2, 3, 1)$ and $(3, 1, 1)$

adding 2 pseudo-documents, one containing each word and one containing none of them.

Then the parameter vector is $\hat{\theta} = (\frac{3}{6}, \frac{4}{6}, \frac{2}{6}) = (0.5, 0.67, 0.33)$

$\hat{\theta}_0 = (\frac{4}{6}, \frac{2}{6}, \frac{2}{6}) = (0.67, 0.33, 0.33)$

For $e^* = abbdabb$

The bit vector is $X = (1, 1, 0)$

The likelihoods is $P(X|\theta) = 0.5 \times 0.67 \times (1-0.33) = 0.222$

$$P(X|\theta) = 0.67 \times 0.33 \times (1-0.33) = 0.148$$

Then the likelihood ratio is $\frac{P(X|\theta)}{P(X|\bar{\theta})} = \frac{0.222}{0.148} = \frac{3}{2} > 1$

$$\text{Because } \frac{P(\theta)}{P(\bar{\theta})} = \frac{\frac{1}{2}}{\frac{1}{2}} = 1$$

$$\text{Then } \frac{P(\theta|X)}{P(\bar{\theta}|X)} = \frac{P(X|\theta)}{P(X|\bar{\theta})} \cdot \frac{P(\theta)}{P(\bar{\theta})} = \frac{3}{2} \times 1 = \frac{3}{2} > 1$$

Therefore, ML classification of x is spam.

Question 3

(a)

Because the sigmoid function is close to (0,1) piecewise and continuously differentiable.

Linear regression does not apply to classification problems because the regression line changes when a new sample is added.

(b)

Because $y|x \sim \text{Bernoulli}(p^*)$, $p^* = \sigma(x^T w^*)$, $\sigma(z) = \frac{1}{1+e^{-z}}$

$$\text{Then } L(w^*) = \prod_{i=1}^n \sigma(x_i^T w^*)^{y_i} [1 - \sigma(x_i^T w^*)]^{1-y_i}$$

$$\begin{aligned}
 \log L(w^*) &= \sum_{i=1}^n x_i \log \sigma(x_i^T w^*) + (1-x_i) \log [1 - \sigma(x_i^T w^*)] \\
 &= n\bar{x} \log [\sigma(x^T w^*)] + n(1-\bar{x}) \log [1 - \sigma(x^T w^*)] \\
 &= n\bar{x} \log \left[\frac{1}{1+e^{-x^T w^*}} \right] + n(1-\bar{x}) \log \left[1 - \frac{1}{1+e^{-x^T w^*}} \right] \\
 &= -n\bar{x} \log (1+e^{-x^T w^*}) - n(1-\bar{x}) [\bar{x}^T w + \log (1+e^{-x^T w^*})]
 \end{aligned}$$

$$\frac{\partial \log L(w^*)}{\partial w^*} = -n\bar{x} \frac{x e^{-x^T w^*}}{1+e^{-x^T w^*}} - n(1-\bar{x}) \left[x + \frac{x e^{-x^T w^*}}{1+e^{-x^T w^*}} \right]$$

$$\text{Let } \frac{\partial \log L(w^*)}{\partial w^*} = 0$$

$$\text{Then } \bar{x} \frac{x e^{-x^T w^*}}{1+e^{-x^T w^*}} = (\bar{x}-1)x + (\bar{x}-1) \frac{x e^{-x^T w^*}}{1+e^{-x^T w^*}}$$

$$\frac{e^{-x^T w^*}}{1+e^{-x^T w^*}} = \bar{x}-1$$

$$\frac{1}{1+e^{-x^T w^*}} - 1 = \bar{x}-1$$

$$e^{-x^T w^*} = \frac{1}{2-\bar{x}} - 1$$

$$\log e^{-x^T w^*} = \log \frac{\bar{x}-1}{2-\bar{x}}$$

$$x^T w^* = \log (2-\bar{x}) - \log (\bar{x}-1)$$

$$w^* = (x^T)^{-1} [\log (2-\bar{x}) - \log (\bar{x}-1)]$$