

# Exercise 1

## 1.1

**Probability** : To find value about some event occur when given distribution. At countable event, probability value is same with likelihood value. But at uncountable event, probability is not same with likelihood.

**Likelihood** : To find appropriate distribution when given fixed value. At countable event, likelihood value is same with probability value. And at uncountable event, likelihood value is same with probability density function value.

## 1.2

**Difference between Sampling and Resampling**: Sampling is the extraction from a population. Sameple is not perfectly same with population. So there are some noise. Resampling is to extract a subset of samples from the sample that I have already. Therefore, the same sample is used several times to measure performance.

**Advantage does Resampling**: Resampling can improve accuracy. Using limited data several times so solve the problem than imbalanced data. Also resampling need fewer assumption. For example, there are bootstrapping, k-fold and so on.

# Exercise 2

## 2.1

At our ppt slide, logistic regression

$$\frac{odds(X+1)}{odds(X)} = e^{\beta}, \quad odds = \frac{P(Y=1)}{1-P(Y=1)}$$

we know the  $\beta$  value -0.0003 so we can calculate

$$\frac{odds(X+1)}{odds(X)} = e^{\beta} = e^{-0.0003} = 0.9997$$

so we can think that

$$Odd(X+1) = 0.9997 \times Odd(X)$$

this mean **Virginica** is more probable than **Versicolor**

So our Interpretation is

∴ **Species Virginica is 1.0003 times more probable than Species Versicolor when the length of sepals Increase**

## 2.2

$$P(Y = 1) = \frac{\exp(X\beta)}{1+\exp(X\beta)}, \quad 1 - P(Y = 1) = \frac{1}{1+\exp(X\beta)}$$

$$\log\left(\frac{P(Y=1)}{1-P(Y=1)}\right) = X\beta = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_p X_p, \quad (\beta_0 = 42.6)$$

$$X_1\beta = \beta_0 + (-0.0003 \times 6.4) + (6.7 \times 3.1) + (-9.4 \times 4.3) + (-18.3 \times 1.3) = -0.84192$$

$$X_2\beta = \beta_0 + (-0.0003 \times 6.9) + (6.7 \times 3.0) + (-9.4 \times 3.9) + (-18.3 \times 1.4) = 0.41793$$

$$P(Y_1 = 1|x_1) = \frac{e^{-0.84192}}{1+e^{-0.84192}} = 0.30113$$

$$P(Y_1 = 0|x_1) = 1 - 0.30113 = 0.69887$$

$$P(Y_2 = 1|x_2) = \frac{e^{0.41793}}{1+e^{0.41793}} = 0.603$$

$$P(Y_2 = 0|x_2) = 1 - 0.6030 = 0.3970$$

Above the result,

∴ **x1 is Virginica and X2 is Versicolor!**

# Exercise 3

## 3.1

$$K(x,z) = (x^T z)^2 = x_1^2 z_1^2 + 2x_1 x_2 z_1 z_2 + x_2^2 z_2^2 \quad x = (x_1, x_2), z = (z_1, z_2)$$

$$K(x,z) = \phi(x)^T \phi(z)$$

So we can know that

$$\phi(x) = \begin{bmatrix} x_1^2 \\ \sqrt{2}x_1 x_2 \\ x_2^2 \end{bmatrix}, \quad \phi(z) = \begin{bmatrix} z_1^2 \\ \sqrt{2}z_1 z_2 \\ z_2^2 \end{bmatrix}$$

## 3.2

We get the formula  $\phi(x)$  at 3.1 so input  $x_1, x_2, x_3$  value.

$$\phi(x) = \begin{bmatrix} x_1^2 \\ \sqrt{2}x_1 x_2 \\ x_2^2 \end{bmatrix}$$

$$x_1 = (1, 2) \quad \phi(x_1) = \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix}$$

$$x_2 = (4, 3) \quad \phi(x_2) = \begin{bmatrix} 16 \\ 12\sqrt{2} \\ 9 \end{bmatrix}$$

$$x_3 = (2, 0) \quad \phi(x_3) = \begin{bmatrix} 4 \\ 0 \\ 0 \end{bmatrix}$$

## 3.3

$$K = \begin{bmatrix} K(x_1, x_1) & K(x_1, x_2) & K(x_1, x_3) \\ K(x_2, x_1) & K(x_2, x_2) & K(x_2, x_3) \\ K(x_3, x_1) & K(x_3, x_2) & K(x_3, x_3) \end{bmatrix}$$

We can calculate by  $K(x_i, x_j) = \phi(x_i)^T \phi(x_j)$ . So,

$$K(x_1, x_1) = 25$$

$$K(x_1, x_2) = K(x_2, x_1) = 100$$

$$K(x_1, x_3) = K(x_3, x_1) = 4$$

$$K(x_2, x_2) = 625$$

$$K(x_2, x_3) = K(x_3, x_2) = 64$$

$$K(x_3, x_3) = 16$$

$$\therefore K = \begin{bmatrix} 25 & 100 & 4 \\ 100 & 625 & 64 \\ 4 & 64 & 16 \end{bmatrix}$$

# Exercise 4

## 4.1

by definition of naive bayes, we can say

$$P(y = A|X1 = a, X2 = b) = \frac{(P(X1=a|y=A) \times P(X2=b|y=A)) \times P(y=A)}{P(X1=a) \times P(X2=b)}$$

$$P(X1 = a|y = A) = \frac{1}{2}, \quad P(X1 = a) = \frac{3}{7}$$

$$P(X2 = b|y = A) = \frac{1}{2}, \quad P(X2 = b) = \frac{4}{7}$$

$$P(y = A) = \frac{4}{7}$$

$$\therefore P(y = A|X1 = a, X2 = b) = \frac{7}{12}$$

## 4.2

To determine whether A or B, I calculate for each case. y = A and y = B

$$P(y = A|X1 = a, X2 = b, x3 = b) = \frac{(P(X1=a|y=A) \times P(X2=b|y=A) \times P(X3=b|y=A)) \times P(y=A)}{P(X1=a) \times P(X2=b) \times P(X3=b)}$$

$$P(X1 = a|y = A) = \frac{1}{2}, \quad P(X1 = a) = \frac{3}{7}$$

$$P(X2 = b|y = A) = \frac{1}{2}, \quad P(X2 = b) = \frac{4}{7}$$

$$P(X3 = b|y = A) = \frac{1}{2}, \quad P(X3 = b) = \frac{3}{7}$$

$$P(y = A) = \frac{4}{7}$$

$$\therefore P(y = A|X1 = a, X2 = b, x3 = b) = \frac{49}{72}$$

$$P(y = B|X1 = a, X2 = b, x3 = b) = \frac{(P(X1=a|y=B) \times P(X2=b|y=B) \times P(X3=b|y=B)) \times P(y=B)}{P(X1=a) \times P(X2=b) \times P(X3=b)}$$

$$P(X1 = a|y = B) = \frac{1}{3}, \quad P(X1 = a) = \frac{3}{7}$$

$$P(X2 = b|y = B) = \frac{2}{3}, \quad P(X2 = b) = \frac{4}{7}$$

$$P(X3 = b|y = B) = \frac{1}{3}, \quad P(X3 = b) = \frac{3}{7}$$

$$P(y = B) = \frac{3}{7}$$

$$\therefore P(y = B|X1 = a, X2 = b, x3 = b) = \frac{49}{162}$$

Because  $P(y=A|X1 = a, X2= b, x3= b)$  is higher than  $P(y=B|X1 = a, X2= b, x3= b)$ ,

∴ **The new data being assigned to Class B.**

# Exercise 5

Because This problem is 10 points on final, I prepared a conceptual question that can be solved quickly.

By using given dataset find the values! -ve point A(-1,1), B(1,3) / +ve C(-4,2).

Exercise 5.1 Find kernel matrix K (2pt)

Exercise 5.2 Find  $\alpha_1, \alpha_2, \alpha_3$  and  $b$  (2pt)

Exercise 5.3 Find  $w$  and hyperplane equation. (3pt)

Exercise 5.4 Prove margin and find margin (3pt)

## 5.1

$$K = \begin{bmatrix} K(A, A) & K(A, B) & K(A, C) \\ K(B, A) & K(B, B) & K(B, C) \\ K(C, A) & K(C, B) & K(C, C) \end{bmatrix}$$

We can calculate by  $K(x_i, x_j) = \phi(x_i)^T \phi(x_j)$ . So,

$$K(A, A) = 2$$

$$K(A, B) = K(B, A) = 2$$

$$K(A, C) = K(C, A) = 6$$

$$K(B, B) = 10$$

$$K(B, C) = K(C, B) = -2$$

$$K(C, C) = 20$$

$$\therefore K = \begin{bmatrix} 2 & 2 & 6 \\ 2 & 10 & -2 \\ -6 & -2 & 20 \end{bmatrix}$$

## 5.2

$$\Sigma \alpha_i y_i = 0$$

$$\Sigma \alpha_i y_i K(x_i, x) + b = 1$$

$$\Sigma \alpha_i y_i K(x_i, x) + b = -1$$

so we can set this!

$$-\alpha_1 - \alpha_2 + \alpha_3 = 0$$

$$-2\alpha_1 - 2\alpha_2 - 6\alpha_3 + b = -1$$

$$-2\alpha_1 - 10\alpha_2 - 2\alpha_3 + b = -1$$

$$6\alpha_1 + 2\alpha_2 + 20\alpha_3 + b = 1$$

solving system of equation

$$\therefore \alpha_1 = \frac{1}{32}, \quad \alpha_2 = \frac{1}{32}, \quad \alpha_3 = \frac{1}{16}, \quad b = -\frac{1}{2}$$

## 5.3

$$W = \Sigma \alpha_i y_i x_i = -\frac{1}{32} \begin{bmatrix} -1 \\ 1 \end{bmatrix} - \frac{1}{32} \begin{bmatrix} 1 \\ 3 \end{bmatrix} - \frac{1}{16} \begin{bmatrix} 4 \\ -2 \end{bmatrix} = \begin{bmatrix} \frac{1}{4} \\ -\frac{1}{4} \end{bmatrix}$$

$$Wx + b = 0$$

$$\therefore \text{hyperplane equation is } x_1 - x_2 - 2 = 0, \quad W = \begin{bmatrix} \frac{1}{4} \\ -\frac{1}{4} \end{bmatrix}$$

## 5.4

we know,  $x^+ = x^- + \lambda w$

$$w^T x^+ + b = w^T (x^- + \lambda w) + b = 1$$

$$w^T x^- + b + \lambda w^T w = 1$$

$$\lambda = \frac{2}{w^T w}$$

we know, Margin =  $\|x^+ - x^-\|_2$

$$= \|x^- + \lambda w + x^-\|_2 = \|\lambda w\|_2$$

$$= \lambda \sqrt{w^T w} = \frac{2}{\sqrt{w^T w}} = \frac{2}{\|w\|_2}$$

$$\therefore \text{Margin} = \frac{2}{\|w\|_2}$$

$$\text{our } W = \begin{bmatrix} \frac{1}{4} \\ -\frac{1}{4} \end{bmatrix}$$

$$\text{so, Margin} = 4\sqrt{2}$$