

(2)

(a)

$$\hat{y}_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} = f(x_i) \text{ where } x_i \text{ is the } i^{\text{th}} \text{ sample } [x_{1i}, x_{2i}]$$

3 samples

$$RSS = \sum_{i=1}^N (y_i - f(x_i))^2 = \sum_{i=1}^N (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i}))^2$$

(b)

$$MSE = \frac{1}{N} \sum_{i=1}^N (y_i - f(x_i))^2$$

$$\frac{\partial MSE}{\partial \beta_j} = \frac{\partial}{\partial \beta_j} \left(\frac{1}{N} \sum_{i=1}^N (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i}))^2 \right)$$

$$= \frac{1}{N} \sum_{i=1}^N \frac{\partial}{\partial \beta_j} (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i}))^2$$

$$= \frac{1}{N} \sum_{i=1}^N 2 (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i})) (-x_{ji})$$

(c)

$$\beta_0 = \beta_0 - 0.5 \cdot \frac{2}{3} \sum_{i=1}^3 (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i})) (-x_{0i})$$

$$= -1 - \frac{1}{3} \cdot ((-1) \cdot (-1) + (-1) \cdot (-1) + (-1) \cdot (-1)) = -2$$

$$\beta_1 = \beta_1 - 0.5 \cdot \frac{2}{3} \sum_{i=1}^3 (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i})) (-x_{1i})$$

$$= 0 - \frac{1}{3} \cdot ((-1) \cdot (-1) + (-1) \cdot 0 + (-1) \cdot 0) = -\frac{1}{3}$$

$$y_1 - (\beta_0 + \beta_1 x_{11} + \beta_2 x_{21}) = -2 - (-1 + 0 + 1 \cdot 0) = -1$$

$$y_2 - (\beta_0 + \beta_1 x_{12} + \beta_2 x_{22}) = 1 - (-1 + 3) = -1$$

$$y_3 - (\beta_0 + \beta_1 x_{13} + \beta_2 x_{23}) = 0 - (-1 + 2) = -1$$

$$\begin{aligned}\beta_2 &= \beta_2 - 0.5 \frac{2}{3} \sum_{i=1}^3 (y_i - (\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i})) (-x_{2i}) \\ &= 1 - \frac{1}{3} ((-1) \cdot 0 + (-1)(-3) + (-1)(-2)) = 1 - \frac{5}{3} = -\frac{2}{3}\end{aligned}$$

Resulting weight vector: $[-2, -1/3, -2/3]^T$

New RSS.

$$y_1 - (\beta_0 + \beta_1 x_{11} + \beta_2 x_{21}) = -2 - (-2 - \frac{1}{3} \cdot 1 - \frac{2}{3} \cdot 0) = \frac{1}{3}$$

$$y_2 - (\beta_0 + \beta_1 x_{12} + \beta_2 x_{22}) = 1 - (-2 - \frac{1}{3} \cdot 0 - \frac{2}{3} \cdot 3) = 5$$

$$y_3 - (\beta_0 + \beta_1 x_{13} + \beta_2 x_{23}) = 0 - (-2 - \frac{1}{3} \cdot 0 - \frac{2}{3} \cdot 2) = \frac{10}{3}$$

$$RSS = (\frac{1}{3})^2 + (5)^2 + (\frac{10}{3})^2 \sim 36.23$$

(3)

$$\hat{\theta}_{MLE} = \underset{\theta}{\operatorname{argmax}} L(\theta) = \underset{\theta}{\operatorname{argmax}} (\log(L(\theta)))$$

$$L(\theta) = \left(\frac{2\theta}{3}\right)^3 \cdot \left(\frac{\theta}{3}\right) \cdot (1-\theta)^2 \quad \text{since } (y_1, y_2, y_3, y_4, y_5, y_6) = (1, 2, 3, 3, 1, 1)$$

$$= \prod_{i=1}^n P(y_i | \theta)$$

$$\log(L(\theta)) = \log\left(\prod_{i=1}^n P(y_i | \theta)\right) = \log\left(\left(\frac{2\theta}{3}\right)^3 \cdot \left(\frac{\theta}{3}\right) \cdot (1-\theta)^2\right)$$

$$= \log\left(\frac{2\theta}{3}\right)^3 + \log\left(\frac{\theta}{3}\right) + \log(1-\theta)^2$$

$$= 3 \log\left(\frac{2\theta}{3}\right) + \log\left(\frac{\theta}{3}\right) + 2 \log(1-\theta)$$

$$= 3 \log\left(\frac{2}{3}\right) + 3 \log \theta + \log\left(\frac{1}{3}\right) + \log \theta + 2 \log(1-\theta)$$

$$= 4 \log \theta + 2 \log(1-\theta) + 3 \log\left(\frac{2}{3}\right) + \log\left(\frac{1}{3}\right)$$

$$\frac{d}{d\theta} \log(L(\theta)) = \frac{4}{\theta} - \frac{2}{1-\theta} = 0$$

$$\frac{4}{\theta} = \frac{2}{1-\theta} \Rightarrow 4(1-\theta) = 2\theta \Rightarrow 4 = 6\theta$$

$$\Rightarrow \hat{\theta}_{MLE} = \frac{2}{3}$$

(4)

$$(a) P(\text{at least one boy gets a book}) = 1 - P(\text{none of the boys get a book})$$

$$= 1 - \frac{20! \cdot 21!}{41!}$$

(b) A: randomly drawn ball is white

B: randomly drawn ball is numbered 2012

2 white
7 black

2012

A and B are independent $\Leftrightarrow P(A|B) = P(A)$

3 white
5 black

2016

$$P(A) = \frac{2+3+5}{24} = \frac{10}{24}$$

5 white
2 black

2020

$$P(B) = \frac{2+7}{24} = \frac{9}{24}$$

24 balls

$$P(A \cap B) = \frac{2}{24}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{\frac{2}{24}}{\frac{9}{24}} = \frac{2}{9}$$

$$P(A|B) = \frac{2}{9} \neq \frac{10}{24} = P(A) \Rightarrow A \text{ and } B \text{ are NOT independent}$$

5

a

Observations

1

2

3

4

5

6

Volatile

1

1

0

0

0

1

Dividend

0

1

0

1

1

0

MACD_CONV

1

0

0

1

0

0

MACD_DIV

0

1

1

0

0

0

MACD_LNR

0

0

0

0

1

1

Risk

High

High

High

Low

Low

High

one-hot-encoded table

$$P(X_1=1 | \text{High}) = \frac{3}{4}$$

$$P(X_1=0 | \text{High}) = \frac{1}{4}$$

$$P(X_2=1 | \text{High}) = \frac{1}{4}$$

$$P(X_2=0 | \text{High}) = \frac{3}{4}$$

$$P(X_3=1 | \text{High}) = \frac{1}{4}$$

$$P(X_3=0 | \text{High}) = \frac{3}{4}$$

$$P(X_4=1 | \text{High}) = \frac{1}{2}$$

$$P(X_4=0 | \text{High}) = \frac{1}{2}$$

$$P(X_5=1 | \text{High}) = \frac{1}{4}$$

$$P(X_5=0 | \text{High}) = \frac{3}{4}$$

$$P(Y=\text{High}) = \frac{2}{3}$$

$$P(X_1=1 | \text{Low}) = 0$$

$$P(X_1=0 | \text{Low}) = 1$$

$$P(X_2=1 | \text{Low}) = 1$$

$$P(X_2=0 | \text{Low}) = 0$$

$$P(X_3=1 | \text{Low}) = \frac{1}{2}$$

$$P(X_3=0 | \text{Low}) = \frac{1}{2}$$

$$P(X_4=1 | \text{Low}) = 0$$

$$P(X_4=0 | \text{Low}) = 1$$

$$P(X_5=1 | \text{Low}) = \frac{1}{2}$$

$$P(X_5=0 | \text{Low}) = \frac{1}{2}$$

$$P(Y=\text{Low}) = \frac{1}{3}$$

Bayes Optimal Classifier $\hat{y} = \underset{y}{\operatorname{argmax}} P(y|x)$

$$= \underset{y}{\operatorname{argmax}} P(x|y) P(y)$$

Naive \Rightarrow Cond. ind. assumption: $P(x_1, \dots, x_d) = P(x_1|Y) \cdot \dots \cdot P(x_d|Y)$

for $i=1, 2, 3, 4, 5$ we calculated $P(X_i | Y=\text{High})$ and $P(X_i | Y=\text{Low})$,
and $P(Y=\text{High})$, $P(Y=\text{Low})$

(b)

for [Yes, No, Linear] $\xrightarrow{\text{by OHE}}$ [1, 0, 0, 0, 1]

$$\hat{y} = \underset{y}{\operatorname{argmax}} P(x_1, x_2, x_3, x_4, x_5 | Y=y), y \in \{\text{High, Low}\}$$

$$\text{For } y = \text{High}, P(x_1=1, x_2=0, x_3=0, x_4=0, x_5=1 | Y=\text{High})$$

$$= P(x_1=1 | Y=\text{High}) P(x_2=0 | Y=\text{High}) P(x_3=0 | Y=\text{High})$$

$$P(x_4=0 | Y=\text{High}) P(x_5=1 | Y=\text{High})$$

$$= \frac{3}{4} \cdot \frac{3}{4} \cdot \frac{3}{4} \cdot \frac{1}{2} \cdot \frac{1}{4} = \frac{27}{512}$$

$$P(Y=\text{High} | x_1, x_2, x_3, x_4, x_5) = \frac{27}{512} \cdot P(Y=\text{High})$$

$$= \frac{27}{512} \cdot \frac{2}{3}$$

$$= \frac{18}{512} = \frac{9}{256}$$

$$\text{For } y = \text{Low}, P(x_1=1, x_2=0, x_3=0, x_4=0, x_5=1 | Y=\text{Low})$$

$$= P(x_1=1 | Y=\text{Low}) P(x_2=0 | Y=\text{Low}) P(x_3=0 | Y=\text{Low}) P(x_4=0 | Y=\text{Low})$$

$$P(x_5=1 | Y=\text{Low})$$

$$= 0 \cdot 0 \cdot \frac{1}{2} \cdot 1 \cdot \frac{1}{2} = 0$$

$$P(Y=\text{Low} | x_1, x_2, x_3, x_4, x_5) = 0 \cdot \frac{1}{3}$$

$$= \frac{0}{256} = 0$$

Therefore, $\hat{y} = \text{High}$