

# Chapter 2

Xuan

2020 年 9 月 14 日

## 1 条件互信息

在有三个变量的情况下，符号  $x_i$  与符号对  $(y_j, z_k)$  之间的互信息量定义为

$$I(x_i; y_j, z_k) = \log \frac{p(x_i | y_j, z_k)}{p(x_i)}$$

条件互信息量是在给定  $z_k$  条件下， $x_i$  与  $y_j$  之间的互信息量，定义为

$$I(x_i; y_j | z_k) = \log \frac{p(x_i | y_j, z_k)}{p(x_i | z_k)}$$

结合上述公式可写为

$$I(x_i; y_j, z_k) = I(x_i; z_k) + I(x_i; y_j | z_k)$$

### 1.1 相关公式推导

$$\begin{aligned} I(X; Y | Z) &= \sum p(x, y, z) \log \frac{p(x, y | z)}{p(x | z)p(y | z)} \\ &= \sum_z D(p(x, y | z) || p(x | z)p(y | z)) * p(z) \end{aligned} \tag{1}$$

$$\begin{aligned} \frac{p(x, y | z)}{p(x | z)p(y | z)} &= \frac{p(x | y, z)p(y | z)p(x)}{p(x)p(x | z)p(y | z)} \\ &= \frac{p(x, y, z)}{p(x)p(y, z)} * \frac{p(x)}{p(x | z)} \end{aligned} \tag{2}$$

将 (2) 带入 (1)

$$(1) = \sum p(x, y, z) \log \frac{p(x, y, z)}{p(x)p(y, z)} - \sum p(x, y, z) \log \frac{p(x|z)p(z)}{p(x)p(z)} \quad (3)$$

$$\begin{aligned} \sum p(x, y, z) \log \frac{p(x|z)p(z)}{p(x)p(z)} &= \sum_{x, y, z} (p(x, z) \log \frac{p(x, z)}{p(x)p(z)}) p(y|x, z) \\ &= \sum_{x, z} \sum_y \\ &= \sum_{x, z} p(x, z) \log \frac{p(x, z)}{p(x)p(z)} \sum_y p(y|x, z) \end{aligned} \quad (4)$$

将 (4) 带入 (3)

$$\begin{aligned} (3) &= \sum p(x, y, z) \log \frac{p(x, y, z)}{p(x)p(y, z)} - \sum p(x, z) \log \frac{p(x, z)}{p(x)p(z)} \\ &= I(X; Y, Z) - I(X; Z) \\ &= H(X) - H(X|Y, Z) - (H(X) - H(X|Z)) \\ &= H(X|Z) - H(X|Y, Z) \end{aligned} \quad (5)$$

## 2 信息不增性

对于马尔可夫链  $X \rightarrow Y \rightarrow Z$ ，有

$$I(X; Z) \leq I(Y; Z)$$

### 2.1 公式证明

$$\left. \begin{aligned} I(X; Y, Z) &= I(X; Y) + I(X; Z|Y) \\ I(X; Z|Y) &= \sum_y D(p(x, z|y) || p(x|y)p(z|y)) p(y) = 0 \end{aligned} \right\} I(X; Y, Z) = I(X; Y) \quad (6)$$

$$I(X; Y, Z) = I(X; Z) + I(X; Y|Z) \quad (7)$$

$$p(x, y, z) = p(y)p(x|y)p(z|y) \iff p(x, z|y) = p(x|y)p(z|y)$$

综上所述， $I(X; Y) = I(X; Y, Z) \geq I(X; Z)$

## 2.2 文氏图证明

见 chapter2.jpg

## 3 连续信源的熵和互信息

### 3.1 连续信源熵

$$H_c(x) = - \int_{-\infty}^{\infty} p_x(x) \log p_x(x) dx \geq 0$$

### 3.2 能量确定条件下, $\int_{-\infty}^{\infty} x^2 p(x) dx$ 在高斯分布情况下最大化熵率

$$H_c(x) = - \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(x-m)^2}{2\sigma^2}} \log \left[ \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(x-m)^2}{2\sigma^2}} \right] dx \quad (8)$$

### 3.3 范围确定条件下, 均匀分布最大化熵率

变量  $X$  的幅度取值限定在  $[a, b]$ , 则有  $\int_a^b p_x(x) dx = 1$ , 当任意  $p_x(x)$  符合平均条件,

$$p_x(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0, & \text{other} \end{cases}$$

时, 信源达到最大熵

$$\begin{aligned} H_c(x) &= - \int_a^b \frac{1}{b-a} \log \frac{1}{b-a} dx \\ &= \log(b-a) \end{aligned} \quad (9)$$