

CSE250A HW1

$$\begin{aligned} 1.1 (a) \quad P(X, Y|E) &= \frac{P(X, Y, E)}{P(E)} && \text{Definition of conditional probability} \\ &= \frac{P(X|Y, E) \cdot P(Y|E) \cdot P(E)}{P(E)} && \text{Product rule expansion} \\ &= P(X|Y, E) \cdot P(Y|E) \end{aligned}$$

$$\begin{aligned} (b) \quad P(X|Y, E) &= \frac{P(X, Y, E)}{P(Y, E)} && \text{Definition of conditional probability} \\ &= \frac{P(E) \cdot P(X|E) \cdot P(Y|X, E)}{P(E) \cdot P(Y|E)} && \text{Product rule expansion} \\ &= \frac{P(Y|X, E) \cdot P(X|E)}{P(Y|E)} \end{aligned}$$

$$\begin{aligned} (c) \quad \text{From definition of conditional probability: } P(X|E) &= \frac{P(X, E)}{P(E)} \\ P(X, Y=y|E) &= \frac{P(X, Y=y, E)}{P(E)} \end{aligned}$$

$$\begin{aligned} \therefore P(X, E) &= \sum_y P(X, Y=y, E) \\ \therefore \frac{P(X, E)}{P(E)} &= \frac{\sum_y P(X, Y=y, E)}{P(E)} \end{aligned}$$

$$\therefore P(X|E) = \sum_y P(X, Y=y|E)$$

1.2 (1) (2) (3) can be expanded based on the definition of conditional probability.

$$(1) \quad \frac{P(X, Y, E)}{P(E)} = \frac{P(X, E)}{P(E)} \cdot \frac{P(Y, E)}{P(E)} \Rightarrow P(X, Y, E) = \frac{P(X, E) \cdot P(Y, E)}{P(E)}$$

$$(2) \quad \frac{P(X, Y, E)}{P(Y, E)} = \frac{P(X, E)}{P(E)} \Rightarrow P(X, Y, E) = \frac{P(X, E) \cdot P(Y, E)}{P(E)}$$

$$(3) \quad \frac{P(X, Y, E)}{P(X, E)} = \frac{P(Y, E)}{P(E)} \Rightarrow P(X, Y, E) = \frac{P(X, E) \cdot P(Y, E)}{P(E)}$$

\therefore (1) (2) (3) are equivalent

1.3 (a) X: There is a fire in the forest.

Y: The weather is hot and dry.

Z: Someone was smoking in the forest just now.

(b) X: There is a fire in the forest.

Y: Someone was smoking in the forest just now.

Z: It is raining now.

(c) X: There is a fire in the forest.

Y: Someone was smoking in the forest just now.

Z: Someone is burning the forest now.

1.4 Given: $P(D=1) = 1\%$ $P(T=1 | D=0) = 5\%$ $P(T=0 | D=1) = 10\%$

$$(a) P(D=0 | T=0) = \frac{P(T=0 | D=0) \cdot P(D=0)}{P(T=0)} \quad \text{Bayes rule}$$

$$= \frac{P(T=0 | D=0) \cdot P(D=0)}{P(D=0, T=0) + P(D=1, T=0)} \quad \text{Marginalization}$$

$$= \frac{P(T=0 | D=0) \cdot P(D=0)}{P(D=0) \cdot P(T=0 | D=0) + P(D=1) \cdot P(T=0 | D=1)} \quad \text{Product rule}$$

$$= \frac{(1-0.05) \times (1-0.01)}{(1-0.01) \times (1-0.05) + 0.01 \times 0.10}$$

$$= 99.89\%$$

$$(b) P(D=1 | T=1) = \frac{P(T=1 | D=1) \cdot P(D=1)}{P(T=1)} \quad \text{Bayes rule}$$

$$= \frac{P(T=1 | D=1) \cdot P(D=1)}{1 - P(T=0)}$$

$$= \frac{(1-0.10) \times 0.01}{1 - [(1-0.01) \times (1-0.05) + 0.01 \times 0.10]}$$

$$= 15.38\%$$

1.5 (a) $f(x) = \log(x) - (x-1)$

$$f'(x) = \frac{1}{x} - 1$$

$$f'(x) = \begin{cases} > 0 & 0 < x < 1 \\ = 0 & x = 1 \\ < 0 & x > 1 \end{cases}$$

$\therefore f(x)$ is maximum at $x=1$

$$f(x=1) = \log(1) - 1 + 1 = 0$$

$$\therefore f(x) = \log(x) - (x-1) \leq 0$$

$$\therefore \log(x) \leq x-1$$

See the plot in later page.

(b) $\therefore \log(x) \leq x-1 \quad \therefore -\log(x) \geq 1-x$

$$\therefore KL(p, q) = \sum_i p_i \log\left(\frac{p_i}{q_i}\right)$$

$$= -\sum_i p_i \log\left(\frac{q_i}{p_i}\right)$$

$$\geq \sum_i p_i \left(1 - \frac{q_i}{p_i}\right) = \sum_i (p_i - q_i) = \sum_i p_i - \sum_i q_i = 0$$

$$\therefore KL(p, q) \geq 0$$

(c) $KL(p, q) = \sum_i p_i \log\left(\frac{p_i}{q_i}\right)$

$$= -\sum_i p_i \log\left(\frac{q_i}{p_i}\right)$$

$$= -2 \sum_i p_i \log\left(\sqrt{\frac{q_i}{p_i}}\right)$$

$$\geq -2 \sum_i p_i \left(\sqrt{\frac{q_i}{p_i}} - 1\right) = -2 \sum_i (\sqrt{p_i q_i} - p_i)$$

$$= 2 \sum_i p_i - 2 \sum_i \sqrt{p_i q_i}$$

$$= \sum_i p_i - 2 \sum_i \sqrt{p_i q_i} + \sum_i q_i \quad \Leftarrow \sum_i p_i = \sum_i q_i = 1$$

$$= \sum_i (p_i - 2\sqrt{p_i q_i} + q_i)$$

$$= \sum_i (\sqrt{p_i} - \sqrt{q_i})^2$$

$$\therefore KL(p, q) \geq \sum_i (\sqrt{p_i} - \sqrt{q_i})^2$$

(d) Assume $p = [0.5 \ 0.5]$ and $q = [0.9 \ 0.1]$ satisfies $\sum_i p_i = \sum_i q_i = 1$

$$KL(p, q) = 0.5 \log\left(\frac{0.5}{0.9}\right) + 0.5 \log\left(\frac{0.5}{0.1}\right) = 0.5108$$

$$KL(q, p) = 0.9 \log\left(\frac{0.9}{0.5}\right) + 0.1 \log\left(\frac{0.1}{0.5}\right) = 0.3681$$

$$\therefore KL(p, q) \neq KL(q, p)$$

$$\begin{aligned}
 1.7 \text{ (a)} \quad I(X, Y) &= \sum_x \sum_y P(x, y) \log \left[\frac{P(x, y)}{P(x)P(y)} \right] \\
 &= - \sum_x \sum_y P(x, y) \log \left[\frac{P(x)P(y)}{P(x, y)} \right] \\
 &\geq \sum_x \sum_y P(x, y) \left[1 - \frac{P(x)P(y)}{P(x, y)} \right] = \sum_x \sum_y [P(x, y) - P(x)P(y)] \\
 &= \sum_x \sum_y P(x, y) - \sum_x P(x) \cdot \sum_y P(y) \\
 &= 1 - 1 \\
 &= 0
 \end{aligned}$$

$$\therefore I(X, Y) \geq 0$$

$$(b) \text{ If } X \text{ and } Y \text{ are independent} \Rightarrow P(x, y) = P(x) \cdot P(y)$$

$$\therefore I(X, Y) = \sum_x \sum_y P(x, y) \cdot \log(1) = 0$$

$$\text{If } I(X, Y) = 0 \Rightarrow P(x, y) \neq 0 \text{ and } \log \left[\frac{P(x, y)}{P(x) \cdot P(y)} \right] = 0$$

$$\therefore P(x, y) = P(x) \cdot P(y) \quad X \text{ and } Y \text{ are independent}$$