

6.1)

CSE 250A - Homework 6

$$a) P(a, b | c, d) = \frac{P(a, b, c, d)}{P(c, d)} \quad \text{Baye's Rule}$$

$$P(a, b, c, d) =$$

$$\underbrace{P(d | a, b, c)}_{\text{causal chain}} \cdot P(c | a, b) \cdot P(b | a) \cdot P(a) = \text{Chain Rule}$$

$$P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)$$

$$P(c, d) = \sum_a \sum_b P(a, b, c, d) =$$

$$\sum_a \sum_b \underbrace{P(d | a, b, c)}_{\text{causal chain}} \cdot P(c | a, b) \cdot P(b | a) \cdot P(a) = \text{Chain Rule}$$

$$\sum_a \sum_b P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)$$

$$P(a, b | c, d) = \frac{P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)}{\sum_a \sum_b P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)}$$

$$b) P(a | c, d) = \sum_b P(a, b | c, d) =$$

$$\frac{\sum_b P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)}{\sum_a \sum_b P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)}$$

$$P(b | c, d) = \sum_a P(a, b | c, d) =$$

$$\frac{\sum_a P(d | b, c) \cdot P(c | a, b) \cdot P(b | a) \cdot P(a)}{\sum_{a'} \sum_b P(d | b, c) \cdot P(c | a', b) \cdot P(b | a') \cdot P(a')}$$

$$c) P(C=c_t, D=d_t) =$$

$$\sum_a \sum_b P(a, b, C=c_t, D=d_t) = \text{From part a}$$

$$\sum_a \sum_b P(D=d_t | b, C=c_t) \cdot P(C=c_t | a, b) \cdot P(b | a) \cdot P(a)$$

$$\downarrow$$

$$\mathcal{L} = \sum_t \log(P(C=c_t, D=d_t))$$

$$\mathcal{L} = \sum_t \log(\sum_a \sum_b P(D=d_t | b, C=c_t) \cdot P(C=c_t | a, b) \cdot P(b | a) \cdot P(a))$$

$$\frac{P(Y=y, X=x, Z_i=1)}{P(X=x, Z_i=1)} \cdot \frac{P(X_i=1, Z_i=1)}{P(Z_i=1)} =$$

$$\frac{P(Y=y, X=x)}{P(X=x)}$$

$$\frac{P(Y=y, X=x, Z_i=1) \cdot P(X_i=1, Z_i=1) \cdot P(X=x)}{P(X=x, Z_i=1) \cdot P(Y=y, X=x) \cdot P(Z_i=1)} =$$

6.2)

$$a) P(Y=1|X) = \sum_{Z \in \{0,1\}^n} P(Y=1, Z|X)$$

$$\text{Compute } P(Y=1, Z|X) = P(Y=1|Z) \cdot P(Z|X)$$

Note that

$$P(Y=1|Z) = \begin{cases} 1 & \text{if } Z_i = 1 \text{ for any } i \\ 0 & \text{if } Z_i = 0 \text{ for all } i \end{cases}$$

$$P(Y=1|Z) \cdot P(Z|X) = \prod_{i=1}^n P(Z_i|X_i)$$

Substitute this back into $P(Y=1|X)$

$$P(Y=1|X) =$$

$$\sum_{Z \in \{0,1\}^n} P(Y=1, Z|X) =$$

$$\sum_{Z \in \{0,1\}^n} \left(\prod_{i=1}^n P(Z_i|X_i) \right) =$$

Note that

$$P(Z_i=1|X_i=0) = 0$$

$$P(Z_i=1|X_i=1) = p_i$$

This also implies that

$$P(Z_i=0|X_i=0) = 1$$

$$P(Z_i=1|X_i=1) = 1 - p_i$$

$$\sum_{Z \in \{0,1\}^n} \prod_{i=1}^n P(Z_i=1|X_i)^{X_i} \cdot P(Z_i=0|X_i)^{1-X_i} = \sum_{Z \in \{0,1\}^n} \prod_{i=1}^n (p_i^{X_i} \cdot (1-p_i)^{1-X_i})$$

Note: for the cases where Z has at least one $Z_i=1$, the product term for each i such that $Z_i=0$ becomes 0, since $(1-p_i)^{1-X_i}$ is 0 when $X_i=0$.

Thus, for the sum, we can simplify it to

only consider cases where Z has at least one $z_i = 1$.

$$P(Y=1|X) = \sum_{\substack{z \in \{0,1\}^n \\ \text{at least one } z_i = 1}} \prod_{i=1}^n \left((p_i^{x_i}) \cdot (1-p_i)^{1-x_i} \right) =$$

This is effectively a noisy-OR implementation.

$$P(Y=1|X) = 1 - \prod_{i=1}^n (1-p_i^{x_i})$$

b) $P(Z_i=1, X_i=1 | X=x, Y=y) =$ Bayes' rule

$$\frac{P(X=x, Y=y | Z_i=1, X_i=1) \cdot P(Z_i=1 | X_i=1)}{P(X=x, Y=y)}$$

compute $P(X=x, Y=y | Z_i=1, X_i=1)$

Recall from the extended belief network that Z_i is the intermediary between X_i and Y . Also recall that $P(Z_i=1 | X_i=1) = p_i$.

Thus, $Z_i=1$ implies that $X_i=1$, so we can simplify as follows:

$$P(X=x, Y=y | Z_i=1, X_i=1) =$$

$$P(X=x, Y=y | X_i=1) =$$

$$P(Y=y | X=x) \leftarrow \text{noisy-OR}$$

For $P(Z_i=1 | X_i=1)$, we are given that $P(Z_i=1 | X_i=1) = p_i$

Regarding the denominator (next page),

$$P(X=x, Y=y) = \sum_{z \in \{0,1\}^n} P(X=x, Y=y | z) \cdot P(z)$$

Recall from the extended belief network's structure and from part a that

$$P(X=x, Y=y | z) = \prod_{i=1}^n P(z_i | x_i)$$

and

$$P(z) = \prod_{i=1}^n (p_i^{x_i} \cdot (1-p_i)^{1-x_i})$$

Thus,

$$\begin{aligned} P(X=x, Y=y) &= \sum_{z \in \{0,1\}^n} P(X=x, Y=y | z) \cdot P(z) = \\ &= \sum_{z \in \{0,1\}^n} \prod_{i=1}^n P(z_i | x_i) \cdot \prod_{i=1}^n (p_i^{x_i} \cdot (1-p_i)^{1-x_i}) = \text{from part a} \\ P(Y=1 | X) &= 1 - \prod_{i=1}^n (1-p_i^{x_i}) \end{aligned}$$

Putting these all together, we get

$$P(Z_i=1, X_i=1 | X=x, Y=y) =$$

$$\frac{P(X=x, Y=y | Z_i=1, X_i=1) \cdot P(Z_i=1 | X_i=1)}{P(X=x, Y=y)} =$$

$$\frac{P(Y=y | X=x) \cdot p_i}{1 - \prod_{i=1}^n (1-p_i^{x_i})}$$

Regarding $P(Y=y | X=x)$, note that if $X_i=1$, it implies that $Y=1$ and $X_i=1$ based on the definitions given in this problem. Thus, the event that $X=x$ and $Y=y$ can only occur when $y \cdot x_i = 1$.

This shows that

$$P(Z_i=1, X_i=1 | X=x, Y=y) = \frac{y \cdot x_i \cdot p_i}{1 - \prod_{i=1}^n (1 - p_i)^{x_i}}$$

- c) From part a, recall that $Z_i=1$ implies that $X_i=1$. From the problem statement, $P(Z_i=1 | X_i=1) = p_i$, but since $Z_i=1$ implies that $X_i=1$, we can simply compute that $P(X_i=1) = p_i$.

The special case of the general formula from lecture is

$$P(X_i=x) = \frac{1}{T_i} \sum_t P(X_i=x | V_t=v_t)$$

Our data set is $\{\vec{x}^{(t)}, \vec{y}^{(t)}\}_{t=1}^T$.

Thus, we have

$$p_i \leftarrow \frac{1}{T_i} \sum_t P(X_i=x | X=x^{(t)}, Y=y^{(t)})$$

6.3)

$$\begin{aligned} a) f'(x) &= \frac{d}{dx} \log(\cosh(x)) = \\ &= \frac{1}{\cosh(x)} \cdot \frac{d}{dx} (\cosh(x)) = \\ &= \frac{\sinh(x)}{\cosh(x)} \end{aligned}$$

Find critical points of $f'(x)$

$$\frac{\sinh(x)}{\cosh(x)} = 0 \text{ when } x=0 \text{ because}$$

$$\left. \frac{\sinh(x)}{\cosh(x)} \right|_{x=0} =$$

$$\frac{\sinh(0)}{\cosh(0)} =$$

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

$$f''(x) = \frac{d}{dx} \left(\frac{\sinh(x)}{\cosh(x)} \right) =$$

$$\frac{\cosh(x) \cdot \cosh(x) - \sinh(x) \cdot \sinh(x)}{\cosh^2(x)} =$$

$$\frac{\cosh^2(x) - \sinh^2(x)}{\cosh^2(x)} =$$

$$\frac{1}{\cosh^2(x)}$$

This implies that $f''(x) > 0$ for all values of x .

By the second derivative $f''(x) > 0$ and $f'(x) = 0$ when $x=0$, this implies that $x=0$ is the minimum.

b) $f''(x) = \frac{1}{\cosh^2(x)}$ from part a.

Note that $\cosh(x) \geq 1$ for all x . This means that $\cosh^2(x) \geq 1$ for all x too.

Because $\cosh^2(x)$ is in the denominator, this means that $f''(x) \leq 1$ for all x . Thus, $f''(x) \leq 1$ for all x .

c) $Q(x, y) = \log(\cosh(y)) + \frac{\sinh(y)}{\cosh(y)} \cdot (x-y) + \frac{1}{2} \cdot (x-y)^2$

d)

i) $Q(x, x) = f(x)$

$$\begin{aligned} Q(x, x) &= f(x) + f'(x) \cdot (x - x) + \frac{1}{2} \cdot (x - x)^2 = \\ &= f(x) + f'(x) \cdot 0 + \frac{1}{2} \cdot 0^2 = \\ &= f(x) + 0 + 0 = \\ &= f(x) \end{aligned}$$

Thus, $Q(x, x) = f(x)$.

ii) $Q(x, y) \geq f(x)$

First, let's solve parts of the integral:

$$f(x) = f(y) + \int_y^x du \cdot [f'(y) + \int_y^u dv \cdot f''(v)] =$$

$$f(y) + \int_y^x du \cdot f'(y) + \int_y^x \int_y^u dv \cdot f''(v) =$$

$$f(y) + [f'(y) \cdot u]_{u=y}^{u=x} + \int_y^x \int_y^u f''(v) \cdot dv =$$

$$f(y) + (x - y) \cdot f'(y) + \int_y^x \int_y^u f''(v) \cdot dv$$

We'll come back to this in a bit.

Recall that

$$Q(x, y) = \underbrace{f(y) + f'(y) \cdot (x - y)} + \frac{1}{2} \cdot (x - y)^2$$

Notice that this is equal to the first 2 parts of the solution for $f(x)$

$$\begin{aligned} f(x) &= f(y) + f'(y) \cdot (x - y) + \int_y^x \int_y^u f''(v) \cdot dv \rightarrow \\ f(x) - \int_y^x \int_y^u f''(v) \cdot dv &= f(y) + f'(y) \cdot (x - y) \end{aligned}$$

Consider $Q(x, y) \geq f(x)$. Rewrite as
 $Q(x, y) - f(x) \geq 0$

Solve for $Q(x, y) - f(x)$ on the next page.

$$\begin{aligned}
 Q(x, y) - f(x) &= \\
 f(y) + f'(y) \cdot (x - y) + \frac{1}{2} \cdot (x - y)^2 - f(x) &= \\
 f(x) - \int_y^x \int_y^u f''(v) \cdot dv + \frac{1}{2} \cdot (x - y)^2 - f(x) &= \\
 \frac{1}{2} \cdot (x - y)^2 - \underbrace{\int_y^x \int_y^u f''(v) \cdot dv}
 \end{aligned}$$

Let's analyze this.

From part b, we know that $f''(x) \leq 1$ for all x . This implies that

$$\int_y^u f''(v) \cdot dv \leq \int_y^u 1 \cdot dv = u - y \rightarrow$$

$$\begin{aligned}
 \int_y^x \int_y^u f''(v) \cdot dv &\leq \int_y^x \int_y^u 1 \cdot dv = \\
 \int_y^x (u - y) \cdot du &= \\
 \left. \frac{1}{2} \cdot u^2 - u \cdot y \right|_{u=y}^{u=x} &= \\
 \frac{1}{2} \cdot x^2 - x \cdot y - \left(\frac{1}{2} \cdot y^2 - y^2 \right) &= \\
 \frac{1}{2} \cdot x^2 - x \cdot y - \left(-\frac{1}{2} \cdot y^2 \right) &= \\
 \frac{1}{2} \cdot x^2 - x \cdot y + \frac{1}{2} \cdot y^2 &= \\
 \frac{1}{2} \cdot (x - y)^2
 \end{aligned}$$

So, plugging this back into $Q(x, y) - f(x)$

$$\begin{aligned}
 Q(x, y) - f(x) &= \quad \quad \quad \leftarrow \text{Because this is } \leq \frac{1}{2} \cdot (x - y)^2 \\
 \frac{1}{2} \cdot (x - y)^2 - \int_y^x \int_y^u f''(v) \cdot dv &\geq \\
 \frac{1}{2} \cdot (x - y)^2 - \frac{1}{2} \cdot (x - y)^2 &= \\
 0
 \end{aligned}$$

So, this implies that $Q(x, y) - f(x) \geq 0$ and that

$$Q(x, y) \geq f(x)$$