

# Homework 3

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## Problem 1

### Lemma

Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space. Let  $A, B, C \in \mathcal{F}$  be arbitrary events. Then,  $\mathbb{P}(A|B \cap C) = \mathbb{P}(A|C)$  iff  $A \perp\!\!\!\perp B|C$ .

### Proof

$$\mathbb{P}(A|B \cap C) = \frac{\mathbb{P}(A \cap B \cap C)}{\mathbb{P}(B \cap C)} = \frac{\mathbb{P}(A \cap B|C) \mathbb{P}(C)}{\mathbb{P}(B \cap C)} \stackrel{A \perp\!\!\!\perp B|C}{=} \frac{\mathbb{P}(A|C) \mathbb{P}(B|C) \mathbb{P}(C)}{\mathbb{P}(B \cap C)} = \mathbb{P}(A|C). \quad \blacksquare$$

1. All paths from  $\mathbf{x}_i$  to  $\mathbf{x}_n$  are blocked by  $\mathbf{z}_n$  - case **a** - for  $i < n \Rightarrow \mathbf{x}_1, \dots, \mathbf{x}_{n-1} \perp\!\!\!\perp \mathbf{x}_n | \mathbf{z}_n$ .
2. All paths from  $\mathbf{x}_i$  to  $\mathbf{z}_n$  are blocked by  $\mathbf{z}_{n-1}$  - case **a** - for  $i < n \Rightarrow \mathbf{x}_1, \dots, \mathbf{x}_{n-1} \perp\!\!\!\perp \mathbf{z}_n | \mathbf{z}_{n-1}$ .

$$\begin{aligned} p(\mathbf{x}_{n+1}, \dots, \mathbf{x}_N, \mathbf{Z}) &= \sum_{x_1} \cdots \sum_{x_n} p(\mathbf{X}, \mathbf{Z}) = p(\mathbf{z}_1) \prod_{k=2}^N p(\mathbf{z}_k | \mathbf{z}_{k-1}) \sum_{x_1} \cdots \sum_{x_n} \prod_{m=1}^N p(\mathbf{x}_m | \mathbf{z}_m) \\ &= p(\mathbf{z}_1) \prod_{k=2}^N p(\mathbf{z}_k | \mathbf{z}_{k-1}) \prod_{m=n+1}^N p(\mathbf{x}_m | \mathbf{z}_m). \end{aligned}$$

$$\begin{aligned} \frac{p(\mathbf{x}_{n+1}, \dots, \mathbf{x}_N, \mathbf{z}_n, \mathbf{z}_{n+1})}{p(\mathbf{z}_{n+1})} &= \frac{1}{p(\mathbf{z}_{n+1})} \sum_{z_1} \cdots \sum_{z_{n-1}} \sum_{z_{n+2}} \cdots \sum_{z_N} p(\mathbf{z}_1) \prod_{k=2}^N p(\mathbf{z}_k | \mathbf{z}_{k-1}) \prod_{m=n+1}^N p(\mathbf{x}_m | \mathbf{z}_m) \\ &= \frac{1}{p(\mathbf{z}_{n+1})} p(\mathbf{z}_n) p(\mathbf{z}_{n+1} | \mathbf{z}_n) p(\mathbf{x}_{n+1} | \mathbf{z}_{n+1}) p(\mathbf{x}_{n+2}, \dots, \mathbf{x}_N | \mathbf{z}_{n+1}) \\ &= p(\mathbf{z}_n | \mathbf{z}_{n+1}) p(\mathbf{x}_{n+1}, \dots, \mathbf{x}_N | \mathbf{z}_{n+1}). \end{aligned}$$

$$4. \quad p(\mathbf{z}_{N+1}, \mathbf{X} | \mathbf{z}_N) = \frac{p(\mathbf{z}_{N+1}, \mathbf{X}, \mathbf{z}_N)}{p(\mathbf{z}_N)} = \frac{\sum_{z_1} \cdots \sum_{z_{N-1}} p(\mathbf{X}, \mathbf{Z})}{p(\mathbf{z}_N)} p(\mathbf{z}_{N+1} | \mathbf{z}_N) = p(\mathbf{X} | \mathbf{z}_N) p(\mathbf{z}_{N+1} | \mathbf{z}_N).$$

## Problem 2

Note that the maximal cliques are of the form  $\{x_i, x_{i+1}\}$ . Thus, we can transform the given MRF into a FG where the corresponding potential function for the mentioned clique is  $\psi_{i,i+1}(x_i, x_{i+1})$ .

Thus, let  $x_n$  be the root and start from the leaves:

$$\begin{aligned} \mu_{x_1 \rightarrow \psi_{1,2}}(x_1) &= 1 & \mu_{x_N \rightarrow \psi_{N-1,N}}(x_N) &= 1 \\ \mu_{\psi_{1,2} \rightarrow x_2}(x_2) &= \sum_{x_1} \psi_{1,2}(x_1, x_2) \cdot 1 & \mu_{\psi_{N-1,N} \rightarrow x_{N-1}}(x_{N-1}) &= \sum_{x_N} \psi_{N-1,N}(x_{N-1}, x_N) \cdot 1 \\ \mu_{x_2 \rightarrow \psi_{2,3}}(x_2) &= \mu_{\psi_{1,2} \rightarrow x_2}(x_2) & \mu_{x_{N-1} \rightarrow \psi_{N-2,N-1}}(x_{N-1}) &= \mu_{\psi_{N-1,N} \rightarrow x_{N-1}}(x_{N-1}) \\ &\vdots & &\vdots \end{aligned}$$

Define,

$$\begin{aligned} \mu_\alpha(x_n) &:= \mu_{\psi_{n-1,n} \rightarrow x_n}(x_n) = \sum_{x_{n-1}} \psi_{n-1,n}(x_{n-1}, x_n) \cdot \mu_{x_{n-1} \rightarrow \psi_{n-1,n}}(x_{n-1}) \\ \mu_\beta(x_n) &:= \mu_{\psi_{n+1,n} \rightarrow x_n}(x_n) = \sum_{x_{n+1}} \psi_{n+1,n}(x_n, x_{n+1}) \cdot \mu_{x_{n+1} \rightarrow \psi_{n+1,n}}(x_{n+1}) \end{aligned}$$

Therefore,  $\tilde{p}(x_n) = \mu_\alpha(x_n) \mu_\beta(x_n) \Rightarrow p(x_n) = \frac{1}{Z} \mu_\alpha(x_n) \mu_\beta(x_n)$ .

### Problem 3

If we observe  $\mathbf{x}_N$  to have a value  $\hat{\mathbf{x}}_N$ , this has the same effect as multiplying the joint distribution by  $\mathbb{1}[\mathbf{x}_N = \hat{\mathbf{x}}_N]$ . Since  $\tilde{p}(\mathbf{x})$  is a product over the potential functions, we can collect this factor into the factor  $\psi_{N-1,N}$ .

Summations over  $\mathbf{x}_N$  then contain only one term in which  $\mathbf{x}_N = \hat{\mathbf{x}}_N$ . Note that for calculating  $p(\mathbf{x}_n|\mathbf{x}_N)$  we can apply the same message passing algorithm and renormalize by  $p(\mathbf{x}_N = \hat{\mathbf{x}}_N)$ .

### Problem 4

$$\begin{aligned}
 \tilde{p}(x_1) &= \mu_{f_a \rightarrow x_1}(x_1) = \sum_{x_2} f_a(x_1, x_2) \mu_{x_2 \rightarrow f_a}(x_2) & \tilde{p}(x_3) &= \mu_{f_b \rightarrow x_3}(x_3) = \sum_{x_2} f_b(x_2, x_3) \mu_{x_2 \rightarrow f_b}(x_2) \\
 &= \sum_{x_2} f_a(x_1, x_2) \mu_{f_b \rightarrow x_2}(x_2) \mu_{f_c \rightarrow x_2}(x_2) & &= \sum_{x_2} f_b(x_2, x_3) \mu_{f_a \rightarrow x_2}(x_2) \mu_{f_c \rightarrow x_2}(x_2) \\
 &= \sum_{x_2} f_a(x_1, x_2) \sum_{x_3} f_b(x_2, x_3) \sum_{x_4} f_c(x_2, x_4) & &= \sum_{x_2} f_b(x_2, x_3) \sum_{x_1} f_a(x_1, x_2) \sum_{x_4} f_c(x_2, x_4) \\
 &= \sum_{x_2} \sum_{x_3} \sum_{x_4} f_a(x_1, x_2) f_b(x_2, x_3) f_c(x_2, x_4) & &= \sum_{x_1} \sum_{x_2} \sum_{x_4} f_a(x_1, x_2) f_b(x_2, x_3) f_c(x_2, x_4) \\
 &= \sum_{\mathbf{x} \setminus x_1} \tilde{p}(\mathbf{x}). & &= \sum_{\mathbf{x} \setminus x_3} \tilde{p}(\mathbf{x}).
 \end{aligned}$$

$$\begin{aligned}
 \tilde{p}(x_1, x_2) &= \sum_{\mathbf{x} \setminus \{x_1, x_2\}} \tilde{p}(\mathbf{x}) = \sum_{x_3} \sum_{x_4} f_a(x_1, x_2) f_b(x_2, x_3) f_c(x_2, x_4) \\
 &= f_a(x_1, x_2) \left( \sum_{x_3} f_b(x_2, x_3) \right) \left( \sum_{x_4} f_c(x_2, x_4) \right) \\
 &= f_a(x_1, x_2) \mu_{f_b \rightarrow x_2}(x_2) \mu_{f_c \rightarrow x_2}(x_2) \\
 &= f_a(x_1, x_2) \mu_{x_2 \rightarrow f_a}(x_2) (1) \\
 &= f_a(x_1, x_2) \prod_{i \in ne(f_a)} \mu_{x_i \rightarrow f_a}(x_i).
 \end{aligned}$$

### Problem 5

Following Bishop's notation in section 8.4,

$$\begin{aligned}
 p(\mathbf{x}_s) &= f_s(\mathbf{x}_s) \prod_{i \in ne(f_s)} \mu_{x_i \rightarrow f_s}(x_i) \\
 &= f_s(\mathbf{x}_s) \prod_{i \in ne(f_s)} \prod_{l \in ne(x_i) \setminus f_s} \mu_{f_l \rightarrow x_i}(x_i) \\
 &= f_s(\mathbf{x}_s) \prod_{i \in ne(f_s)} \prod_{l \in ne(x_i) \setminus f_s} \sum_{X_l} F_l(x_i, X_l) \\
 &= \sum_{\mathbf{x} \setminus \mathbf{x}_s} f_s(\mathbf{x}_s) \prod_{i \in ne(f_s)} \prod_{l \in ne(x_i) \setminus f_s} F_l(x_i, X_l) \\
 &= \sum_{\mathbf{x} \setminus \mathbf{x}_s} p(\mathbf{x}).
 \end{aligned}$$