

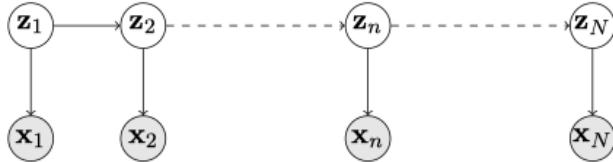
ML2 HOME ASSIGNMENT 4

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PROBLEM 1

Given BN with $X = \{x_1, \dots, x_N\}$, $Z = \{z_1, \dots, z_N\}$.

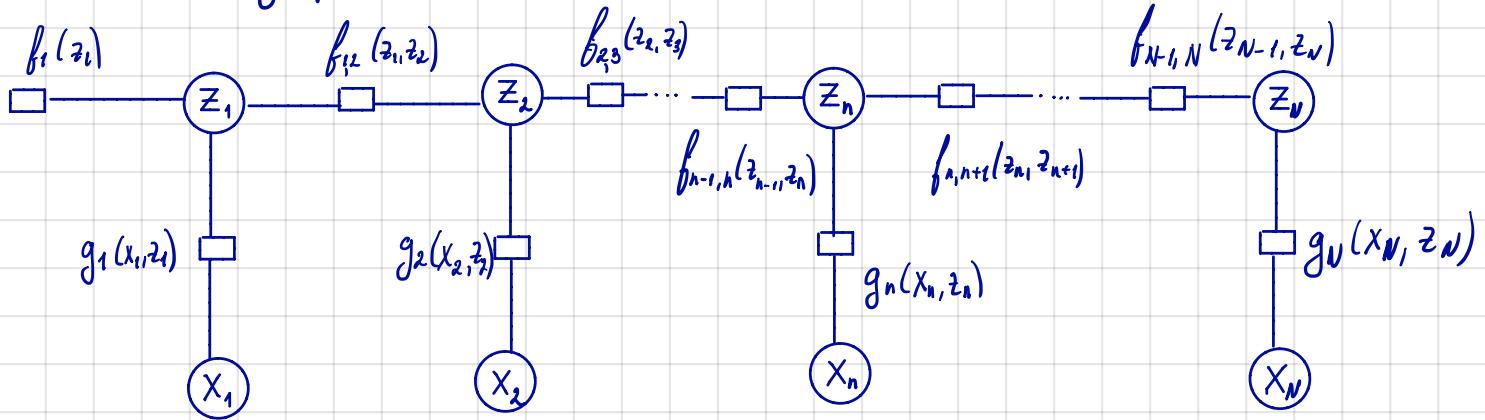


1. The factorized joint distribution $p(Z, X)$:

$$p(Z, X) = p(z_1) \cdot p(x_1|z_1) \cdot p(z_2|z_1) p(x_2|z_2) \cdots p(z_N|z_{N-1}) \cdot p(x_N|z_N)$$

$$= p(z_1) p(x_1|z_1) \prod_{i=2}^N p(z_i|z_{i-1}) p(x_i|z_i)$$

2. Factor graph!



where $\{x_1, \dots, x_N, z_1, \dots, z_N\}$ - variables and $\{f_1(z_1), f_{1,2}(z_1, z_2), \dots, f_{N-1,N}(z_{N-1}, z_N)\}$, $g_1(x_1, z_1), \dots, g_N(x_N, z_N)\}$ - factors.

3.

$$p(X, Z) = \frac{1}{Z} f_1(z_1) f_{i,z}(z_i, z_2) \dots f_{N-1,N}(z_{N-1}, z_N) \cdot g_1(x_1, z_1) \dots g_N(x_N, z_N) =$$

$$= \frac{1}{Z} f_1(z_1) \prod_{i=1}^{N-1} f_{i,i+1}(z_i, z_{i+1}) \cdot \prod_{i=1}^N g_i(x_i, z_i)$$

Because factor graph was derived from a directed graph $Z = 1$ and
 $f_i(z_i) = p(z_i)$, $f_{i,i+1}(z_i, z_{i+1}) = p(z_{i+1} | z_i)$ $\forall i = 1, N-1$ and
 $g_i(x_i, z_i) = p(x_i | z_i)$ $\forall i = 1, N$

q. $p(z_n | X) = \frac{p(X | z_n) p(z_n)}{p(X)} = \frac{\alpha(z_n) \beta(z_n)}{p(X)}$

for $n=1$: $p(z_1 | X) = \frac{p(X | z_1) p(z_1)}{p(X)} = [x_1 \perp\!\!\!\perp x_j | z_1, \text{ because } x_1 \text{ and } x_j \text{ are d-separated from } z_1]$

$$= \frac{p(x_1 | z_1) \cdot p(x_2, \dots, x_N | z_1)}{p(X)} p(z_1)$$

for $n=2$:

$$p(z_2 | X) = \frac{p(X | z_2) p(z_2)}{p(X)} = \begin{cases} x_1 \perp\!\!\!\perp x_2 | z_2 \\ x_1 \perp\!\!\!\perp x_j | z_2 \quad \forall j > 2 \\ x_2 \perp\!\!\!\perp x_j | z_2 \end{cases}$$

$$= \frac{p(x_1 | z_2) p(x_2 | z_2) \cdot p(x_3, \dots, x_N | z_2)}{p(X)} p(z_2)$$

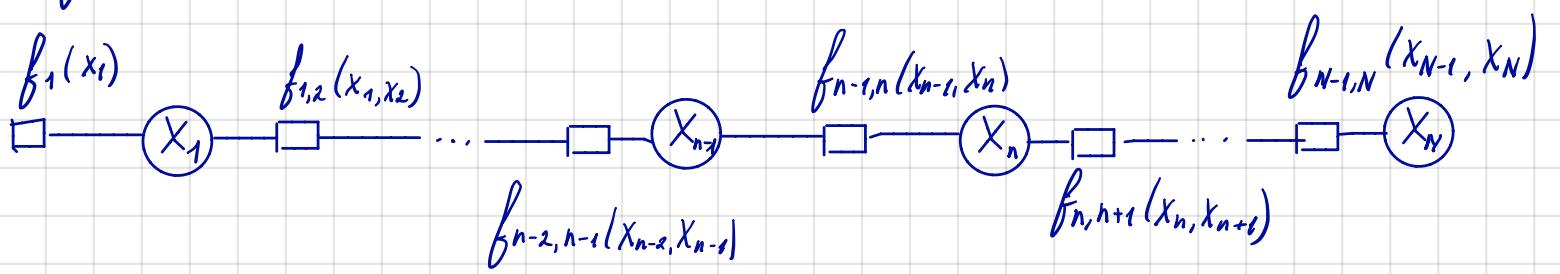
So in general case: $p(z_n | X) = \frac{p(z_n)}{p(X)} p(x_n | z_n) p(\{x_1, \dots, x_{n-1}\} | z_n) \circ p(\{x_{n+1}, \dots, x_N\} | z_n)$.

$$\alpha(z_n) = p(x_n | z_n) p(\{x_1, \dots, x_{n-1}\} | z_n) \circ p(z_n) = p(x_1, \dots, x_n, z_n)$$

$$\begin{aligned}
\alpha(z_n) &= \sum_{z_{n-1}} p(x_1, \dots, x_n, z_n | z_{n-1}) p(z_{n-1}) = \left[\{x_1, \dots, x_{n-1}, z_n \} \sqcup \{x_n, z_n\} | z_{n-1} \right] = \\
&= \sum_{z_{n-1}} p(x_1, \dots, x_{n-1} | z_{n-1}) p(x_n, z_n) p(z_{n-1}) = \sum_{z_{n-1}} p(x_1, \dots, x_{n-1}, z_{n-1}) p(x_n, z_n | z_{n-1}) \\
&= \sum_{z_{n-1}} \alpha(z_{n-1}) p(x_n, z_n | z_{n-1}) \\
\beta(z_n) &= p(x_{n+1}, \dots, x_N | z_n) = \sum_{z_{n+1}} p(x_{n+1}, \dots, x_N, z_{n+1} | z_n) = \sum_{z_{n+1}} \frac{p(x_{n+1}, \dots, x_N, z_n | z_{n+1}) p(z_{n+1})}{p(z_n)} = \\
&= \left[\{x_{n+1}, z_n\} \sqcup \{x_{n+2}, \dots, x_N\} | z_{n+1} \right] = \sum_{z_{n+1}} \frac{p(x_{n+1}, z_n, z_{n+1}) \cdot p(x_{n+2}, \dots, x_N | z_{n+1})}{p(z_n)} = \sum_{z_{n+1}} \beta(z_{n+1}) p(x_{n+1}, z_{n+1} | z_n)
\end{aligned}$$

PROBLEM 2

Let us draw a corresponding factor graph to apply sum-product algorithm:



$$\begin{aligned}
\text{with } f_{i-1,i}(x_{i-1}, x_i) &= \psi_{i-1,i}(x_{i-1}, x_i) \quad \forall i=2, N \\
f_1(x_1) &= 1.
\end{aligned}$$

$$\begin{aligned}
\tilde{p}(x_n) &= \prod_{s \in \text{ne}(x)} \mu_{fs \rightarrow x_n}(x_n) = \mu_{f_{n-1,n} \rightarrow x_n}(x_n) \circ \mu_{f_{n,n+1} \rightarrow x_n}(x_n) \quad \forall n=2, N-1 \\
\tilde{p}(x_1) &= \mu_{f_{1,2} \rightarrow x_1}(x_1) \text{ and } \tilde{p}(x_N) = \mu_{f_{N-1,N} \rightarrow x_N}(x_N).
\end{aligned}$$

Let's run sum-product algorithm to find $\mu_{fs \rightarrow x_n}(x_n)$. I choose x_N as a root node and x_1 as a leaf. So starting with a leaf node:

$$\mu_{x_i \rightarrow f_{1,2}}(x_i) = 1$$

$$\mu_{f_{1,2} \rightarrow x_2}(x_2) = \sum_{x_1} f_1(x_1, x_2)$$

$$\mu_{x_2 \rightarrow f_{2,3}}(x_2) = \mu_{f_{1,2} \rightarrow x_2}(x_2)$$

$$\mu_{f_{2,3} \rightarrow x_3}(x_3) = \sum_{x_2} f_{2,3}(x_2, x_3) \circ \mu_{x_2 \rightarrow f_{2,3}}(x_2).$$

$$\text{so } \forall i = 2, \dots, N-1: \mu_{x_i \rightarrow f_{i,i+1}}(x_i) = \mu_{f_{i,i} \rightarrow x_i}(x_i)$$

$$\begin{aligned} \mu_{f_{i,i+1} \rightarrow x_{i+1}}(x_{i+1}) &= \sum_{x_i} f_{i,i+1}(x_i, x_{i+1}) \cdot \mu_{x_i \rightarrow f_{i,i+1}}(x_i) = \\ &= \sum_{x_i} f_{i,i+1}(x_i, x_{i+1}) \cdot \mu_{f_{i,i} \rightarrow x_i}(x_i). \end{aligned}$$

Now we can propagate messages from the root node to the leaf:

$$\mu_{x_N \rightarrow f_{N-1,N}}(x_N) = 1$$

$$\mu_{f_{N-1,N} \rightarrow x_{N-1}}(x_{N-1}) = \sum_{x_N} f_{N-1,N}(x_{N-1}, x_N)$$

$$\mu_{x_{N-1} \rightarrow f_{N-2,N-1}}(x_{N-1}) = \mu_{f_{N-1,N} \rightarrow x_{N-1}}(x_{N-1})$$

$$\mu_{f_{N-2,N-1} \rightarrow x_{N-2}}(x_{N-2}) = \sum_{x_{N-1}} f_{N-2,N-1}(x_{N-2}, x_{N-1}) \circ \mu_{x_{N-1} \rightarrow f_{N-2,N-1}}(x_{N-1})$$

So general formula for messages propagated in the backward direction:

$$\mu_{x_i \rightarrow f_{i-1,i}}(x_i) = \mu_{f_{i,i+1} \rightarrow x_i}(x_i) \quad \forall i = \overline{1, N-1}$$

$$\begin{aligned} \mu_{f_{i,i+1} \rightarrow x_i}(x_i) &= \sum_{x_{i+1}} f_{i,i+1}(x_i, x_{i+1}) \mu_{x_{i+1} \rightarrow f_{i,i+1}}(x_{i+1}) = \\ &= \sum_{x_{i+1}} f_{i,i+1}(x_i, x_{i+1}) \mu_{f_{i+1,i+2} \rightarrow x_{i+1}}(x_{i+1}) \end{aligned}$$

Using the notation from the Figure 2:

$$\mu_{f_{n-1,n} \rightarrow x_n}(x_n) = \mu_\alpha(x_n) = \sum_{x_{n-1}} \psi_{n-1,n}(x_{n-1}, x_n) \mu_\alpha(x_{n-1})$$

$$\mu_{f_{n,n+1} \rightarrow x_n}(x_n) = \mu_\beta(x_n) = \sum_{x_{n+1}} \psi_{n+1,n}(x_{n+1}, x_n) \mu_\beta(x_{n+1})$$

So we have shown that $\tilde{p}(x_n) = \mu_\alpha(x_n) \cdot \mu_\beta(x_n)$, because we worked with an undirected graph $p(x_n) = \frac{1}{Z} \tilde{p}(x_n)$ where

Z is a normalizing coefficient.

PROBLEM 3

Considering that $p(x_n | x_N) \propto p(x_n, x_N)$ to evaluate $p(x_n | x_N)$ we can find $p(x_n, x_N)$.

$$p(x_n, x_N) = \sum_{X \setminus x_n, x_N} p(X) = \frac{1}{Z} \sum_{x_{n-1}} \psi_{n-1, n}(x_{n-1}, x_n).$$

$$\sum_{x_{n-2}} \psi_{n-2, n-1}(x_{n-2}, x_{n-1}) \cdot \dots \cdot \sum_{x_1} \psi_{1, 2}(x_1, x_2) \cdot \sum_{x_{n+1}} \psi_{n, n+1}(x_n, x_{n+1})$$

$$\begin{aligned} & \sum_{x_{n+2}} \psi_{n+1, n+2}(x_{n+1}, x_{n+2}) \cdot \dots \cdot \sum_{x_{N-1}} \psi_{N-2, N-1}(x_{N-2}, x_{N-1}) \psi_{N-1, N}(x_{N-1}, x_N) = \\ & = \frac{1}{Z} \mu_\alpha(x_n) \circ \tilde{\mu}_\beta(x_n, x_N). \end{aligned}$$

$$\text{with } \tilde{\mu}_\beta(x_n, x_N) =$$

$$= \sum_{x_{n+1}} \psi_{n, n+1}(x_n, x_{n+1}) \cdot \dots \cdot \sum_{x_{N-1}} \psi_{N-2, N-1}(x_{N-2}, x_{N-1}) \psi_{N-1, N}(x_{N-1}, x_N) =$$

$$= \sum_{x_{n+1}} \psi_{n, n+1}(x_n, x_{n+1}) \tilde{\mu}_\beta(x_{n+1}, x_N)$$

$$\text{and } \tilde{\mu}_\beta(x_{N-1}, x_N) = \sum_{x_{N-1}} \psi_{N-2, N-1}(x_{N-2}, x_{N-1}) \psi_{N-1, N}(x_{N-1}, x_N)$$

$$\mu_\alpha(x_n) = \sum_{x_{n-1}} \psi_{n-1, n}(x_{n-1}, x_n) \cdot \mu_\alpha(x_{n-1}).$$

μ_α does not change while μ_β should be adjusted.

PROBLEM 4

Using Bishop's equations 8.61, 8.62, 8.63, 8.64, 8.66, 8.68, 8.69:

$$\begin{aligned} p(x_s) &= \sum_{X \setminus x_s} f_s(x_s) \prod_{i \in \text{ne}(f_s)} \prod_{j \in \text{ne}(x_i) \setminus f_s} F_j(x_i, X_{ij}) = \\ &= f_s(x_s) \prod_{i \in \text{ne}(f_s)} \sum_{X \setminus x_s} \prod_{j \in \text{ne}(x_i) \setminus f_s} F_j(x_i, X_{ij}) = \\ &= f_s(x_s) \prod_{i \in \text{ne}(f_s)} \sum_{X \setminus x_s} G_i(x_i, X_{si}) = \\ &= f_s(x_s) \prod_{i \in \text{ne}(f_s)} \mu_{x_i \rightarrow f_s}(x_i) \end{aligned}$$