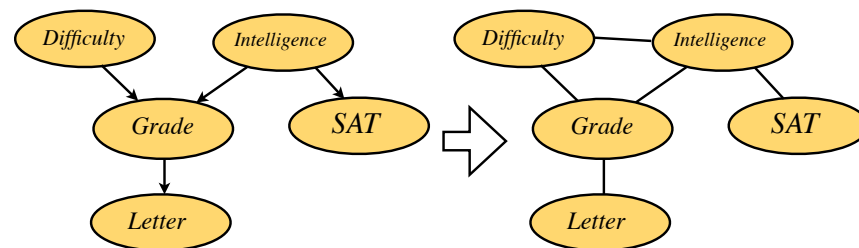


From directed to undirected

- Complete algorithm
 - Make the graph moral.
 - Initialize maximal clique potential to one.
 - Multiply each factor in $p()$ into an appropriate clique potential.
 - Note that $Z=1$

Example of converting directed to undirected



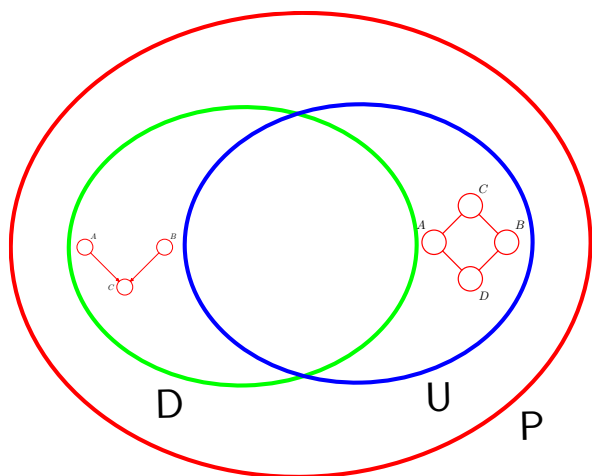
$$P(I, D, G, L, S) = P(I)P(D)P(G|I, D)P(L|G)P(S|I)$$

$$P = \psi(D, G, I)\psi(S, I)\psi(L, G)$$

$$\psi(D, G, I) = P(I)P(D)P(G|I, D) \quad \psi(S, I) = P(S|I) \quad \psi(L, G) = P(L|G)$$

$$\psi(D, G, I) = P(D)P(G|I, D) \quad \psi(S, I) = P(I)P(S|I) \quad \psi(L, G) = P(L|G)$$

Directed and undirected perfect maps



Inference on graphs

- Given a graph and its conditionals or potentials compute

$$p(\theta|e) \quad (\text{particular } \theta \text{ and } e, \text{ marginalizing out other variables})$$

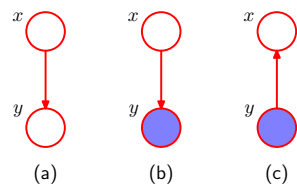
$$p(X) \quad (\text{particular event, marginalizing out other variables})$$

$$\text{argmax } p(\theta|e) \quad (\text{particular } \theta \text{ and } e; \text{ marginalizing other variables})$$

$$\text{argmax } p(\theta, \theta_N, e_N | e) \quad (\text{all variables, will nuisance / unobserved})$$

Inference on graphs

- Simplest example (Bayes' rule)
 - (a) model
 - (b) illustrates observed
 - (c) inference **reverses** the arrow



- Computationally

$$p(x|y) = \frac{p(y|x)p(x)}{\sum_{x'} p(y|x')p(x')}$$

Marginals on a chain

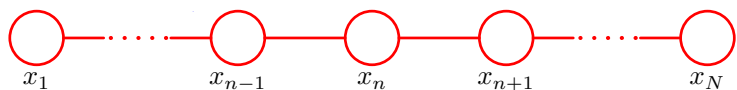
Recall $p(\mathbf{x}) = p(x_1)p(x_2|x_1)p(x_3|x_2) \cdots p(x_{N-1}|x_{N-2})p(x_N|x_{N-1})$

Converted to $p(\mathbf{x}) = \psi_{1,2}(x_1, x_2)\psi_{2,3}(x_2, x_3) \cdots \psi_{N-2,N-1}(x_{N-2}, x_{N-1})\psi_{N-1,N}(x_{N-1}, x_N)$

Assume N discrete variables, with K values each.

Compute the marginal of a node in the middle, $p(x_n)$

Marginals on a chain



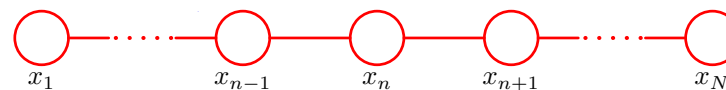
$$p(\mathbf{x}) = \psi_{1,2}(x_1, x_2)\psi_{2,3}(x_2, x_3) \cdots \psi_{N-2,N-1}(x_{N-2}, x_{N-1})\psi_{N-1,N}(x_{N-1}, x_N)$$

Direct calculation of $p(x_n)$

$$p(x_n) = \sum_{x_1} \sum_{x_2} \cdots \sum_{x_{n-1}} \sum_{x_{n+1}} \cdots \sum_{x_{N-1}} \sum_{x_N} p(\mathbf{x})$$

↑
Skip x_n

Marginals on a chain

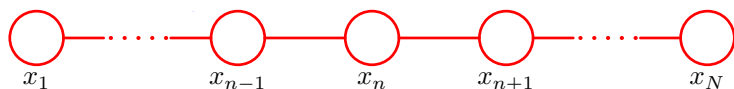


$$p(\mathbf{x}) = \psi_{1,2}(x_1, x_2)\psi_{2,3}(x_2, x_3) \cdots \psi_{N-2,N-1}(x_{N-2}, x_{N-1})\psi_{N-1,N}(x_{N-1}, x_N)$$

Direct calculation of $p(x_n)$

$$p(x_n) = \sum_{x_1} \sum_{x_2} \cdots \sum_{x_{n-1}} \sum_{x_{n+1}} \cdots \sum_{x_{N-1}} \sum_{x_N} p(\mathbf{x})$$

Computational complexity is $O(K^N)$. **Way too slow!**



$$p(x) = \psi_{1,2}(x_1, x_2) \psi_{2,3}(x_2, x_3) \dots \psi_{N-2,N-1}(x_{N-2}, x_{N-1}) \psi_{N-1,N}(x_{N-1}, x_N)$$

Main idea is to rearrange terms to exploit conditional independence.

Fancy formulas from algebra

$$\sum_{\substack{x_1, x_2, \dots, x_N \\ \text{all values of each}}} f(x_1, x_2, \dots, x_N) = \sum_{\substack{x_1 \quad x_2 \quad \dots \quad x_N \\ \text{any order you like}}} f(x_1, x_2, \dots, x_N)$$

(essentially a definition)

$$(\sum a_i)(\sum b_j) = \sum \sum a_i b_j$$

Back to
marginals
on a chain



$$p(x) = \psi_{1,2}(x_1, x_2) \psi_{2,3}(x_2, x_3) \dots \psi_{N-2,N-1}(x_{N-2}, x_{N-1}) \psi_{N-1,N}(x_{N-1}, x_N)$$

$$\begin{aligned} p(x_n) &= \sum_{x_1} \sum_{x_2} \dots \sum_{x_{n-1}} \sum_{x_{n+1}} \dots \sum_{x_{N-1}} \sum_{x_N} f_L(x_1, x_2, \dots, x_n) f_R(x_n, x_{n+1}, \dots, x_N) \\ &= \left(\sum_{x_1} \sum_{x_2} \dots \sum_{x_{n-1}} f_L(x_1, x_2, \dots, x_n) \right) \left(\sum_{x_{n+1}} \dots \sum_{x_{N-1}} \sum_{x_N} f_R(x_n, x_{n+1}, \dots, x_N) \right) \\ &= \left(\sum_{x_1} \sum_{x_2} \dots \sum_{x_{n-1}} \prod_{i=1}^{n-1} \psi_{i,i+1}(x_i, x_{i+1}) \right) \left(\sum_{x_{n+1}} \dots \sum_{x_{N-1}} \sum_{x_N} \prod_{i=n}^{N-1} \psi_{i,i+1}(x_i, x_{i+1}) \right) \\ &= \left(\sum_{x_{n-1}} \sum_{x_{n-2}} \dots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i-1, n-i}(x_{n-i-1}, x_{n-i}) \right) \left(\sum_{x_{n+1}} \dots \sum_{x_{N-1}} \sum_{x_N} \prod_{i=n}^{N-1} \psi_{i,i+1}(x_i, x_{i+1}) \right) \end{aligned}$$

$$p(x_n) = \left(\sum_{x_{n-1}} \sum_{x_{n-2}} \dots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i-1, n-i}(x_{n-i-1}, x_{n-i}) \right) \left(\sum_{x_{n+1}} \dots \sum_{x_{N-1}} \sum_{x_N} \prod_{i=n}^{N-1} \psi_{i,i+1}(x_i, x_{i+1}) \right)$$

$$\prod_{i=1}^{n-1} \psi_{n-i-1, n-i}(x_{n-i-1}, x_{n-i}) = \psi_{n-1, n}(x_{n-1}, x_n) \psi_{n-2, n-1}(x_{n-2}, x_{n-1}) \dots \psi_{2,3}(x_2, x_3) \psi_{1,2}(x_1, x_2)$$

Warmup example

$$\sum_{x_2} \sum_{x_1} \underbrace{\psi(x_2, x_3)}_{\substack{\text{No dependency on } x_1 \\ \text{hence we can move} \\ \text{factor outside sum} \\ \text{over } x_1.}} \psi(x_1, x_2) = \sum_{x_2} \psi(x_2, x_3) \sum_{x_1} \psi(x_1, x_2)$$

(Recall the distributive law: $ba + ca = a(b + c)$)

$$p(x_n) = \left(\sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) \right) \left(\sum_{x_{n+1}} \cdots \sum_{x_{N-1}} \sum_{x_N} \prod_{i=n}^{N-1} \psi_{i, i+1}(x_i, x_{i+1}) \right)$$

$$\prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) = \psi_{n-1, n}(x_{n-1}, x_n) \psi_{n-2, n-1}(x_{n-2}, x_{n-1}) \cdots \psi_{2, 3}(x_2, x_3) \psi_{1, 2}(x_1, x_2)$$

$$\begin{aligned} \sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) &= \sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_2} \prod_{i=1}^{n-2} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) \left\{ \sum_{x_1} \psi_{1,2}(x_1, x_2) \right\} \\ &= \sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_3} \prod_{i=1}^{n-3} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) \left\{ \sum_{x_2} \psi_{2,3}(x_2, x_3) \left\{ \sum_{x_1} \psi_{1,2}(x_1, x_2) \right\} \right\} \\ &\quad \dots \\ &= \left\{ \sum_{x_{n-1}} \psi_{n-1, n}(x_{n-1}, x_n) \cdots \left\{ \sum_{x_3} \psi_{3,4}(x_3, x_4) \left\{ \sum_{x_2} \psi_{2,3}(x_2, x_3) \left\{ \sum_{x_1} \psi_{1,2}(x_1, x_2) \right\} \right\} \right\} \right\} \end{aligned}$$

$$p(x_n) = \left(\sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) \right) \left(\sum_{x_{n+1}} \cdots \sum_{x_{N-1}} \sum_{x_N} \prod_{i=n}^{N-1} \psi_{i, i+1}(x_i, x_{i+1}) \right)$$

$$\sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) = \left\{ \sum_{x_{n-1}} \psi_{n-1, n}(x_{n-1}, x_n) \cdots \left\{ \sum_{x_3} \psi_{3,4}(x_3, x_4) \left\{ \sum_{x_2} \psi_{2,3}(x_2, x_3) \left\{ \sum_{x_1} \psi_{1,2}(x_1, x_2) \right\} \right\} \right\} \right\} \dots$$

and

$$\sum_{x_{n+1}} \cdots \sum_{x_{N-1}} \sum_{x_N} \prod_{i=n}^{N-1} \psi_{i, i+1}(x_i, x_{i+1}) = \left\{ \sum_{x_{n+1}} \psi_{n, n+1}(x_n, x_{n+1}) \cdots \left\{ \sum_{x_{N-1}} \psi_{N-2, N-1}(x_{N-2}, x_{N-1}) \left\{ \sum_{x_N} \psi_{N-1, N}(x_{N-1}, x_N) \right\} \right\} \right\} \dots$$

(Deriving the right factor (red) is similar.)

Matrix interpretation (for two variables)

$$\sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) = \left\{ \sum_{x_{n-1}} \psi_{n-1, n}(x_{n-1}, x_n) \cdots \left\{ \sum_{x_3} \psi_{3,4}(x_3, x_4) \left\{ \sum_{x_2} \psi_{2,3}(x_2, x_3) \left\{ \sum_{x_1} \psi_{1,2}(x_1, x_2) \right\} \right\} \right\} \right\} \dots$$

$$\psi_{i, i+1}(x_i, x_{i+1}) \Leftrightarrow Q_{i+1, i} \quad (\text{note transposition!})$$

$$\sum_i \psi_{i, i+1}(x_i, x_{i+1}) \Leftrightarrow \text{sums columns to get a vector } V_{i+1}$$

$$\sum_{x_{i+1}} \psi_{i+1, i+2}(x_{i+1}, x_{i+2}) \left\{ \sum_{x_i} \psi_{i, i+1}(x_i, x_{i+1}) \right\} \Leftrightarrow Q_{i+2, i+1} \cdot V_i$$

Computational Complexity

Suppose each variable has K values

What is the cost of evaluating the first factor?

$$\sum_{x_{n-1}} \sum_{x_{n-2}} \cdots \sum_{x_1} \prod_{i=1}^{n-1} \psi_{n-i, n-i+1}(x_{n-i}, x_{n-i+1}) = \left\{ \sum_{x_{n-1}} \psi_{n-1, n}(x_{n-1}, x_n) \cdots \left\{ \sum_{x_3} \psi_{3,4}(x_3, x_4) \left\{ \sum_{x_2} \psi_{2,3}(x_2, x_3) \left\{ \underbrace{\sum_{x_1} \psi_{1,2}(x_1, x_2)}_{\substack{\text{K sums of K values} \\ \text{K evaluations of K products}}} \right\} \right\} \right\} \right\} \dots$$

The other factor is similar.

We see that the overall cost is $O(N \cdot K^2)$.

Much better than the naive computation where we had $O(K^N)$.