## **Probabilistic Graphical Models Notes**

## **Section 1: Background**

**Conditional Independence**: Given RVs X, Y, Z we say that  $X \perp Y \mid Z$  iff  $P(X, Y \mid Z) = P(X \mid Z)P(Y \mid Z)$ Useful properties:

a) Decomposition: 
$$X \perp Y, Z \mid W \Rightarrow \begin{cases} X \perp Y \mid Z \\ X \perp Y \mid W \end{cases}$$
 b

An ordering  $I: v \mapsto \{1, \dots, n\}$  is **Topological** iff  $j \in \pi_i \Rightarrow I(j) < I(i)$  i.e. parents always come before their children If G is a DAG, then  $\exists$  a Topological Ordering on G

## **Section 2: Directed Graphical Models**

Given DAG 
$$G = (V, E)$$

$$\mathcal{L}(G) = \left\{ p \text{ is a dist over } x_v : \exists \text{ factors } f_i \text{ s.t. } p(x_v) = \prod_{i=1}^n f_i(x_i; x_{\pi_i}) \text{ and } f_i \text{ satisfies: } \begin{cases} f_i > 0 \\ \sum_{x_i} f_i(x_i; x_{\pi_i}) = 1 \\ f_i : Dom(x_i)^2 \mapsto [0, 1] \end{cases} \right\}$$

**Proposition 1 (Leaf Plucking Property).** *if* n *is a leaf of* G,  $p(x_v) \in \mathcal{L}(G - \{n\})$ 

*Proof.*  $P(x_v) = p(x_{1:n-1}, x_n) = f_n(x_n; x_{\pi_n}) \prod_{i \neq n} f_i(x_i, x_{\pi_i})$ . Next marginalizing out  $x_n$  and using that it is a leaf, we have:

$$p(x_{1:n-1}) = \underbrace{\sum_{x_n} f_n(x_n; x_{\pi_n})}_{\text{sums to 1}} \underbrace{\prod_{i=1}^{n-1} f_i(x_i; x_{\pi_i})}_{\text{Does not contain } x_n} = \prod_{i=1}^{n-1} f_i(x_i; x_{\pi_i})$$

$$x_n) \in \mathcal{L}\left(G - \{n\}\right) \text{ as needed.}$$

**Proposition 2 (Factors are Conditional PMFs).** Let  $p \in \mathcal{L}(G)$  and  $\{f_j\}$  be a factorization, then  $\forall i, P(x_i|x_{\pi_i}) = f_i(x_i;x_{\pi_i})$ 

*Proof.* WLOG let  $\{1, \cdots, n\}$  be a Topological Ordering and use Theorem 1 to get that  $p(x_{1:i}) \in \mathcal{L}(G - \{i+1, \cdots, n\})$ , and so  $p(x_{1:i}) = \prod f_j(x_j; x_{\pi_j}) f_i(x_i; x_{\pi_i})$ . We partition  $\{1:i\}$  as  $\{i\} \cup \pi_i \cup A$  and get that:

$$p(x_i \mid x_{\pi_i}) = \frac{\sum_{x_A} f_i(x_i; x_{\pi_i}) g(x_{1:i-1})}{\sum_{x_A} \sum_{x_i'} f_i(x_i'; x_{\pi_i}') g(x_{1:i-1})} = \frac{f_i(x_i; x_{\pi_i}) \sum_{x_A} g(x_{1:i-1})}{\sum_{x_i'} f_i(x_i'; x_{\pi_i}') g(x_{1:i-1})} = f_i(x_i; x_{\pi_i})$$

Note: adding edges adds more distributions i.e.  $E \subseteq E'$  and G' = (V, E') then  $\mathcal{L}(G) \subseteq \mathcal{L}(G')$ 

**Proposition 3.**  $p \in \mathcal{L}(G) \iff x_i \perp x_{nd(i)} \mid \pi_i$ 

$$Proof. \ (\Rightarrow) \ (\Leftarrow)$$

