

1. The *conditional entropy* $H(X|Y)$ is defined as the average entropy in the conditional distribution

$$H(X|Y) = \mathbb{E}[\mathbb{E}\{-\log f(X|Y) | Y\}] = \sum_y \left\{ \sum_x -f(x|y) \log f(x|y) \right\} f(y).$$

- (a) Use the information inequality to show that

$$H(X|Y) \leq H(X),$$

i.e. the *entropy is always reduced by conditioning*

- (b) Show that

$$H(X, Y) = H(X|Y) + H(Y).$$

- (c) For three discrete random variables, show that

$$H(X, Y, Z) + H(Z) \leq H(X, Z) + H(Y, Z).$$

- (d) Show further that

$$X \perp\!\!\!\perp Y | Z \iff H(X, Y, Z) + H(Z) = H(X, Z) + H(Y, Z).$$

2. Consider a directed graph $\mathcal{D} = (V, E)$ and assume given $k_v, v \in V$ with $k_v \geq 0$ and $\sum_{x_v \in \mathcal{X}_v} k_v(x_v | x_{\text{pa}(v)}) = 1$. Define

$$p(x) = \prod_{v \in V} k_v(x_v | x_{\text{pa}(v)}).$$

- (a) Show that when \mathcal{D} is acyclic, i.e. a DAG, this yields a well-defined probability distribution;
 (b) Show that it holds that

$$p(x_v | x_{\text{pa}(v)}) = k_v(x_v | x_{\text{pa}(v)}). \quad (1)$$

- (c) Give a counterexample in the case where \mathcal{D} has directed cycles.

Hint: Use induction for (a) and (b), exploiting that a DAG always has a terminal vertex v_0 , i.e. a vertex with no children.

3. Consider a DAG \mathcal{D} with arrows $1 \rightarrow 2, 2 \rightarrow 5, 2 \rightarrow 3, 5 \rightarrow 6, 4 \rightarrow 5, 4 \rightarrow 7, 5 \rightarrow 7$.
- Draw the DAG;
 - List all conditional independence relations corresponding to the local, directed Markov property;
 - List all conditional independence relations corresponding to the ordered Markov property for the well-ordering induced by the given numbering;
 - Find the ancestral sets generated by the following subsets:
 - $\{5\}$;
 - $\{2, 7\}$;
 - $\{4, 6\}$;
 - Which of the following separation statements are true? For those that are not true, identify an active trail.
 - $2 \perp_{\mathcal{D}} 4 \mid 5$;
 - $2 \perp_{\mathcal{D}} 7 \mid 5$,
 - $1 \perp_{\mathcal{D}} 7 \mid 5, 6$;
 - $1 \perp_{\mathcal{D}} 4 \mid 6$;
4. Let P be a distribution which factorizes over the DAG \mathcal{D} and let $G(P)$ be its dependence graph. Show that $G(P) \subseteq \mathcal{D}^m$, where \mathcal{D}^m is the moral graph of \mathcal{D} .
5. A DAG \mathcal{D} is said to be *perfect* if all parents are married, i.e. if it holds that

$$\alpha, \beta \in \text{pa}(\gamma) \Rightarrow \alpha \rightarrow \beta \text{ or } \beta \rightarrow \alpha.$$

- Show that a perfect DAG \mathcal{D} is Markov equivalent to its *skeleton* i.e. the undirected graph obtained by ignoring directions on all arrows;
- Show the converse, i.e. that if \mathcal{D} is Markov equivalent to its skeleton, the \mathcal{D} is perfect.
- Show that the skeleton of a perfect \mathcal{D} is chordal.
- Show that the edges of a chordal graph \mathcal{G} can be directed to create a Markov equivalent DAG.

Hint: Exploit the existence of a perfect numbering of a chordal graph.