

## Independencies encoded in BN

- We said: All you need is the local Markov assumption
$\square\left(\mathrm{X}_{\mathrm{i}} \perp\right.$ NonDescendants $\left._{\mathrm{xi}_{\mathrm{i}}} \mid \mathrm{Pa}_{\mathrm{xi}_{\mathrm{i}}}\right)$
- But then we talked about other (in)dependencies
$\square$ e.g., explaining away
$A \rightarrow B \rightarrow C \rightarrow D$
$A \perp D \mid B$

- What are the independencies encoded by a BN?

Only assumption is local Markov
But many others can be derived using the algebra of conditional independencies!!!



## An active trail - Example



When are $A$ and $H$ independent?

## Active trails formalized

- A trail $X_{1}-X_{2}-\cdots-X_{k}$ is an active trail when variables $\mathbf{O} \subseteq\left\{X_{1}, \ldots, X_{n}\right\}$ are observed if for each consecutive triplet in the trail:
$\square X_{i-1} \rightarrow X_{i} \rightarrow X_{i+1}$, and $X_{i}$ is not observed ( $X_{i} \notin \mathbf{O}$ )
$\square X_{i-1} \leftarrow X_{i} \leftarrow X_{i+1}$, and $X_{i}$ is not observed $\left(X_{i} \notin \mathbf{O}\right)$
$\square \mathrm{X}_{\mathrm{i}-1} \leftarrow \mathrm{X}_{\mathrm{i}} \rightarrow \mathrm{X}_{\mathrm{i}+1}$, and $\mathrm{X}_{\mathrm{i}}$ is not observed $\left(\mathrm{X}_{\mathrm{i}} \notin \mathbf{O}\right)$
$\square X_{i-1} \rightarrow X_{i} \leftarrow X_{i+1}$, and $X_{i}$ is observed ( $X_{i} \in O$ ), or one of its descendents


## Active trails and independence?

- Theorem: Variables $X_{i}$ and $X_{j}$ are independent given $Z \subseteq\left\{X_{1}, \ldots, X_{n}\right\}$ if the is no active trail between $X_{i}$ and $X_{j}$ when variables $Z \subseteq\left\{X_{1}, \ldots, X_{n}\right\}$ are observed



## More generally: <br> Soundness of d-separation

- Given BN structure G
- Set of independence assertions obtained by d-separation:

$$
\square \mathbf{I}(G)=\left\{(\mathbf{X} \perp \mathbf{Y} \mid \mathbf{Z}): d-\operatorname{sep}_{G}(\mathbf{X} ; \mathbf{Y} \mid \mathbf{Z})\right\}
$$

- Theorem: Soundness of d-separation
$\square$ If $P$ factorizes over $G$ then $I(G) \subseteq I(P)$
- Interpretation: d-separation only captures true independencies
- Proof discussed when we talk about undirected models


## Existence of dependency when not d-separated

- Theorem: If $X$ and $Y$ are not d-separated given $\mathbf{Z}$, then $X$ and $Y$ are dependent given $\mathbf{Z}$ under some $P$ that factorizes over G
- Proof sketch:

Choose an active trail between $X$ and $Y$ given $Z$ Make this trail dependent
Make all else uniform
 (independent) to avoid "canceling" out influence

## More generally: <br> Completeness of d-separation

- Theorem: Completeness of d-separation
$\square$ For "almost all" distributions where $P$ factorizes over to $G$, we have that $I(G)=I(P)$
- "almost all" distributions: except for a set of measure zero of parameterizations of the CPTs (assuming no finite set of parameterizations has positive measure)
- Means that if all sets $\mathbf{X}$ \& $\mathbf{Y}$ that are not d-separated given $\mathbf{Z}$, then $\neg(\mathbf{X} \perp \mathbf{Y} \mid \mathbf{Z})$
- Proof sketch for very simple case:


## Interpretation of completeness

- Theorem: Completeness of d-separation

For "almost all" distributions that $P$ factorize over to $G$, we have that $I(G)=I(P)$

- BN graph is usually sufficient to capture all independence properties of the distribution!!!!
- But only for complete independence:
$P \rightarrow(\mathbf{X}=\mathbf{x} \perp \mathbf{Y}=\mathbf{y} \mid \mathbf{Z}=\mathbf{z}), \forall \mathbf{x} \in \operatorname{Val}(\mathbf{X}), \mathbf{y} \in \operatorname{Val}(\mathbf{Y}), \mathbf{z} \in \operatorname{Val}(\mathbf{Z})$
- Often we have context-specific independence (CSI)
$\square \exists \mathbf{x} \in \operatorname{Val}(\mathbf{X}), \mathbf{y} \in \operatorname{Val}(\mathbf{Y}), \mathbf{z} \in \operatorname{Val}(\mathbf{Z}): P \rightarrow(\mathbf{X}=\mathbf{x} \perp \mathbf{Y}=\mathbf{y} \mid \mathbf{Z}=\mathbf{z})$
Many factors may affect your grade
$\square$ But if you are a frequentist, all other factors are irrelevant ©


## Algorithm for d-separation

- How do $I$ check if X and Y are dseparated given $\mathbf{Z}$
$\square$ There can be exponentially-many trails between X and Y
- Two-pass linear time algorithm finds all d-separations for $X$
- 1. Upward pass
$\square$ Mark descendants of $\mathbf{Z}$
- 2. Breadth-first traversal from $X$ Stop traversal at a node if trail is "blocked"
(Some tricky details apply - see reading)



## What you need to know

- d-separation and independence
sound procedure for finding independencies
$\square$ existence of distributions with these independencies
$\square$ (almost) all independencies can be read directly from graph without looking at CPTs


## Announcements

- Homework 1:
$\square$ Due next Wednesday - beginning of class!
$\square$ It's hard - start early, ask questions
- Audit policy
$\square$ No sitting in, official auditors only, see course website


## Building BNs from independence properties

- From d-separation we learned:

Start from local Markov assumptions, obtain all independence assumptions encoded by graph
For most $P$ 's that factorize over $G, I(G)=I(P)$
All of this discussion was for a given $G$ that is an I-map for $P$

Now, give me a $P$, how can I get a $G$ ?
i.e., give me the independence assumptions entailed by $P$

Many $G$ are "equivalent", how do I represent this?
Most of this discussion is not about practical algorithms, but useful concepts that will be used by practical algorithms

- Practical algs next time


## Minimal I-maps

- One option:
$G$ is an I-map for $P$
$\square G$ is as simple as possible
- $G$ is a minimal l-map for $P$ if deleting any edges from $G$ makes it no longer an I-map


## Obtaining a minimal I-map

- Given a set of variables and conditional independence assumptions
- Choose an ordering on variables, e.g., $X_{1}, \ldots, X_{n}$
For $\mathrm{i}=1$ to n
$\square$ Add $\mathrm{X}_{\mathrm{i}}$ to the network
$\square$ Define parents of $X_{i}, P a_{x_{X}}$, in graph as the minimal subset of $\left\{\mathrm{X}_{1}, \ldots, \mathrm{X}_{\mathrm{i}-1}\right\}$ such that local Markov assumption holds - $X_{i}$ independent of rest of $\left\{\mathrm{X}_{1}\right.$ $\left., \ldots, \mathrm{X}_{\mathrm{i}-1}\right\}$, given parents $\mathrm{Pa}_{\mathrm{X}_{\mathrm{i}}}$ $\square$ Define/learn CPT - $\mathrm{P}\left(\mathrm{X}_{\mathrm{i}} \mid \mathrm{Pa}_{\mathrm{x}_{\mathrm{i}}}\right)$


## Minimal I-map not unique (or minimum)

- Given a set of variables and assumptions
- Choose an ordering on variables, e.g., $\mathrm{X}_{1}, \ldots, \mathrm{X}_{\mathrm{n}}$
- For $\mathrm{i}=1$ to n
$\square$ Add $\mathrm{X}_{\mathrm{i}}$ to the network
$\square$ Define parents of $X_{i}, \mathrm{~Pa}_{\mathrm{X}}$, in graph as the minimal subset of $\left\{\mathrm{X}_{1}, \ldots, \mathrm{X}_{\mathrm{i}-1}\right\}$ such that local Markov assumption holds $-X_{i}$ independent of rest of $\left\{\mathrm{X}_{1}\right.$ $\left., \ldots, \mathrm{X}_{\mathrm{i}-1}\right\}$, given parents $\mathrm{Pa}_{\mathrm{x}_{\mathrm{i}}}$
Define/learn CPT - $\mathrm{P}\left(\mathrm{X}_{\mathrm{i}} \mid \mathrm{Pa}_{\mathrm{x}_{\mathrm{i}}}\right)$


## Perfect maps (P-maps)

- I-maps are not unique and often not simple enough

■ Define "simplest" $G$ that is I-map for $P$
A BN structure $G$ is a perfect map for a distribution $P$ if $I(P)=I(G)$

- Our goal:

Find a perfect map!
Must address equivalent BNs

## Inexistence of P-maps 1

- XOR (this is a hint for the homework)

Inexistence of P-maps 2

■ (Slightly un-PC) swinging couples example

## Obtaining a P-map

- Given the independence assertions that are true for $P$
- Assume that there exists a perfect map $\mathrm{G}^{*}$
$\square$ Want to find $\mathrm{G}^{*}$
- Many structures may encode same independencies as $\mathrm{G}^{*}$, when are we done?

Find all equivalent structures simultaneously!

## I-Equivalence

- Two graphs $G_{1}$ and $G_{2}$ are I-equivalent if $\mathrm{I}\left(G_{1}\right)=\mathrm{I}\left(G_{2}\right)$
- Equivalence class of BN structures
$\square$ Mutually-exclusive and exhaustive partition of graphs
- How do we characterize these equivalence classes?


## Skeleton of a BN

- Skeleton of a BN structure $G$ is an undirected graph over the same variables that has an edge $\mathrm{X}-\mathrm{Y}$ for every $\mathrm{X} \rightarrow \mathrm{Y}$ or $\mathrm{Y} \rightarrow \mathrm{X}$ in $G$
- (Little) Lemma: Two I -equivalent BN structures must have the same skeleton

- Theorem: If $G_{1}$ and $G_{2}$ have the same skeleton and $V$-structures, then $G_{1}$ and $G_{2}$ are I-equivalent


## Same V-structures not necessary

- Theorem: If $G_{1}$ and $G_{2}$ have the same skeleton and $V$-structures, then $G_{1}$ and $G_{2}$ are l-equivalent
- Though sufficient, same V -structures not necessary


## Immoralities \& I-Equivalence

- Key concept not V-structures, but "immoralities" (unmarried parents ©)
$\square X \rightarrow Z \leftarrow Y$, with no arrow between $X$ and $Y$
$\square$ Important pattern: X and Y independent given their parents, but not given $Z$
$\square$ (If edge exists between X and Y , we have covered the V-structure)
- Theorem: $G_{1}$ and $G_{2}$ have the same skeleton and immoralities if and only if $G_{1}$ and $G_{2}$ are I-equivalent


## Obtaining a P-map

- Given the independence assertions that are true for $P$
$\square$ Obtain skeleton
$\square$ Obtain immoralities
- From skeleton and immoralities, obtain every (and any) BN structure from the equivalence class

Identifying the skeleton 1

- When is there an edge between X and Y ?
- When is there no edge between X and Y ?


## Identifying the skeleton 2

Assume d is max number of parents (d could be $n$ )

- For each $X_{i}$ and $X_{j}$
$\square \mathrm{E}_{\mathrm{ij}} \leftarrow$ true
$\square$ For each $\mathbf{U} \subseteq \mathbf{X}-\left\{\mathrm{X}_{\mathrm{i}}, \mathrm{X}_{\mathrm{j}}\right\},|\mathbf{U}| \leq \mathrm{d}$ - Is $\left(\mathrm{X}_{\mathrm{i}} \perp \mathrm{X}_{\mathrm{j}} \mid \mathrm{U}\right)$ ?
$\mathrm{E}_{\mathrm{ij}} \leftarrow$ false
$\square$ If $\mathrm{E}_{\mathrm{ij}}$ is true
- Add edge X - Y to skeleton


## Identifying immoralities

- Consider $\mathrm{X}-\mathrm{Z}-\mathrm{Y}$ in skeleton, when should it be an immorality?

■ Must be $X \rightarrow Z \leftarrow Y$ (immorality):
$\square$ When X and Y are never independent given $\mathbf{U}$, if $\mathrm{Z} \in \mathbf{U}$

- Must not be $X \rightarrow Z \leftarrow Y$ (not immorality):

When there exists $\mathbf{U}$ with $Z \in \mathbf{U}$, such that $X$ and $Y$ are independent given $\mathbf{U}$

## From immoralities and skeleton to BN structures

- Representing BN equivalence class as a partially-directed acyclic graph (PDAG)
- Immoralities force direction on some other BN edges
- Full (polynomial-time) procedure described in reading


## What you need to know

- Minimal I-map
$\square$ every $P$ has one, but usually many
- Perfect map
$\square$ better choice for BN structure
$\square$ not every $P$ has one
$\square$ can find one (if it exists) by considering l-equivalence
Two structures are I-equivalent if they have same skeleton and immoralities

