

Bayesian Networks

2. Separation in Graphs

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Bayesian Networks



1. Separation in Undirected Graphs

- 2. Properties of Ternary Relations on Sets
- 3. Separation in Directed Graphs

Graphs

Definition 1. Let V be any set and

 $E \subseteq \mathcal{P}^2(V) := \{\{x, y\} \mid x, y \in V\}$

be a subset of sets of unordered pairs of V. Then G := (V, E) is called an **undirected graph**. The elements of V are called **vertices** or **nodes**, the elements of E edges.

Let $e = \{x, y\} \in E$ be an edge, then we call the vertices x, y **incident** to the edge e. We call two vertices $x, y \in V$ **adjacent**, if there is an edge $\{x, y\} \in E$.

The set of all vertices adjacent with a given vertex $x \in V$ is called its **fan**:

$$fan(x) := \{ y \in V \mid \{x, y\} \in E \}$$





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Bayesian Networks / 1. Separation in Undirected Graphs



Paths on graphs

Definition 2. Let V be a set. We call $V^* := \bigcup_{i \in \mathbb{N}} V^i$ the set of finite sequences in V. The length of a sequence $s \in V^*$ is denoted by |s|. Let G = (V, E) be a graph. We call $G^* := V^*_{|G} := \{ p \in V^* \mid \{ p_i, p_{i+1} \} \in E,$ Ħ $i = 1, \ldots, |p| - 1$ Figure 2: Example graph. The sequences the set of paths on G. (A, D, G, H)(C, E, B, D)Any contiguous subsequence of a path (F) $p \in G^*$ is called a **subpath of** p, i.e. any path $(p_i, p_{i+1}, \ldots, p_i)$ with $1 \le i \le j \le n$. are paths on G, but the sequences The subpath $(p_2, p_3, \ldots, p_{n-1})$ is called (A, D, E, C)the interior of p. A path of length $|p| \ge 2$ (A, H, C, F)is called **proper**.

Separation in graphs (u-separation)

Definition 3. Let G := (V, E) be a graph. | We write $I_G(X, Y|Z)$ for the statement, Let $Z \subseteq V$ be a subset of vertices. We say, two vertices $x, y \in V$ are **u**separated by Z in G, if every path from x to y contains some vertex of Z ($\forall p \in$ $G^*: p_1 = x, p_{|p|} = y \Rightarrow \exists i \in \{1, \dots, n\}:$ $p_i \in Z$).

Let $X, Y, Z \subseteq V$ be three disjoint subsets of vertices. We say, the vertices Xand Y are **u-separated by** Z in G, if every path from any vertex from X to any vertex from Y is separated by Z, i.e., contains some vertex of Z.

that X and Y are u-separated by Z in G.

 I_G is called **u-separation relation in** G.



Figure 3: Example for u-separation [CGH97, p. 179].

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Separation in graphs (u-separation)



Figure 4: More examples for u-separation [CGH97, p. 179].

Properties of u-separation / no chardality

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For u-separation the chordality property does not hold (in general).

Figure 5: Counterexample for chordality in undirected graphs (u-separation) [CGH97, p. 189].

Bayesian Networks

Checking u-separation

To test, if for a given graph $G = (V, E)$ two given sets $X, Y \subseteq V$ of vertices are u-separated by a third given set $Z \subseteq V$ of vertices, we may use standard breadth-first search to compute all ver- tices that can be reached from X (see, e.g., [OW02], [CLR90]).	 For checking u-separation we have to tweak the algorithm 1. not to add vertices from <i>Z</i> to the border and 2. to stop if a vertex of <i>Y</i> has been reached.
1 breadth-first search(G, X) : 2 border := X 3 reached := Ø 4 while border $\neq \emptyset$ do 5 reached := reached \cup border 6 border := fan _G (border) \ reached 7 od 8 return reached	<pre>1 check-u-separation(G, X, Y, Z): 2 border := X 3 reached := \emptyset 4 <u>while</u> border $\neq \emptyset$ <u>do</u> 5 reached := reached \cup border 6 border := fan_G(border) \ reached \ Z 7 <u>if</u> border $\cap Y \neq \emptyset$ 8 <u>return</u> false 9 <u>fi</u> 10 <u>od</u> 11 <u>return</u> true</pre>
Figure 6: Breadth-first search algorithm for enu- merating all vertices reachable from <i>X</i> .	Figure 7: Breadth-first search algorithm for checking u-separation of X and Y by Z .
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1. Separation in Undirected Graphs

2. Properties of Ternary Relations on Sets

3. Separation in Directed Graphs

Symmetry

Definition 4. Let *V* be any set and *I* a ternary relation on $\mathcal{P}(V)$, i.e., $I \subseteq (\mathcal{P}(V))^3$.

I is called **symmetric**, if

 $I(X,Y|Z) \Rightarrow I(Y,X|Z)$

Figure 8: Examples for symmetry [CGH97, p. 186].

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Decomposition and Composition

Definition 5. I is called (right-)decomposable, if

 $I(X,Y|Z) \Rightarrow I(X,Y'|Z) \quad \text{for any } Y' \subseteq Y$

I is called (right-)composable, if

 $I(X,Y|Z) \text{ and } I(X,Y'|Z) \Rightarrow I(X,Y\cup Y'|Z)$

Figure 9: Examples for decomposition [CGH97, p. 186].

Union

Definition 6. I is called strongly unionable, if

 $I(X,Y|Z) \Rightarrow I(X,Y|Z \cup Z') \quad \text{ for all } Z' \text{ disjunct with } X,Y$

I is called (right-)weakly unionable, if

 $I(X,Y|Z) \Rightarrow I(X,Y'|(Y \setminus Y') \cup Z) \quad \text{for any } Y' \subseteq Y$

Figure 10: Examples for a) strong union and b) weak union [CGH97, p. 186,189]. Lars Schmidt-Thieme, Information Systems and Machine Learning Lab (ISMLL), Institute of Computer Science, University of Hildesheim Course on Bayesian Networks, winter term 2013/14

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Contraction and Intersection

Definition 7. *I* is called (right-)contractable, if

 $I(X,Y|Z) \text{ and } I(X,Y'|Y\cup Z) \Rightarrow I(X,Y\cup Y'|Z)$

I is called (right-)intersectable, if

 $I(X,Y|Y'\cup Z) \text{ and } I(X,Y'|Y\cup Z) \Rightarrow I(X,Y\cup Y'|Z)$

Figure 11: Examples for a) contraction and b) intersection [CGH97, p. 186].

Transitivity

Definition 8. I is called strongly transitive, if

 $I(X,Y|Z) \Rightarrow I(X,\{v\}|Z) \text{ or } I(\{v\},Y|Z) \quad \forall v \in V \setminus Z$

I is called weakly transitive, if

 $I(X,Y|Z) \text{ and } I(X,Y|Z\cup\{v\}) \Rightarrow I(X,\{v\}|Z) \text{ or } I(\{v\},Y|Z) \quad \forall v \in V \backslash Z$

Figure 12: Examples for a) strong transitivity and b) weak transitivity. [CGH97, p. 189]

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Chordality

Definition 9. *I* is called chordal, if

 $I(\{a\},\{c\}|\{b,d\}) \text{ and } I(\{b\},\{d\}|\{a,c\}) \Rightarrow I(\{a\},\{c\}|\{b\}) \text{ or } I(\{a\},\{c\}|\{d\})$

Figure 13: Example for chordality.

- 1. Separation in Undirected Graphs
- 2. Properties of Ternary Relations on Sets
- **3. Separation in Directed Graphs**

Definition 10. Let V be any set and

$$E \subseteq V \times V$$

be a subset of sets of ordered pairs of V. Then G := (V, E) is called a **directed** graph. The elements of V are called **vertices** or **nodes**, the elements of E edges.

Let $e = (x, y) \in E$ be an edge, then we call the vertices x, y **incident** to the edge e. We call two vertices $x, y \in V$ **adjacent**, if there is an edge $(x, y) \in E$ or $(y, x) \in E$.

The set of all vertices with an edge from a given vertex $x \in V$ is called its **fanout**:

fanout $(x) := \{ y \in V \mid (x, y) \in E \}$

The set of all vertices with an edge to a given vertex $x \in V$ is called its **fanin**:

$$fanin(x) := \{ y \in V \mid (y, x) \in E \}$$

Figure 14: Fanin (orange) and fanout (green) of a node (blue).

Paths on directed graphs

Definition 11. Let G = (V, E) be a directed graph. We call

$$G^* := V^*_{|G|} := \{ p \in V^* \mid (p_i, p_{i+1}) \in E, \\ i = 1, \dots, |p| - 1 \}$$

the **set of paths on** *G*. For two vertices $x, y \in V$ we denote by

$$G^*_{[x,y]} := \{ p \in V^*_{|G|} \mid p_1 = x, p_{|p|} = y \}$$

the set of paths from x to y.

The notions of **subpath**, **interior**, and **proper path** carry over to directed graphs.

A proper path $p = (p_1, \ldots, p_n) \in G^*$ with $p_1 = p_n$ is called **cyclic**. A path without cyclic subpath is called a **simple path**. A graph without a cyclic path is called **directed acyclig graph (DAG)**.

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Paths on directed graphs (2/2)

Definition 12. For a DAG *G* vertices of the fanout are also called **children** $child(x) := fanout(x) := \{y \in V \mid (x, y) \in E\}$ and the vertices of the fanin **parents**: $pa(x) := fanin(x) := \{y \in V \mid (y, x) \in E\}$

Vertices y with a proper path from y to x are called **ancestors of** x:

anc(x) := {
$$y \in V | \exists p \in G^* : |p| \ge 2$$
,
 $p_1 = y, p_{|p|} = x$ }

Vertices y with a proper path from x to y are called **descendents of** x:

$$\label{eq:desc} \begin{split} \operatorname{desc}(x) &:= \{ y \in V \, | \, \exists p \in G^* : |p| \geq 2, \\ p_1 = x, p_{|p|} = y \rbrace \end{split}$$

Vertices that are not a descendent of x are called **nondescendents of** x.

Figure 16: Parents/Fanin (orange) and additional ancestors (light orange), children/fanout (green) and additional descendants (light green) of a node (blue).

Definition 13. Let G := (V, E) be a directed graph. We can construct an **undirected skeleton** u(G) := (V, u(E)) of *G* by dropping the directions of the edges:

$$u(E) := \{\{x,y\} \,|\, (x,y) \in E \text{ or } (y,x) \in E\}$$

The paths on u(G) are called **chains of** G:

$$G^{\blacktriangle} := u(G)^*$$

i.e., a chain is a sequence of vertices that are linked by a forward or a backward edge. If we want to stress the directions of the linking edges, we denote a chain $p = (p_1, \ldots, p_n) \in G^{\blacktriangle}$ by

$$p_1 \leftarrow p_2 \rightarrow p_3 \leftarrow \cdots \leftarrow p_{n-1} \rightarrow p_n$$

The notions of **length**, **subchain**, **interior** and **proper** carry over from undi-

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 (p_{i-1}, p_i, p_{i+1}) there is

 $\begin{cases} p_i \in Z, & \text{if not } p_{i-1} \to p_i \leftarrow p_{i+1} \\ p_i \notin Z \cup \operatorname{anc}(Z), & \text{else} \end{cases}$

Figure 18: Chain (A, B, E, D, F) is blocked by $Z = \{B\}$ at 2.

Figure 17: Chain (A, B, E, D, F) on directed graph and path on undirected skeleton.

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Blocked chains / more examples

Figure 19: Chain (A, B, E, D, F) is blocked by $Z = \emptyset$ at 3.

Figure 20: Chain (A, B, E, D, F) is **not** blocked by $Z = \{E\}$ at 3.

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The moral graph

Definition 15. Let G := (V, E) be a DAG.

As the **moral graph of** G we denote the undirected skeleton graph of G plus additional edges between each two parents of a vertex, i.e. moral(G) := (V, E') with

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Separation in DAGs (d-separation)

Let G := (V, E) be a DAG.

Let $X, Y, Z \subseteq V$ be three disjoint subsets of vertices. We say, the vertices Xand Y are **separated by** Z **in** G, if

- (i) every chain from any vertex from *X* to any vertex from *Y* is blocked by *Z* or equivalently
- (ii) X and Y are u-separated by Z in the moral graph of the ancestral hull of $X \cup Y \cup Z$.

We write $I_G(X, Y|Z)$ for the statement, that *X* and *Y* are separated by *Z* in *G*.

Figure 23: Are the vertices A and D separated by C in G?

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Separation in DAGs (d-separation) / examples

Figure 24: A and D are separated by C in G.

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Separation in DAGs (d-separation) / more examples

Figure 25: A and D are not separated by $\{C, G\}$ in G.

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Checking d-separation

To test, if for a given graph G = (V, E) two given sets $X, Y \subseteq V$ of vertices are d-separated by a third given set $Z \subseteq V$ of vertices, we may

- build the moral graph of the ancestral hull and
- apply the u-separation criterion.
 - *i* check-d-separation(G, X, Y, Z):
 - $2 G' := \operatorname{moral}(\operatorname{anc}_G(X \cup Y \cup Z))$
 - ³ <u>return</u> check-u-separation(G', X, Y, Z)

Figure 26: Algorithm for checking d-separation via u-separation in the moral graph.

A drawback of this algorithm is that we have to rebuild the moral graph of the ancestral hull whenever X or Ychanges.

i enumerate-d-separation
$$(G = (V, E), X, Z)$$
:

```
<sup>2</sup> borderForward := \emptyset
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- *3* borderBackward $:= X \setminus Z$
- 4 reached := \emptyset
- while borderForward $\neq \emptyset$ or borderBackward $\neq \emptyset$ do 5
- reached := reached \cup (borderForward $\setminus Z$) \cup borderBackward 6
- borderForward := fanout_G(borderBackward \cup (borderForward $\setminus Z$)) \setminus reached 7
- $borderBackward := fanin_G(borderBackward \cup (borderForward \cap (Z \cup anc(Z)))) \setminus Z \setminus reached$ 8

non-blocked chains.

```
9 od
```

10 return $V \setminus reached$

Figure 28: Algorithm for enumerating all vertices d-separated from X by Z in G via restricted breadth-first search (see [Nea03, p. 80–86] for another formulation).

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Properties of d-separation / no strong union

For d-separation the strong union property does not hold.

I is called **strongly unionable**, if

 $I(X, Y|Z) \Rightarrow I(X, Y|Z \cup Z')$ for all Z' disjunct with X, Y

Figure 29: Example for strong union in undirected graphs (u-separation) [CGH97, p. 189].

Figure 30: Counterexample for strong unions in DAGs (d-separation).

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Properties of d-separation / no strong transitivity

For d-separation the strong transitivity property does not hold.

I is called strongly transitive, if

 $I(X, Y|Z) \Rightarrow I(X, \{v\}|Z) \text{ or } I(\{v\}, Y|Z) \quad \forall v \in V \setminus Z$

or

Figure 31: Example for strong transitivity in undirected graphs (u-separation) [CGH97, p. 189]. in DAGs (d-separation). Lars Schmidt-Thieme, Information Systems and Machine Learning Lab (ISMLL), Institute of Computer Science, University of Hildesheim Course on Bayesian Networks, winter term 2013/14

Figure 32: Counterexample for strong transitivity

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Properties of d-separation

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