

Probabilistic  
Graphical  
Models



Representation

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Independencies

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I-maps and  
Perfect Maps

# Capturing Independencies in $P$

$$\underline{I(P)} = \{ \underline{(X \perp Y | Z)} : \underline{P} \models \underline{(X \perp Y | Z)} \}$$

*dist* (pointing to  $P$ )  
*independencies that hold in  $P$*  (pointing to the set)

- $P$  factorizes over  $G \Leftrightarrow G$  is an I-map for  $P$ :

$$\text{d-separation } \underline{I(G)} \subseteq I(P)$$

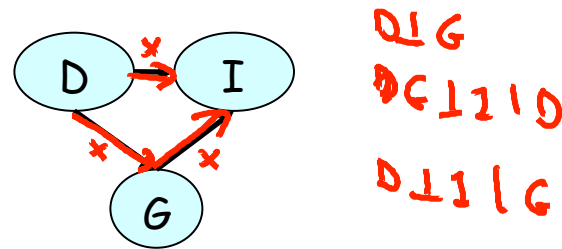
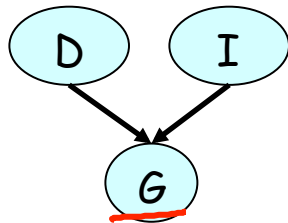
- But not always vice versa: there can be independencies in  $I(P)$  that are not in  $I(G)$

# Want a Sparse Graph

- If the graph encodes more independencies
  - it is sparser (has fewer parameters)
  - and more informative
- Want a graph that captures as much of the structure in  $P$  as possible

# Minimal I-map

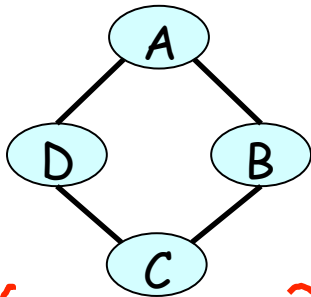
- Minimal I-map: I-map without redundant edges  
 $\textcircled{D} \not\rightarrow \textcircled{I} \textcircled{G} \quad P(Y|X^*) = P(Y|X')$
- Minimal I-map may still not capture  $I(P)$



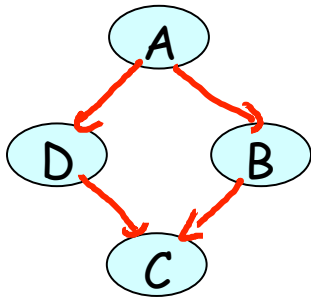
# Perfect Map

- Perfect map:  $I(G)$  =  $I(P)$ 
  - $G$  perfectly captures independencies in  $P$

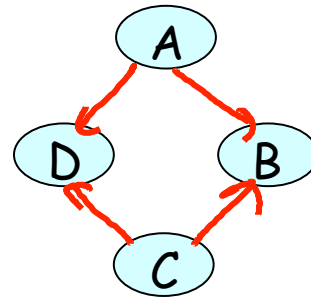
# Perfect Map



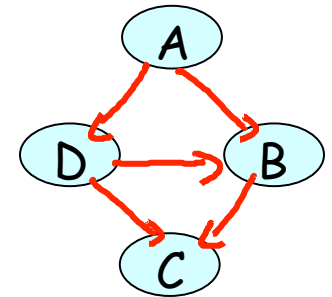
$P = \left\{ \begin{array}{l} A \perp C \mid B, D \\ B \perp D \mid A, C \end{array} \right\}$



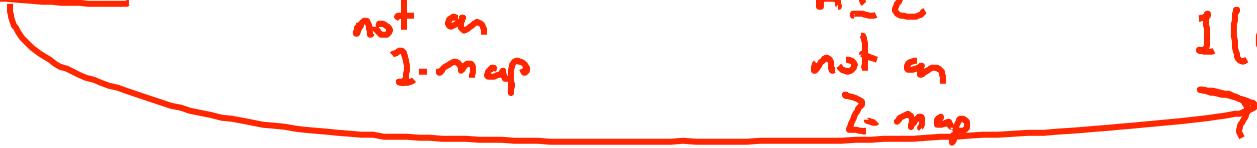
$B \perp D \mid A$   
not an  
I-map



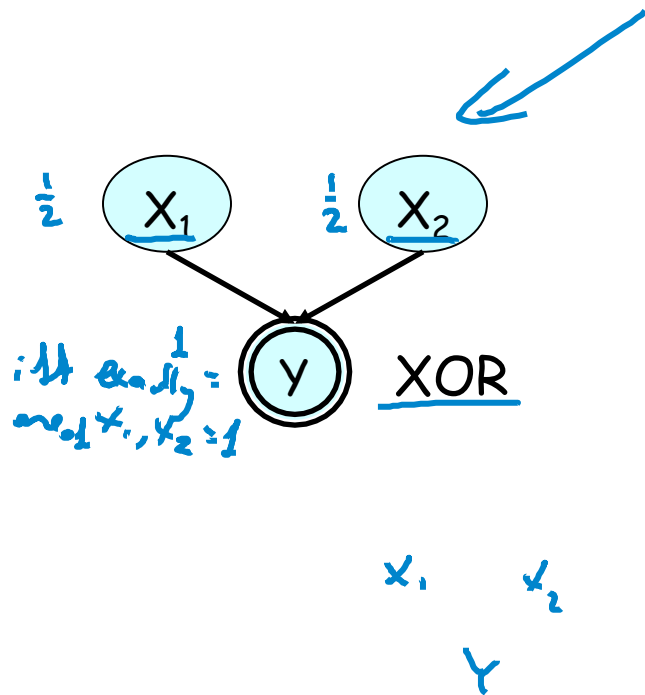
$B \perp D \mid A, C$   
 $A \perp C$   
not an  
I-map



$A \perp C \mid B, D$   
 $I(C) \subseteq I(P)$



# Another imperfect map



$X_1$	$X_2$	<u><math>Y</math></u>	Prob
0	0	0	0.25
0	1	1	0.25
1	0	1	0.25
1	1	0	0.25

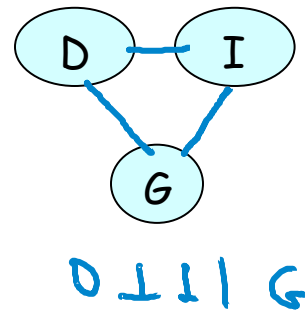
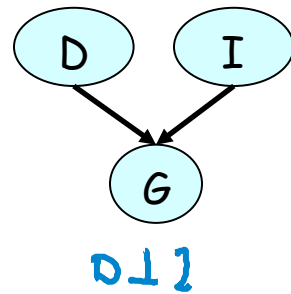
$X_1 \perp X_2$

$X_1 \perp Y$

$X_2 \perp Y$

# MN as a perfect map

- Perfect map:  $I(\underline{H}) = I(P)$ 
  - $H$  perfectly captures independencies in  $P$



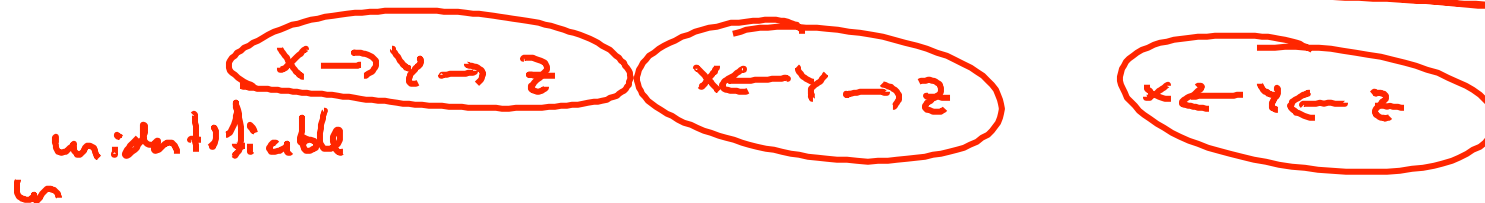


# Uniqueness of Perfect Map

$$\begin{array}{l} G_1 \quad \textcircled{X} \longrightarrow \textcircled{Y} \quad I(G_1) = \emptyset \\ G_2 \quad \textcircled{X} \longleftarrow \textcircled{Y} \quad I(G_2) = \emptyset \end{array} \Rightarrow \text{can represent exactly the same distribution}$$

# I-equivalence

Definition: Two graphs  $G_1$  and  $G_2$  over  $X_1, \dots, X_n$  are I-equivalent if  $I(G_1) = I(G_2)$



Most  $G$ 's have many I-equivalent variants

# Summary

- Graphs that capture more of  $I(P)$  are more compact and provide more insight
- A minimal I-map may fail to capture a lot of structure even if present *and representable as a PGM*
- A perfect map is great, but may not exist
- Converting BNs  $\leftrightarrow$  MNs loses independencies
  - BN to MN: loses independencies in v-structures
  - MN to BN: must add triangulating edges to loops

