

Chapter 10

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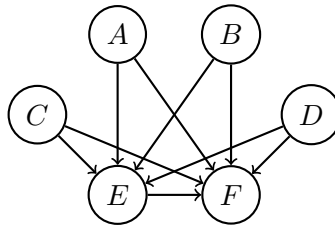
Exercises

Exercise 1

Marginalizing out X gives

$$\begin{aligned} p(A : F) &= \sum_{x'} p(A : F | X = x') \\ &= p(A)p(B)p(C)p(E) \sum_{x'} p(E, F | C, D, X = x') p(X = x' | A : D) \\ &= p(A)p(B)p(C)p(D)p(E, F | A : D) \\ &= p(A)p(B)p(C)p(D)p(F | A : E) p(E | A : D) \end{aligned}$$

It follows that the graph G' is



Note that we can reverse $E \rightarrow F$.

We know that this is a minimal I-map because removing any edges creates conditional independence assumptions not made by G .

Edge removed in G'	New CI in G'	Counterexample in G
$A \rightarrow E$	$A \perp_{G'} E B, C, D$	$A \rightarrow X \rightarrow E$
$A \rightarrow F$	$A \perp_{G'} F B, C, D, E$	$A \rightarrow X \rightarrow F$
$B \rightarrow E$	$B \perp_{G'} E A, C, D$	$B \rightarrow X \rightarrow E$
$B \rightarrow F$	$B \perp_{G'} F A, C, D, E$	$B \rightarrow X \rightarrow F$
$C \rightarrow E$	$C \perp_{G'} E A, B, D$	$C \rightarrow X \rightarrow E$
$C \rightarrow F$	$C \perp_{G'} F A, B, D, E$	$C \rightarrow X \rightarrow F$
$D \rightarrow E$	$D \perp_{G'} E A, B, C$	$D \rightarrow X \rightarrow E$
$D \rightarrow F$	$D \perp_{G'} F A, B, C, E$	$D \rightarrow X \rightarrow F$
$E \rightarrow F$	$E \perp_{G'} F A, B, C, D$	$E \leftarrow X \rightarrow F$

The first row of the above table can be read as: removing the $A \rightarrow E$ edge in G' creates the conditional independence $A \perp_{G'} E | B, C, D$; however, this CI is not held in G because of the $A \rightarrow X \rightarrow E$ path.

Note that $A \perp_{G'} E | B, C, D$ rather than $A \perp_{G'} E | B, C, D, F$ because of Berkson's paradox (this also applies for all above CIs conditioned on three nodes).

This exercise highlights the power of hidden nodes: despite G' has 50% more edges than G despite having fewer nodes.

Exercise 3

Let $Y_i = ch(X_i)$, $B_i = pa(X_i)$, $Z_i = copa(X_i)$ and $U_i = X_{-i} / (Z_i \cup Z_i \cup B_i)$.

$$\begin{aligned}
p(X_i | X_{-i}) &= \frac{p(X_i, X_{-i})}{p(X_{-i})} \\
&\propto p(X_i, X_{-i}) \\
&= p(X_i, B_i, Y_i, Z_i, U_i) \\
&= p(Y_i | X_i, Z_i) p(X_i | B_i) p(B_i, Z_i, U_i) \\
&\propto p(Y_i | X_i, Z_i) p(X_i | B_i) \\
&= p(X_i | B_i) \prod_{y \in Y_i} p(y | pa(y)).
\end{aligned}$$