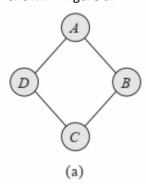
1 Misconception Example:

Consider a scenario where we have four students who get together in pairs to work on the homework for a class. For various reasons, only the following pairs meet: Alice and Bob; Bob and Charles; Charles and Debbie; and Debbie and Alice. (Alice and Charles just can't stand each other, and Bob and Debbie had a relationship that ended badly.) The study pairs are shown in figure a.



Factors for misconception example are given as follows:

$\phi_1(A,B)$	$\phi_2(B,C)$	$\phi_3(C,D)$	$\phi_4(D, A)$
$ \begin{array}{ccccc} a^0 & b^0 & 30 \\ a^0 & b^1 & 5 \\ a^1 & b^0 & 1 \\ a^1 & b^1 & 10 \end{array} $	$\begin{vmatrix} b^0 & c^0 & 100 \\ b^0 & c^1 & 1 \\ b^1 & c^0 & 1 \\ b^1 & c^1 & 100 \end{vmatrix}$	$ \begin{vmatrix} c^0 & d^0 & 1 \\ c^0 & d^1 & 100 \\ c^1 & d^0 & 100 \\ c^1 & d^1 & 1 \end{vmatrix} $	$ \begin{vmatrix} d^0 & a^0 & 100 \\ d^0 & a^1 & 1 \\ d^1 & a^0 & 1 \\ d^1 & a^1 & 100 \end{vmatrix} $
(a)	(b)	(c)	(d)

2 Vehicle Localization Task Example

Consider a vehicle localization task, where a moving car tries to track its current location using the data obtained from a, possibly faulty, sensor. The system state can be encoded (very simply) using the: Location — the car's current location, Velocity — the car's current velocity, Weather — the current weather, Failure — the failure status of the sensor, and Obs — the current observation. We have one such set of variables for every point t. A joint probability distribution over all of these sets defines a probability distribution over trajectories of the car.

Assume selected X = [L, O], where L is the location of the object and O its observed location. At first glance, we might be tempted to make the Markov assumption in this setting: after all, the location at time t+1 does not appear to depend directly on the location at time t-1. However, assuming the object's motion is coherent, the location at time t+1 is not independent of the previous locations given only the location at time t, because the previous locations give us information about the object's direction of motion and speed. By adding Velocity, we make the Markov assumption closer to being satisfied. If, however, the driver is more likely to accelerate and decelerate sharply in certain types of weather (say heavy winds), then our V; L model does not satisfy the Markov assumption relative to V; we can, again, make the model more Markovian by adding the Weather variable. Finally, in many cases, a sensor failure at one point is usually accompanied with a sensor failure at nearby time points, rendering nearby Obs variables correlated.

Q. 2. Wri Q. 3. Cor whi and (0,0) (0,0) Pro Q. 4. Cor tha on 0.5 Ma Q. 5. Wri mis Q. 6. Find	ve that X1 is trivially independent of X3. sider a non-positive distribution P over four binary variables A, B, C, D					
Q. 2. Wri Q. 3. Cor whi and (0,0 (0,0 Pro Q. 4. Cor tha on 0.5 Ma Q. 5. Wri mis Q. 6. Find Q. 7. Cor	ite short note on Markov network for computer vision. Insider a distribution P over four binary random variables X1, X2, X3, X4, ich gives probability 1/8 to each of the following eight configurations, if probability zero to all others: 10,0,0) (1,0,0,0) (1,1,0,0) (1,1,1,0) 10,0,1) (0,0,1,1) (0,1,1,1) (1,1,1,1) 10,0 that X1 is trivially independent of X3. 11 insider a non-positive distribution P over four binary variables A, B, C, D					
Q. 3. Cor whi and (0,0 Cor) (0,0 Cor	nsider a distribution P over four binary random variables X1, X2, X3, X4, ich gives probability 1/8 to each of the following eight configurations, if probability zero to all others: 0,0,0) (1,0,0,0) (1,1,0,0) (1,1,1,0) 0,0,1) (0,0,1,1) (0,1,1,1) (1,1,1,1) ve that X1 is trivially independent of X3. Insider a non-positive distribution P over four binary variables A, B, C, D					
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Q. 4. Cor that on the correction of the correcti	0,0,1) (0,0,1,1) (0,1,1,1) (1,1,1,1) we that X1 is trivially independent of X3. Insider a non-positive distribution P over four binary variables A, B, C, D					
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Q. 4. Cor tha on 0.5 Ma Q. 5. Wri mis Q. 6. Find Q. 7. Cor	nsider a non-positive distribution P over four binary variables A, B, C, D	(0,0,0,1) (0,0,1,1) (0,1,1,1) (1,1,1,1)				
Q. 5. Wri		Prove that X1 is trivially independent of X3.				
0.5 Ma Q. 5. Wri mis Q. 6. Find Q. 7. Cor	Consider a non-positive distribution P over four binary variables A, B, C, D that assigns nonzero probability only to cases where all four variables take					
0.5 Ma Q. 5. Wri mis Q. 6. Find Q. 7. Cor						
Q. 5. Wri	on exactly the same value; for example, we might have P (a1, b1, c1, d1) =					
Q. 5. Wri mis Q. 6. Find Q. 7. Cor	0.5 and P (a0, b0, c0, d0) = 0:5. Show that Markov network satisfies the					
Q. 6. Find Q. 7. Cor	Markov blanket condition for all nodes.					
Q. 6. Find Q. 7. Cor	Write the logarithmic representation of the clique potential parameters for					
Q. 7. Cor	misconception example given above.					
	Find Canonical energy function for the Misconception example.					
ma	Consider the Markov network structure H of figure a. Construct minimal I-map Bayesian networks for a nonchordal Markov network in Figure a.					
	A					
	(B) (C)					
	Ť Ť					
	$\stackrel{\leftarrow}{\mathcal{D}}$					
	(F)					
	(a)					
Q. 8. Dra	Draw DBN for monitoring Vehicle Localization Task given in 2.					
Q. 9. Hov	w monitoring system for vehicle conclude that sensors have failed?					
	Write short note on HMMs and Phylo-HMMs for Gene Finding.					
Q. 11. Exp	Explain State Observation model.					
Q. 12. The	The HMM is shown in figure below:					
	0.3 0.5 0.1					
	0.6					
	0.9					
Wri	ite distribution for above HMM,					
Q. 13. Wri	Write short note on HMMs for Speech Recognition.					
Q. 14. Wri	Write short note on Linear Dynamical System.					
Q. 15. Dra	ite short note on Linear Dynamical System.	Draw single plate model for student example.				