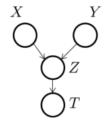
IFT6269-A2020	Hwk 3	Name:
Prof: Simon Lacoste-Julien	Due date: Nov 3, 2020	Student id:

Note: There are 10 bonus points in this assignment. The maximum number of points achievable for this assignment is 80; though you will be graded over 70. If you get more points than 70, they can compensate other assignments.

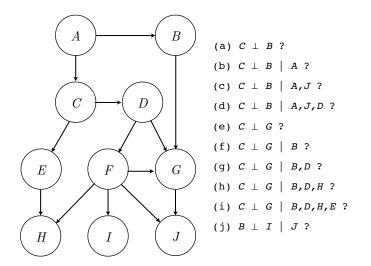
1. DGM (5 points)

Consider the directed graphical model G on the right. Write down the implied factorization for any joint distribution $p \in \mathcal{L}(G)$. Is it true that $X \perp \!\!\!\perp Y \mid T$ for any $p \in \mathcal{L}(G)$? Prove or disprove.



2. D-separation in DGM (5 points)

Indicate (yes or no) which conditional independence statements are true?



3. Positive interactions in-V-structure (10 points)

Let X, Y, Z be binary random variables with a joint distribution parametrized according to the graph: $X \to Z \leftarrow Y$. We define the following:

$$\alpha := P(X = 1), \quad \beta := P(X = 1 \mid Z = 1), \quad \gamma := P(X = 1 \mid Z = 1, Y = 1)$$

- (a) For all the following cases, provide examples of joint probability tables (and calculate the quantities a, b, c), so that each of the following conditions are (individually) true:
 - (i) $\gamma < \alpha$ (ii) $\alpha < \gamma < \beta$ (iii) $\beta < \alpha < \gamma$.
- (b) Think of X and Y as causes and Z as a common effect, for the previous three cases, briefly state (in a sentence or two) why the claims are true for your examples.

Please use the following format to represent the joint distributions you choose for the cases in part *a*). Study the expected format carefully and note how this is sufficient to represent the joint distribution. Also, use the corresponding boxes to write the values of α , β and γ for each of the three sub-questions.

X	Y	P(X,Y)	X	Y	P(Z=1 X,Y)
1	1		1	1	
1	0		1	0	
0	1		0	1	
0	0		0	0	

α	β	γ

4. Equivalence of directed tree DGM with undirected tree UGM (10 points)

Let G be a directed tree and G' its corresponding undirected tree, i.e., the orientation of edges is ignored. Prove that $\mathcal{L}(G) = \mathcal{L}(G')$. *Hint: Recall that by the definition of a directed tree, G does not contain any v-structure.*

5. Hammersley-Clifford counter-example (10 points)

In class, we mentioned that the strict positivity of the joint distribution was crucial in the Hammersley-Clifford theorem. Here is a counter-example (4.4 in Koller & Friedman) that shows the problems when we have zero probabilities.

Consider a joint distribution p over four binary random variables: X_1, X_2, X_3 and X_4 which gives probability $\frac{1}{8}$ to each of the following eight configurations, and zero to all others:

(0,0,0,0) (1,0,0,0) (1,1,0,0) (1,1,1,0) (0,0,0,1) (0,0,1,1) (0,1,1,1) (1,1,1,1)

Let G be the usual four nodes undirected graph $X_1 - X_2 - X_3 - X_4 - X_1$. One can show that p satisfies the global Markov property with respect to this graph G because of trivial deterministic relationships. For example, if we condition on $X_2 = 0$ and $X_4 = 0$, then the only value of X_3 with non-zero probability is $X_3 = 0$, and thus $X_3|(X_2 = 0, X_4 = 0)$ being a deterministic random variable, it is trivially conditionally independent to X_1 . By (painfully) going through all other possibilities, we get similar situations: $X_2 = 0$ and $X_4 = 1$ forces $X_1 = 0$, etc.

Prove that the distribution p cannot factorize according to G, and thus $p \notin \mathcal{L}(G)$. *Hint:* argue by contradiction.

6. Bizarre conditional independence properties (10 BONUS points)

Let (X, Y, Z) be a random vector with a finite sample space. Consider the following statement:

"If $X \perp \!\!\!\perp Y \mid Z$ and $X \perp \!\!\!\perp Y$ then $(X \perp \!\!\!\perp Z \text{ or } Y \perp \!\!\!\perp Z)$."

- (a) Is this true if one assumes that Z is a binary variable? Prove or disprove.
- (b) Is the statement true in general? Prove or disprove.

7. EM and Gaussian mixtures (4 points)

Derive the form of the **M-step updates** for the parameters $\{\pi_k, \mu_k, \sigma_k\}_{k=1}^K$ of a Gaussian mixture model in which the covariance matrices are proportional to the identity:

$$p(x) = \sum_{k=1}^{K} \pi_k \, \mathcal{N}(x \,|\, \boldsymbol{\mu}_k, \sigma_k \mathbf{I})$$

8. Implementation: EM and Gaussian mixtures (26 points)

Follow the instructions in this Colab notebook:

https://colab.research.google.com/drive/1mtyFbcWZAuoNV8zBicoH1Y0Bc_Sv2c7G