

**UNC
CS Department**

**COMP 755– HW 1
Machine Learning**

Instructor: Prof. Junier Oliva

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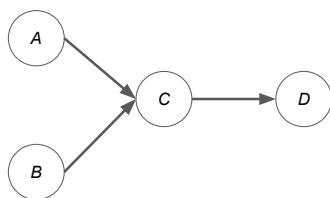
This assignment contains 3 pages (including this cover page) and 1 multi-part questions. The total of points is 100. **The assignment is due September 1st, 2021 11:59PM via https://docs.google.com/forms/d/e/1FAIpQLSekixyPflufkMEh3jXgBza22BRJwKP0cmICZnuVFG4jTXwbQQ/viewform?usp=pp_url.** For any handwritten portions, please submit clear and legible handwritten or typefaced answers. **Do not try to search directly for the answers. You may speak to other students about the assignment at a high-level (e.g. sharing related references/slides). However, Sharing complete or partial answers is strictly prohibited.** Please see the UNC honor code: <https://catalog.unc.edu/policies-procedures/honor-code/> for additional details on maintaining academic integrity.

Distribution of Marks

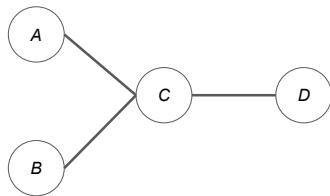
Question	Points	Score
1	100	
Total:	100	

1. Graphical Models

- (a) (25 points) Suppose that we have 4 features A, B, C, D . What are all the possible conditional (and unconditional) independence statements to make between the features? (E.g. $C \perp\!\!\!\perp D, \dots$)
- (b) (25 points) Which of all the possible independence statements (in (a)) are guaranteed to hold for the Bayesian network below?



- (c) (25 points) Which of all the possible independence statements (in (a)) are guaranteed to hold for the undirected graphical model below? Why is this not a valid graphical model for distributions satisfying the Bayesian network above?



- (d) (25 points) What is the smallest edit to the undirected graphical model above, such that it is a valid graphical model for distributions satisfying the Bayesian network (in (b)). What are the independence statements are guaranteed to hold for the edited undirected graphical model?

1(a) $A \perp C | B$, if B is observed and

(either) $(A \leftarrow B \rightarrow C)$ or $(A \rightarrow B \rightarrow C)$ or
SST resulted $(C \rightarrow B \rightarrow A)$ [cascade or common parent
situation]

$A \not\perp C$ if B is unobserved

another case, if $C \rightarrow B \rightarrow A$ and B is
unobserved, then $A \not\perp C$

$A \not\perp C$ if $A \perp C | B$ if B is observed.

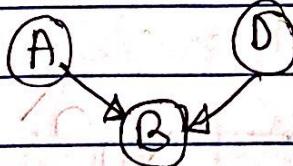
Similarly,

$A \perp D | B$ if B is observed where

B is common parent of A, D or
A, B, D forming cascade situation

$A \not\perp D$ if B is unobserved

for V-structure



$A \perp D$ if B is unobserved

$A \not\perp D | B$ if B is observed

Similarly for the rest of the Conditional Independence we can write; A

$A \perp B | C$ if C is observed \square Cascade or
 $A \not\perp B | C$ if C is unobserved \square Common Parent

$A \perp B$ if C is unobserved \square V-Structure

$A \not\perp B | C$ if C is observed \square A $\perp \!\!\! \perp$ B

$A \perp D | C$ if C is observed \square Cascade or
 $A \not\perp D | C$ if C is unobserved \square Common Parent

$A \perp D$ if C is unobserved \square V-Structure

$A \not\perp D | C$ if C is observed \square A $\perp \!\!\! \perp$ D

$A \perp B | D$ if D is observed \square Cascade or
 $A \not\perp B | D$ if D is unobserved \square Common Parent

$A \perp B$ if D is unobserved \square V-Structure

$A \not\perp B | D$ if D is observed \square A $\perp \!\!\! \perp$ D

$A \perp C | D$, if D is observed \square Cascade or

$A \not\perp C | D$, if D is unobserved \square Common Parent

1) In case of A $\perp\!\!\!\perp$ C, iff D is unobserved (IV-structure)
A $\not\perp\!\!\!\perp$ C | D iff D is observed (IV-structure)

In fact Cascade or Common Parent situation

the following Conditional Independence Statements hold:

B $\perp\!\!\!\perp$ C | A, if A is observed QLA

B $\perp\!\!\!\perp$ D | A, if A is observed QLA

B $\perp\!\!\!\perp$ A | C, if C is observed QLA

B $\perp\!\!\!\perp$ D | C, if C is observed QLA

B $\perp\!\!\!\perp$ A | D, if D is observed QLA

B $\perp\!\!\!\perp$ C | D, if D is observed QLA

C $\perp\!\!\!\perp$ B | A, if A is observed QLA

C $\perp\!\!\!\perp$ D | A, if A is observed QLA

C $\perp\!\!\!\perp$ D | B, if B is observed QLA

C $\perp\!\!\!\perp$ A | B, if B is observed QLA

C $\perp\!\!\!\perp$ A | D, if D is observed QLA

C $\perp\!\!\!\perp$ B | D, if D is observed QLA

$D \perp B | A$, if A is observed

$D \perp C | A$, if A is observed

$D \perp A | B$, if B is observed

$D \perp C | B$, if B is observed

$D \perp A | C$, if C is observed

$D \perp B | C$, if C is observed

The following Conditional Independence statements hold for the NA structure
case:

$B \perp D | C$, if C is unobserved

$B \perp A | C$, if C is unobserved

$B \not\perp A | C$, if C is observed

$B \perp D | C$, if C is unobserved

$B \not\perp D | C$, if C is observed

$B \perp C | A$, if A is unobserved

$B \not\perp C | A$, if A is observed

$B \perp D | A$, if A is unobserved

$B \not\perp D | A$, if A is observed

$B \perp A$, if D is unobserved \square

$B \not\perp A | D$, if D is observed \square

$B \perp C$, if D is unobserved \square

$B \not\perp C | D$, if D is observed \square

$C \perp B$, if A is unobserved \square

$C \not\perp B | A$, if A is observed \square

$C \perp D$, if A is unobserved \square

$C \not\perp D | A$, if A is observed \square

$C \perp D$, if B is unobserved

$C \not\perp D | B$, if B is observed \square

$C \perp A$, if B is unobserved

$C \not\perp A | B$, if B is observed \square

$C \perp A$, if D is unobserved

$C \not\perp A | D$, if D is observed \square

$C \perp B$, if D is unobserved

$C \not\perp B | D$, if D is observed \square

$D \perp B$, if A is unobserved

$\cancel{D \perp B | A}$, if A is observed

$D \perp C$, if A is unobserved

$\cancel{D \perp C | A}$, if A is observed

$D \perp A$, if B is unobserved

$\cancel{D \perp A | B}$, if B is observed

$D \perp C$, if B is unobserved

$\cancel{D \perp C | B}$, if B is observed

$D \perp A$, if C is unobserved

$\cancel{D \perp A | C}$, if C is observed

$D \perp B$, if C is unobserved

$\cancel{D \perp B | C}$, if C is observed

1(b) if C is observed, then we can say that

$$A \perp D | C$$

(A)

(B)

(D)

: C holds all the

information, that is

needed to determine the outcome of D.

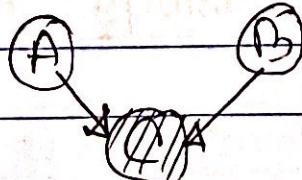
knowing A does not make any change
in the outcome of D.

Similarly, we can say,

$$B \perp D | C, \text{ if } C \text{ is observed}$$

Again, if C is observed, we can see

$$A \not\perp B | C,$$

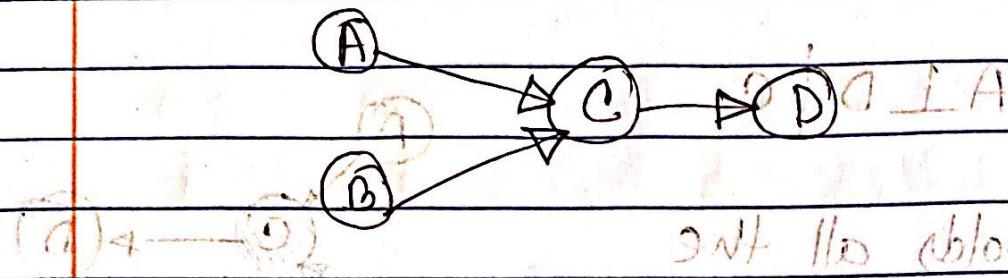


Hence, knowing C

V Structure (explaining away)

Couples A and B

If C is not observed, then we see
 $A \perp D$, if C is observed



$A \perp D$, if C is not observed

$B \perp D$, if C is not observed

D is independent of A and B given C

Both situations are examples of Cascade

parents: two parents for each "A" parent

But in case of A, B, C

$A \perp B$, if C is unobserved

this is an example

having do $C \rightarrow D \rightarrow B$

of V-structure.

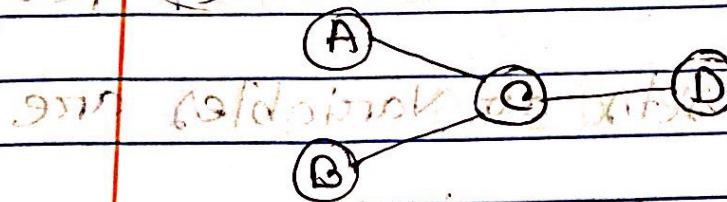


Here, C is a parent of both A and B

and C is a child of both A and B

$B \perp A$ given C

↳ random fields or pairwise

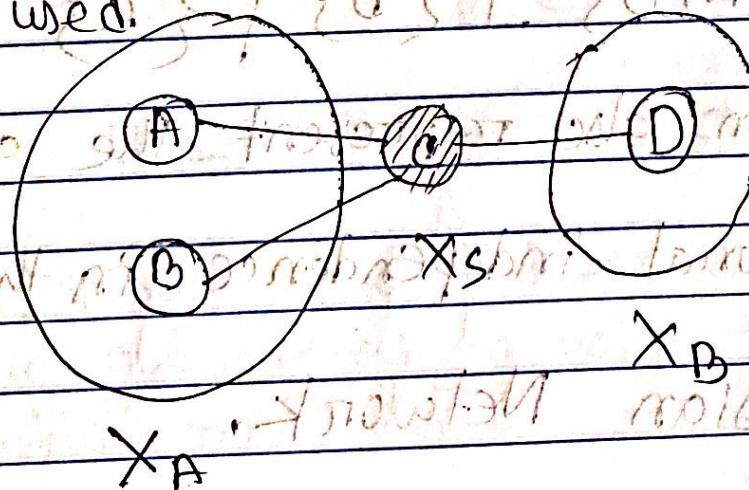


The above one is an undirected

graphical model. To capture Conditional Independence in such an undirected

graphical model "Markov Random Fields"

is used.



Hence $X_S = \{C\}$, X_S is a Separating Subset

$X_A = \{A, B\}$, Sub-set X_A contains A and B

$X_B = \{D\}$, Sub-set X_B contains only D

According to Global Markov property,

any two subsets of variables are

Conditionally independent given a

Separating subset.

So, we can say, given a

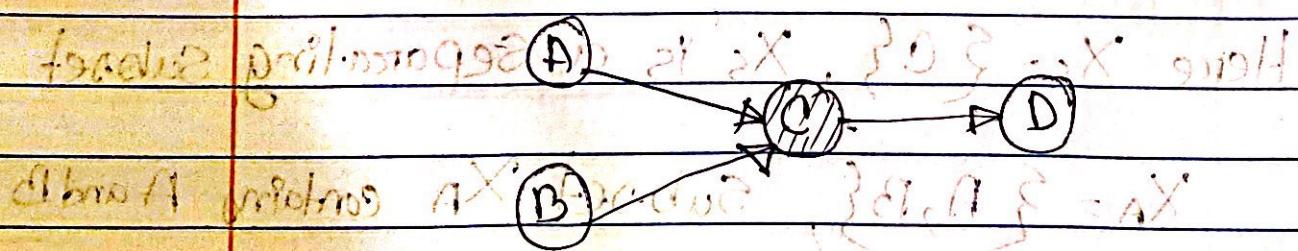
subset $X_A \cup X_B \cup X_S$ disjoint from

$\{A, B\} \cup \{D\} \perp \{C\}$

We can also represent the above

Conditional independence in the following

Bayesian Network.



If node C is observed, then $A \rightarrow C \rightarrow D$

forming a Cascade Pattern, So, we can

say, $A \perp D \mid C$, if C is observed

Again, $B \perp D \mid C$ forming a cascade

pattern. So, we can say, $B \perp D \mid C$

$B \perp D \mid C$, if C is observed.

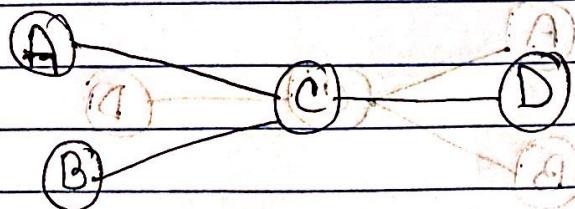
Now, we know that, the following two conditional independence statements hold for the UGM.

$A \perp D \mid C$, if C is observed

$B \perp D \mid C$, if C is observed

The above two independence statements are guaranteed to hold for the undirected graphical model below.

Because

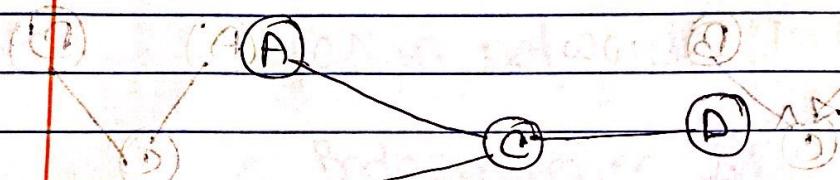


for the above model

Q1(c) Question: (Part 2 of 1(c)) Why is this

graph not a valid graphical model, for

distributions satisfying the Bayesian network above?



bill (B) is too Bayesian if QLA

constraint between

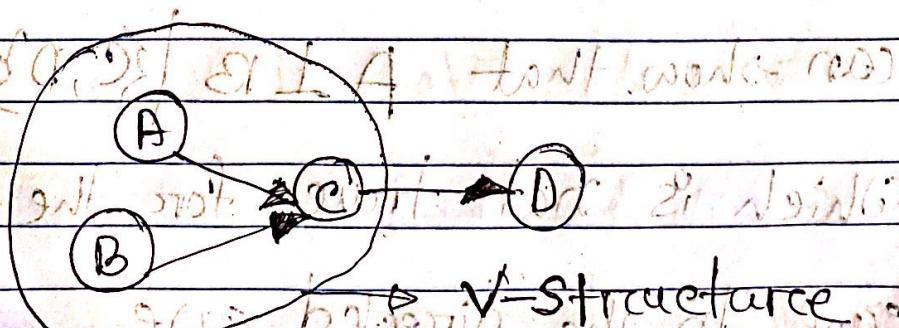
between not mil-

ment. If answer

of constraints.

The above Bayesian network has a

N-structure component marked below

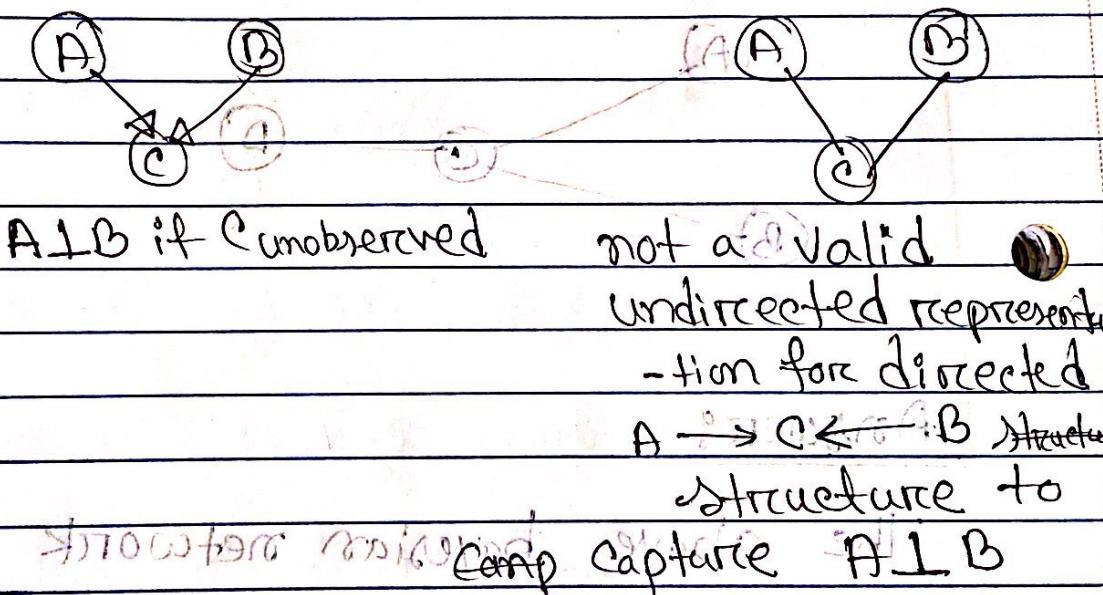


broadening? If yes QLA

We know that, for a directed structure

case, there is no undirected model

that can describe the independence assumption



In the given undirected model, we

can show that $A \perp B \mid \{C, D\}$

which is not true for the directed

case. In the directed case

$A \perp B$ only if C is unobserved.

Ques 1(d) Question: What is the smallest edit to
the undirected graphical model above,
join such that it is a valid graphical
model for distributions satisfying
the Bayesian network (in (b)). What
are the independence statements are
guaranteed to hold for the edited
undirected graphical model?

Answer: 6.000000 0.111111

One of the possible solutions that I can
think of is that adding an
edge between node A and node B.

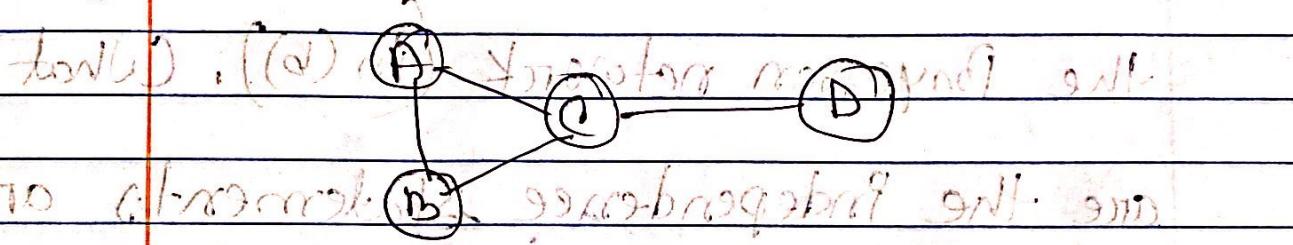
That way, we can eliminate the independence

of fibs ~~itself~~ an assumption $A \perp B \mid \{C, D\}$ [using

Markov property]

So, the resulting undirected graphical

model is



Since $C \perp D \mid \{A, B\}$ [Markov property]

but there is no way we can show

that $A \perp B$ holds for any

specific scenarios.

∴ The only guaranteed independence

statements are $A \perp D \mid \{B, C\}$

$\{A, B\} \perp \{D\} \mid \{C\}$ [using

Markov property]

$A \perp D | \{B, C\}$

[using Pairwise

$B \perp D | \{A, C\}$

Markov Property]

So, the I-map of the edited undirected graphical model is the subset of the given Bayesian network

$$I(G') \subset I(G_{BN})$$

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