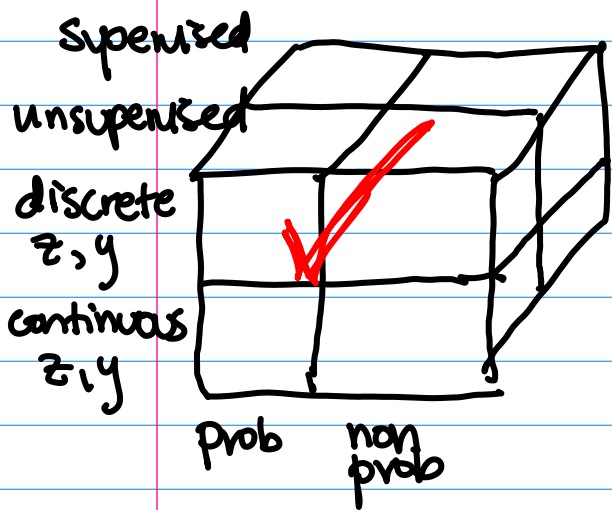


CS181 - Models with Structure (Ch.8)



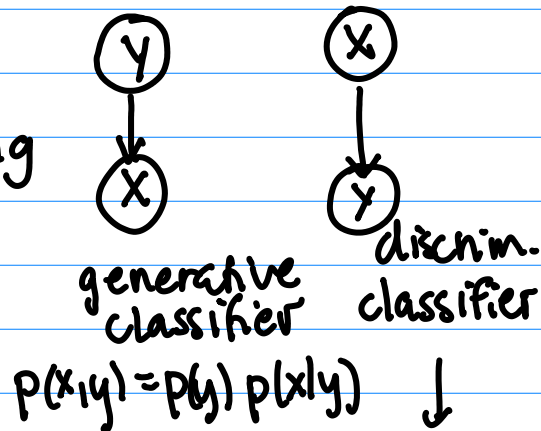
+ model selection,
model classes, CNN)
objectives (SUM)

+ structured models,
decision-making (RL)

Notes • HW5 due Friday
• Practical teams due Friday,
(and look at sample code!)

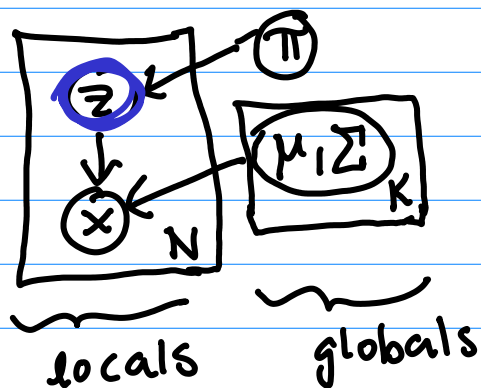
Graphical Models

→ Early: Supervised Learning

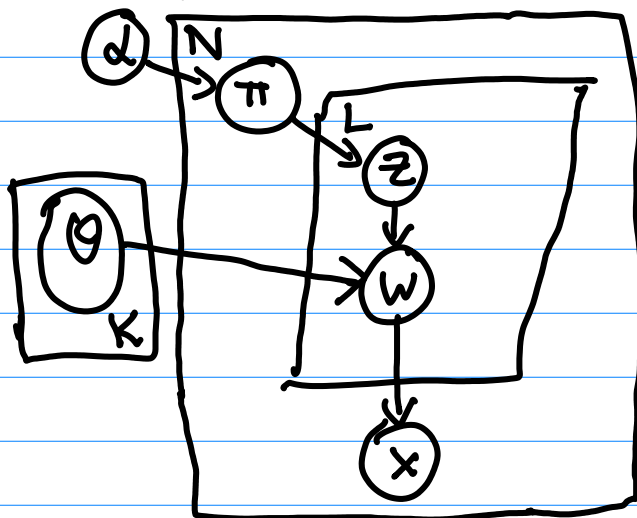


→ More recently:

Mixture models:



Topic Models:



$$p(x, y) = \frac{p(x)p(y)}{p(y|x)}$$

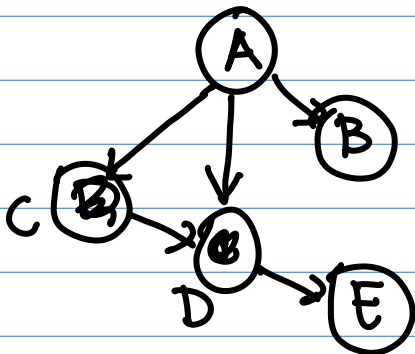
These are examples of Bayesian Networks

→ help encode structure of the data
(including independence relationships)

→ independences are useful for:

- ① inference (block coordinate ascent)
- ② learn smaller models

Today: focus Directed Acyclic Graph (DAG)

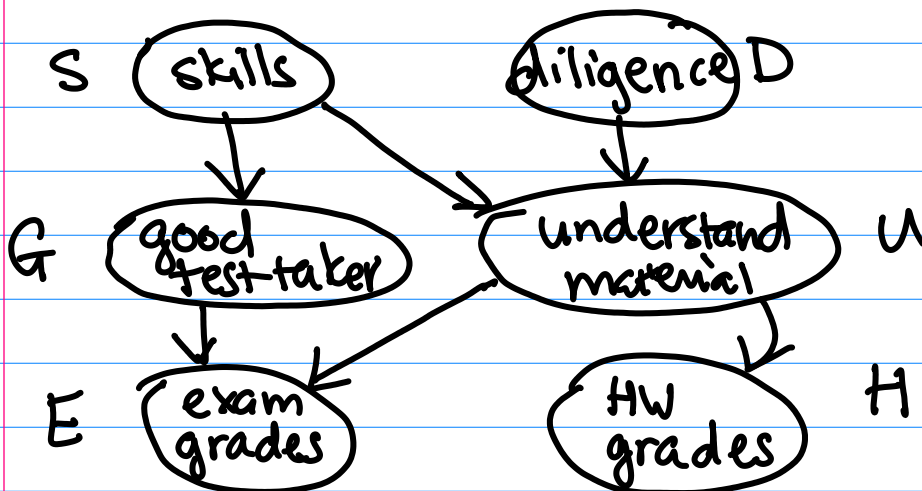


→ interpretation is that we can simplify the joint prob in a specific way:

$$\begin{aligned}
 p(A, B, C, D, E) &= p(A) p(B|A) p(C|B, A) p(D|A, B, C) p(E|C, D) \\
 &= p(A) p(B|A) p(C|A) p(D|C, A) p(E|D)
 \end{aligned}$$

⊗

Local independence: every node is conditionally indep of non-descendants given parents



→ given G, U :
 E is indep of S, D, H
 → given S ,
 are G, U indep?
 → given E ,
 are G, U indep?

Formalize into rules: "D-separation"

(A), (B) are d-separated if every undirected path from (A) to (B) is blocked.

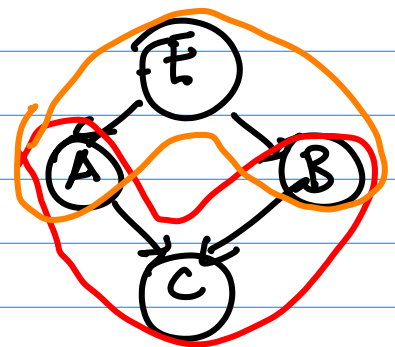
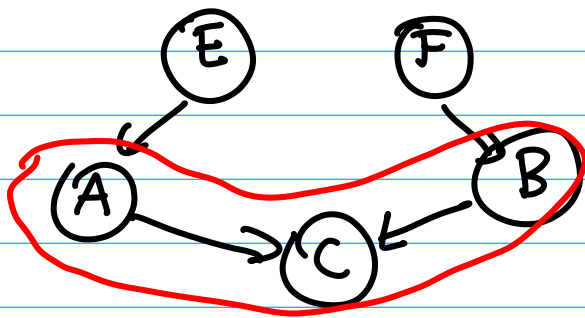
Ways to block:

1. $(A) \rightarrow (C) \rightarrow (B)$, C is observed

2. $(A) \leftarrow (C) \leftarrow (B)$, C is observed

3. $(A) \leftarrow (C) \rightarrow (B)$, C is observed

4. $(A) \rightarrow (C) \leftarrow (B)$, C is NOT observed



Quick note: about uniqueness:

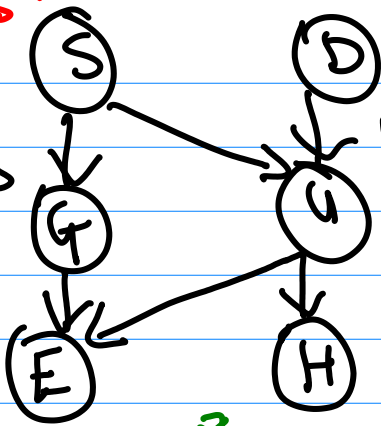
- if we have a causal interpretation $(A) \rightarrow (B)$ then A causes B.

- BUT: in general, statistical interpretation allows for multiple orderings/graphs (but one may convey the fewest params / most independences)

$$P(A, B) = P(A) P(B|A) = P(B) P(A|B)$$

P(S) is true

minimal ordering most independencies



$U | S=0, D=0$
 $U | S=0, D=1$
 $U | S=1, D=0$
 $U | S=1, D=1$

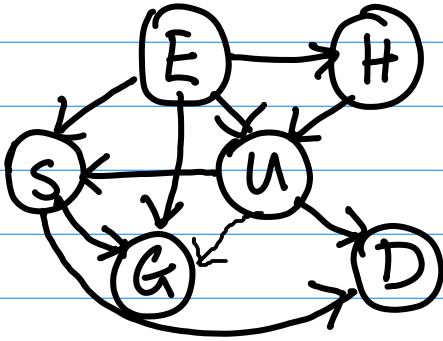
Count the # of params (if we assume all nodes are binary):

S: 1 D: 1 E: 4
 G: 2 U: 4 H: 2

Total: 14 params

$P(S)$ $P(D|S)$ $P(G|S, D)$ $P(U|G, S, D)$ $P(E|G, U)$ $P(H|U)$

What if I drew the graph differently?



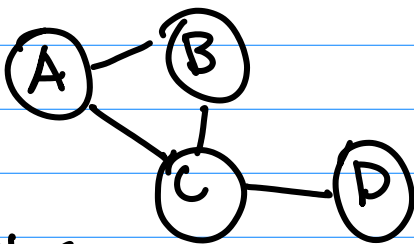
This is a graph with a diff topological ordering, minimal # of connections given that ordering.

count params: 23 params (more than 14)

$P(E)$ $P(H|E)$ $P(U|E, H)$...

Final notes on other graphical models:

Undirected

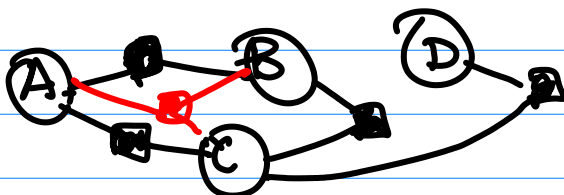


$$P(A, B, C, D) =$$

$$\phi(A, B, C) \phi(C, D)$$

tricky: normalize??

Factor Graphs



$$P(A, B, C, D) = \phi(A, C) \phi(A, B) \phi(B, C) \phi(C, D)$$