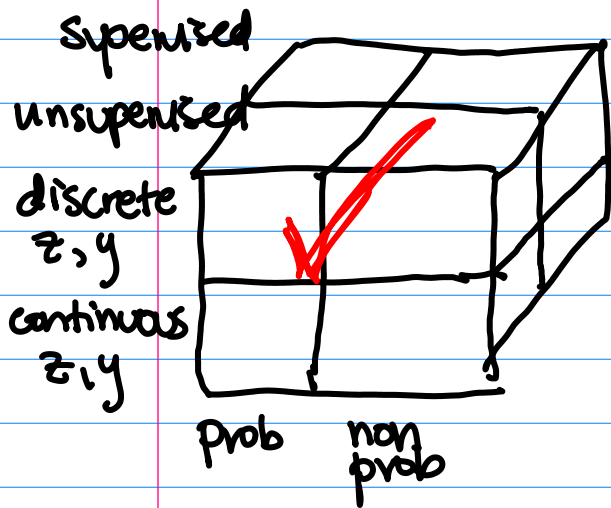


CS181 - Models with Structure (Ch.8)



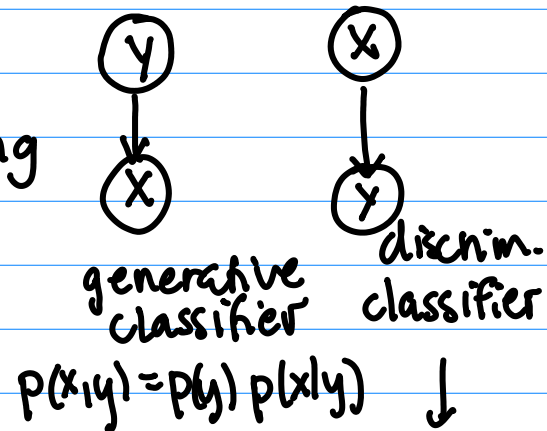
+ model selection,
model classes, CNN
objectives (SUM)

+ structured models,
decision-making (RL)

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

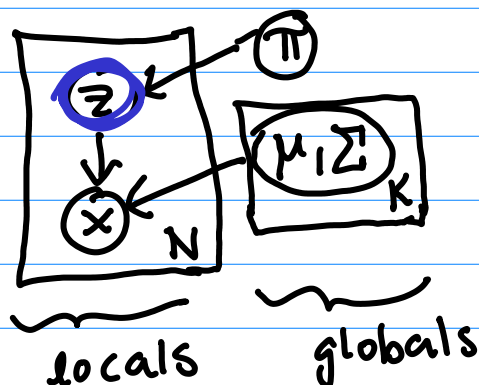
Graphical Models

→ Early: Supervised Learning

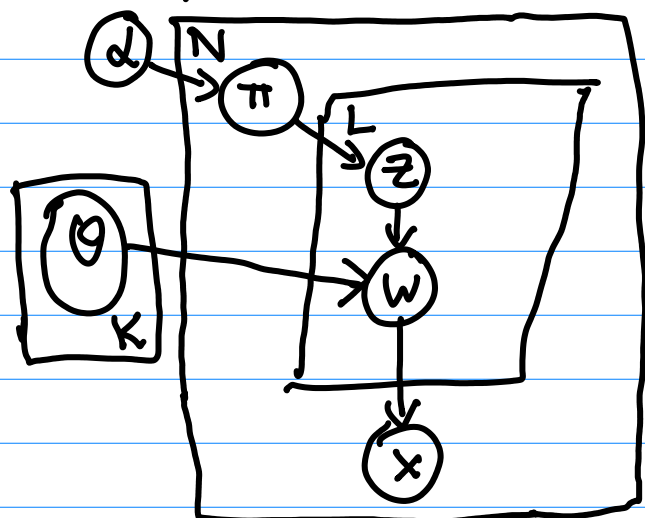


→ More recently:

Mixture models:



Topic Models:



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

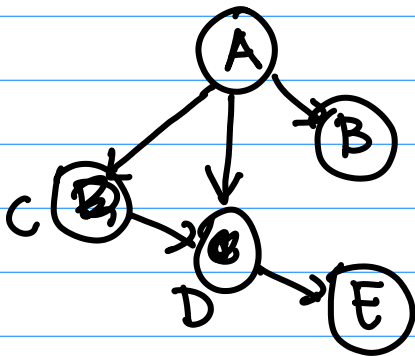
These are examples of Bayesian Networks

→ 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)



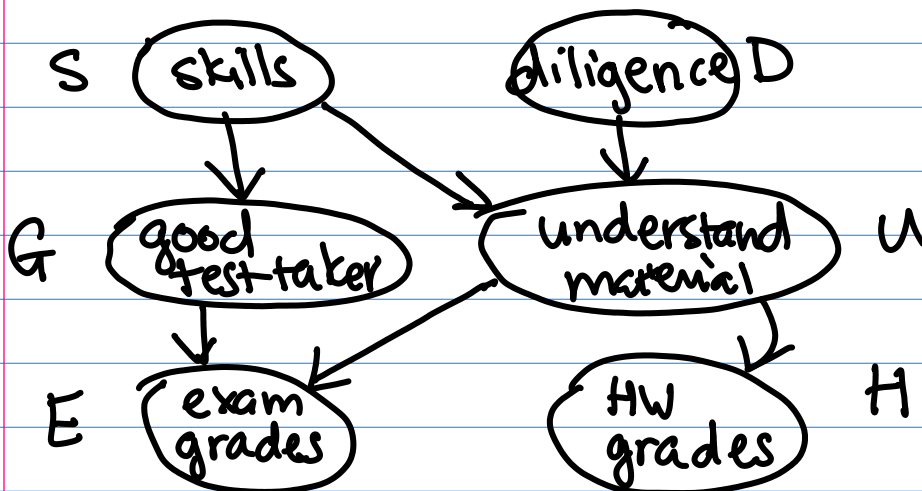
→ interpretation is that we can simplify the joint prob in a specific way:

$$p(A, B, C, D, E) = p(A) p(B|A) p(C|B, A) p(D|A, B, C) p(E|D)$$

$$= p(A) p(B|A) p(C|A) p(D|C, A) p(E|D)$$

⊗

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?

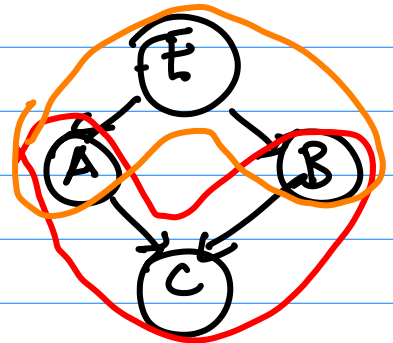
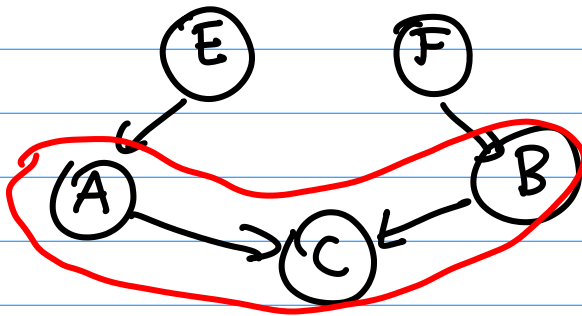
→ given E ,
are G, U indep?

Formalize into rules: "D-separation"

①, ② are d-separated if every undirected path from ① to ② 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



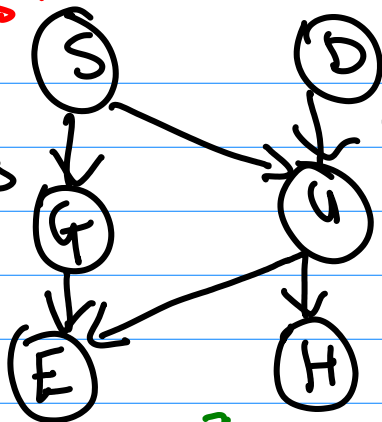
Quick note: about uniqueness:

- 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



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

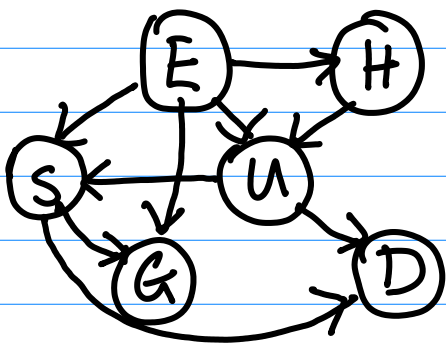
Count the # of params (if we assume all nodes are binary):

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

Total: 14 params

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

What if I drew the graph differently?



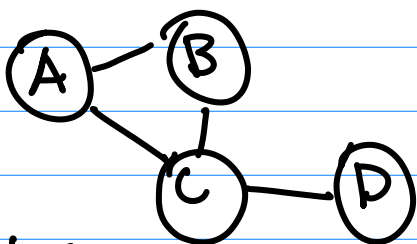
This is a graph with a diff topological ordering, minimal # of connections given that ordering.

can't params: 19 params (more than 14)

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

Final notes on other graphical models:

Undirected

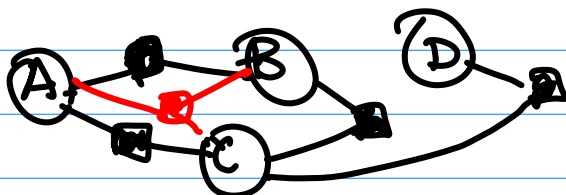


$P(A, B, C, D) =$

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

tricky: normalize??

Factor Graphs



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