

today: • graph theory
• DGM

graphical model

graph. model $\leadsto$ prob theory + C.S.
 $\downarrow$ $\downarrow$
 R.V. graph

graph $\rightarrow$ efficient data structure

e.g. $X_1, X_2, \dots, X_n$ R.V.

$X_i \in \{0, 1\}$ $n=100s$

$\Rightarrow 2^{100}$ #'s table
 $\leadsto$ intractable

QMR



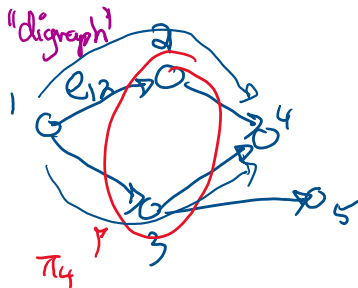
graph theory review:

directed graph $G = (V, E)$

$V = \{1, \dots, n\}$ "nodes/vertices"

$E \subseteq V \times V$ "directed edges"

$e_{12} = (1, 2)$



$\pi_i \triangleq \{j \in V : \exists (j, i) \in E\}$

set of parents of i

$children(i) \triangleq \{j \in V : \exists (i, j) \in E\}$

directed path $1 \leadsto 4$

compatible seq. of edges

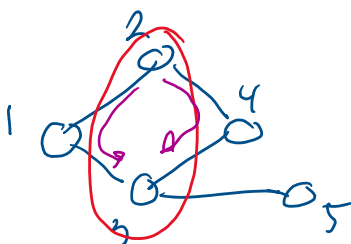
$(1, 2), (2, 4)$

or $(1, 3), (3, 4)$

undirected graph : $G = (V, E)$ where elements of E are 2-sets
 (set of 2 different elements of V)

thus we have $\{i, j\} = \{j, i\}$

vs. $(i, j) \neq (j, i)$ [order matters]



"neighbors" of node 4 or 1

$N(i)$ for neighbors of i
 $\triangleq \{j \in V : \exists \{i, j\} \in E\}$

undirected path $2 \leftrightarrow 3$

⊗ neighbors replace the parent/children terminology for digraphs

def: DAG = directed acyclic graph $\triangleq$ digraph with no cycles

def: an ordering $\pi: V \rightarrow \{1, 2, \dots, |V|\}$ is said to be topological for digraph G iff nodes in π_i appear before i in $\pi \forall i$



$$(j \in \pi_i \Rightarrow \pi(j) < \pi(i))$$

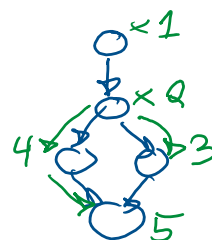
$\Rightarrow$ top. ordering $\Rightarrow$ all edges from left to right
[no "back edge"]

prop: digraph G is a DAG $\iff \exists$ a topological ordering of G

proof: $\Leftarrow$ trivial: no back edge $\Rightarrow$ no cycle

$\Rightarrow$ use DFS algorithm to label nodes in decreasing order when have no children

to construct a top. sort.
in $O(|E| + |V|)$



top. sort $\rightarrow$ finding a topological ordering

notation for graphical model

n discrete R.V. $X_1, \dots, X_n$

$\rightarrow$ discrete R.V. for simplicity

cond. dist. concept is freely to formulate for cts. R.V.

(see Boltzmann-Kolmogorov paradox)

V $\Leftarrow$ set of vertices
one R.V. per node

joint $p(X_1=x_1, X_2=x_2, \dots, X_n=x_n) \stackrel{\text{shorthand}}{=} p(x_1, \dots, x_n) \quad \mathcal{X} = \mathcal{X}_{1:n}$
 $= p(x_V) \stackrel{(\dagger)}{=} p(x)$

question: is $p(x_1, x_2, x_4) \stackrel{?}{=} p(x_2, x_1, x_4)$?

Yes? usually in this class we "fixed convention"

for any $A \subseteq V$

$$p(x_A) = P\{X_i = x_i : i \in A\} = \sum_{x_{A^c}} p(x_A, x_{A^c})$$

$\hookrightarrow$ subset of "subscripts"
e.g. $p(x_1, x_2, x_4) = p(x_{\{1, 2, 4\}})$

summing over all possible values of x_{A^c}
i.e. $(x_i)_{i \in A^c}$

$$A^c \triangleq V \setminus A$$

revisit cond. independence

Let $A, B, C \subseteq V$

~~⑧~~ $X_A \perp X_B \mid X_C$

(F) $\Leftrightarrow p(x_A, x_B | x_C) = p(x_A | x_C) p(x_B | x_C) \quad \forall x_A, x_B, x_C$
 st. $p(x_C) > 0$
 Fackungssatz

(c) conditionung $\Leftrightarrow p(x_A | x_B, x_C) = p(x_A | x_C) \quad \forall x_A, x_B, x_C$
 " " " " $\forall x_B, x_C \text{ s.t. } p(x_B, x_C) > 0$

⑤ "marginal independence": $X_A \perp\!\!\!\perp X_B \left(\begin{smallmatrix} I \\ \bigcirc \end{smallmatrix} \right)$
 $p(x_A, x_B) = p(x_A) p(x_B)$

2 facts about card. indgs

1) can repeat variable in statement
(for convenience)

$X \perp\!\!\!\perp Y \mid Z, W$
 $X \perp\!\!\!\perp Y, \underline{Z} \mid \underline{Z}, W$ is fine to say
 does not do anything

2) decomposition: $x \perp (y, z) | w \Rightarrow \begin{cases} x \perp y | w \\ x \perp z | w \end{cases}$

⊗ pairwise indep. $\nRightarrow$ mutual indep. $\rightarrow$ see lecture 3
 $Z = X \oplus Y$
 see

⊛ chain rule $p(x_V) = \prod_{i=1}^n p(x_i | x_{1:i-1})$ (always true) (last c

last cond.: $p(x_n | x_{1:n-1})$
table with 2^n entries

assumption in Q64

$$p(x_v) = \prod_{i=1}^v p(x_i | x_{\pi_i}) \rightarrow \text{nodes of } \mathcal{Z}^{\max |\pi_i| + 1}$$

$$x_i \perp\!\!\!\perp x_{1:i-1} \mid x_{\pi_i}$$

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directed graph. model

let $G = (V, E)$ be a DAG

a directed graphical model (DGM) (associated with G) (aka a Bayesian network)

is a family of distribution over X_V

$$\mathcal{I}(G) = \left\{ p \text{ is a dist. over } X_V : \exists \text{ legal factors } f_i \text{'s s.t.} \right. \\ \left. p(x_V) = \prod_{i=1}^n f_i(x_i | x_{\pi_i}) \quad \forall x \right\}$$

over X_V define $p(x) = \prod_{i=1}^n f_i(x_i | x_{\pi_i}) \quad \forall x$

st. $f_i : \Omega_{X_i} \times \Omega_{X_{\pi_i}} \rightarrow [0,1]$

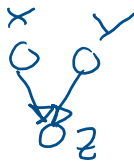
st. $\sum_{x_i} f_i(x_i | x_{\pi_i}) = 1 \quad \forall x_{\pi_i}$

$\Rightarrow f_i$ is like a CPT
(conditional prob. table)

terminology: if we can write $p(x) = \prod_i f_i(x_i | x_{\pi_i}) \rightarrow$ this defines G

then we say that " p factorizes according to G " $\S p \in \mathcal{J}(G)$

one example



$p \in \mathcal{J}(G) \Leftrightarrow \exists f_x, f_y, f_z$ s.t.

$p(x, y, z) = f_x(x) f_y(y) f_z(z | x, y)$

"leaf plucking" property (fundamental property of DGM)

Let $p \in \mathcal{J}(G)$; let n be a leaf in G (i.e. n has no children)

a) then $p(x_{1:n-1}) \in \mathcal{J}(G - \hat{n})$

$\nwarrow$ means remove node n from G

b) moreover, if $p(x_{1:n}) = \prod_{i=1}^n f_i(x_i | x_{\pi_i})$ then $p(x_{1:n-1}) = \prod_{i=1}^{n-1} f_i(x_i | x_{\pi_i})$

proof: $p(x_n, x_{1:n-1}) = f_n(x_n | x_{\pi_n}) \prod_{i=1}^{n-1} f_i(x_i | x_{\pi_i})$

$p(x_{1:n-1}) = \sum_{x_n} p(x_n, x_{1:n-1}) = \left[\sum_{x_n} f_n(x_n | x_{\pi_n}) \right] \left(\prod_{i=1}^{n-1} f_i(x_i | x_{\pi_i}) \right)$

1 by def'n

proposition: if $p \in \mathcal{J}(G)$

let $\{f_i\}$ be a factorization w.r. to G of p

then $\forall i, p(x_i | x_{\pi_i}) = f_i(x_i | x_{\pi_i}) \quad \forall x_{\pi_i}$

i.e. factors are correct conditionals of p
(and are unique)

proof: WLOG, let $(1, \dots, n)$ be a tp. ordering of G (possible since G is a DAG)

• let i be fixed; we want to show that $p(x_i | x_{\pi_i}) = f_i(x_i | x_{\pi_i})$

• n is a leaf $\rightarrow$ "pluck it" to get $p(x_{1:n-1}) = \prod_{i=1}^{n-1} f_i(x_i | x_{\pi_i})$

• we're stuck, we want to show that $p(x_i | x_{\pi_i}) = f_i(x_i | x_{\pi_i})$

• n is a leaf $\rightarrow$ "pluck it" to get $p(x_{1:n-1}) = \prod_{j=1}^{n-1} f_j(x_j | x_{\pi_j})$

factorization for a DAG where x_{n-1} is now a leaf

• pluck $n-1$ as well? (as leaf, because of top. sort.)

• keep doing this [in reverse order $n-1, n-2, \dots, i+1$] to get

$$p(x_{1:i}) = f_i(x_i | x_{\pi_i}) \prod_{j=1}^{i-1} f_j(x_j | x_{\pi_j})$$

$\Rightarrow g(x_{1:i-1})$ ✓ i.e. there is no x_p by top. sort. property

partition $1:i$ as $\{i\} \cup \pi_i \cup A$ [$A \triangleq 1:i-1 \setminus \pi_i$]

$$p(x_{1:i}) = p(x_i, x_{\pi_i}, x_A)$$

$$\frac{p(x_i | x_{\pi_i}) = \sum_{x_A} p(x_i, x_{\pi_i}, x_A)}{\sum_{x_i} \sum_{x_A} p(x_i, x_{\pi_i}, x_A)} = \frac{f_i(x_i | x_{\pi_i}) \sum_{x_A} g(x_{1:i-1})}{\left(\sum_{x_i} f_i(x_i | x_{\pi_i}) \right) \sum_{x_A} g(x_{1:i-1})} = f_i(x_i | x_{\pi_i})$$

time is no x_i ? there

⊛ (now), we can safely write

$$\mathcal{G}(G) = \{p : p(x_v) = \prod_{i=1}^n p(x_i | x_{\pi_i})\}$$

properties of DGM:

adding edges $\Rightarrow$ more distributions

i.e. $G=(V,E)$ and $G'=(V,E')$ with $E' \supseteq E$
then $\mathcal{G}(G) \subseteq \mathcal{G}(G')$

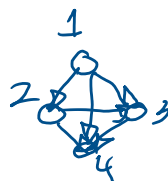
examples:

• $E=\emptyset$
(trivial graph)

$\Rightarrow \mathcal{G}(G)$ contains only fully independent distributions
i.e. $p(x_v) = \prod_{i=1}^n p(x_i)$

• complete DAG

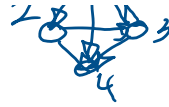
for all $i \neq j$ either $(i,j) \in E$
or $(j,i) \in E$



get wlog

$$p(x_v) = \prod_{i=1}^n p(x_i | x_{1:i-1})$$

for all $i \neq j$ either $(i,j) \in E$
or $(j,i) \in E$



$$p(x_V) = \prod_{i=1}^n p(x_i | x_{1:i-1})$$

(like chain rule)

$\Rightarrow$ all dist. on x_V
are in $\mathcal{G}(\text{complete})$

② removing edges impose more factorization constraints
(and thus more cond. indep. assumptions)

"absence of edge" is meaningful on a DGM