

# Basics of probability

(Axioms, joint, conditional, marginal, independence, Bayes' rule, expectation)

## PGM

### Representation

BN (DAG)

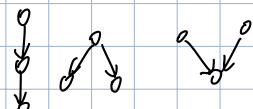
↓ Moralize

$$P(X_1, \dots, X_N) = \prod_{n=1}^N P(X_n | \text{Par}(X_n))$$

$X \rightarrow Y$  if  $X$  is a parent of  $Y$

Dependencies / Independencies

"v-structure", explaining away



[induction proofs not req]

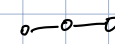
MN (undirected)

$$P(X_1, \dots, X_N) = \frac{1}{Z} \prod_{k=1}^K \phi_k \quad \phi: \text{factor/potential/score}$$

MN:  $X_i - X_j$  if they occur in one factor

Factor graph: Bipartite graph of nodes and factors

Dependencies / independencies



→ CRF: conditional version of MRF  $P(Y|X)$

### Inference (exact inference) (undirected)

Sum-product, max-product

(Commutative semiring not req)

Variable elimination: To eliminate a node  $X$

Apply evidence potential  
consider all the factors involving  $X$   
sum-prod/max-prod  
result is a factor  $\psi(N(X))$

Message Passing (Tree of variables, factors, junction tree not req)

- 2 pass to answer multiple queries about a node
- $i \xrightarrow{\text{pass}} j$  if  $i$  has received all messages except from  $j$

$$M_{ij} = \sum_{\text{variables to eliminate}} \phi_{\text{local}} \prod_{k \in N(i) \setminus j} m_{k \rightarrow i}$$

### Learning (Fully observable BN)

MLE decomposes to each node conditioned on the values of its parents

Categorical distribution

MLE:  $\frac{1}{\pi}^{(MLE)} = \arg\max_{\pi} P(\mathcal{D} | \pi) \quad \frac{1}{\pi_k}^{(MLE)} = \frac{N_k}{N}$

MAP:  $\frac{1}{\pi}^{(MAP)} = \arg\max_{\pi} P(\pi) P(\mathcal{D} | \pi)$   
 $P(\pi) = \text{Dir}(\pi; \alpha)$

add-( $\alpha-1$ )  
smoothing

Bayesian

$$P(X | \mathcal{D}) = \int P(X | \pi) P(\pi | \mathcal{D}) d\pi$$

add- $\alpha$   
smoothing

learning  $\alpha$ : MLE, MAP, Bayesian-  
(hierarchical Bayes)

(Memorizing formulas of  $N$ , Dir is not req.)

(Computing  $\int$  is not req unless taught in lec)

Emphasis: small derivations on standard knowledge. Unimportant: { memorization, tricky derivation steps }