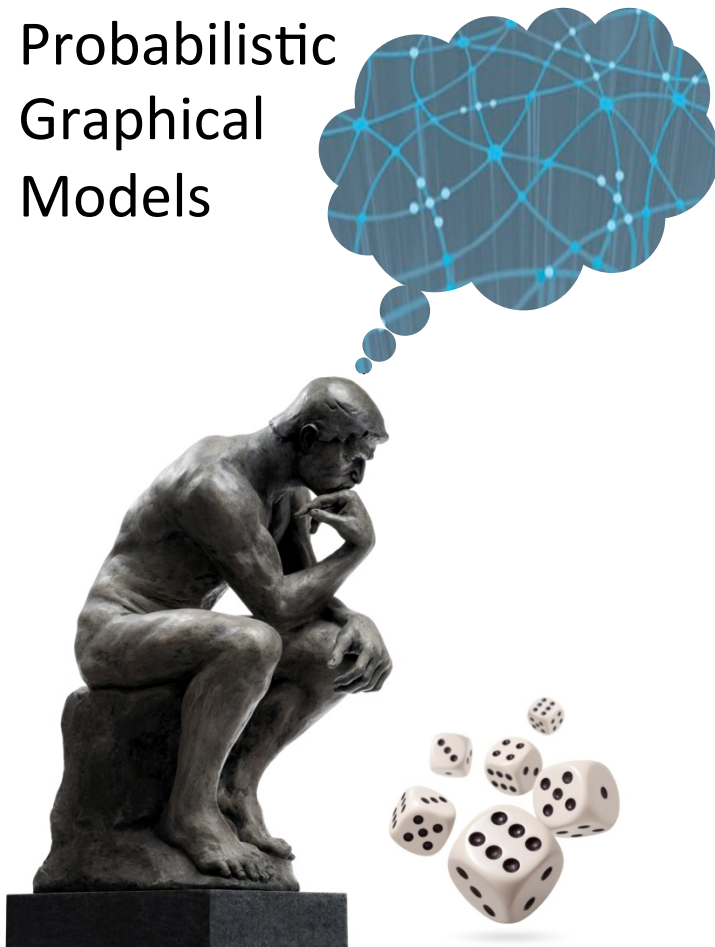


Probabilistic
Graphical
Models



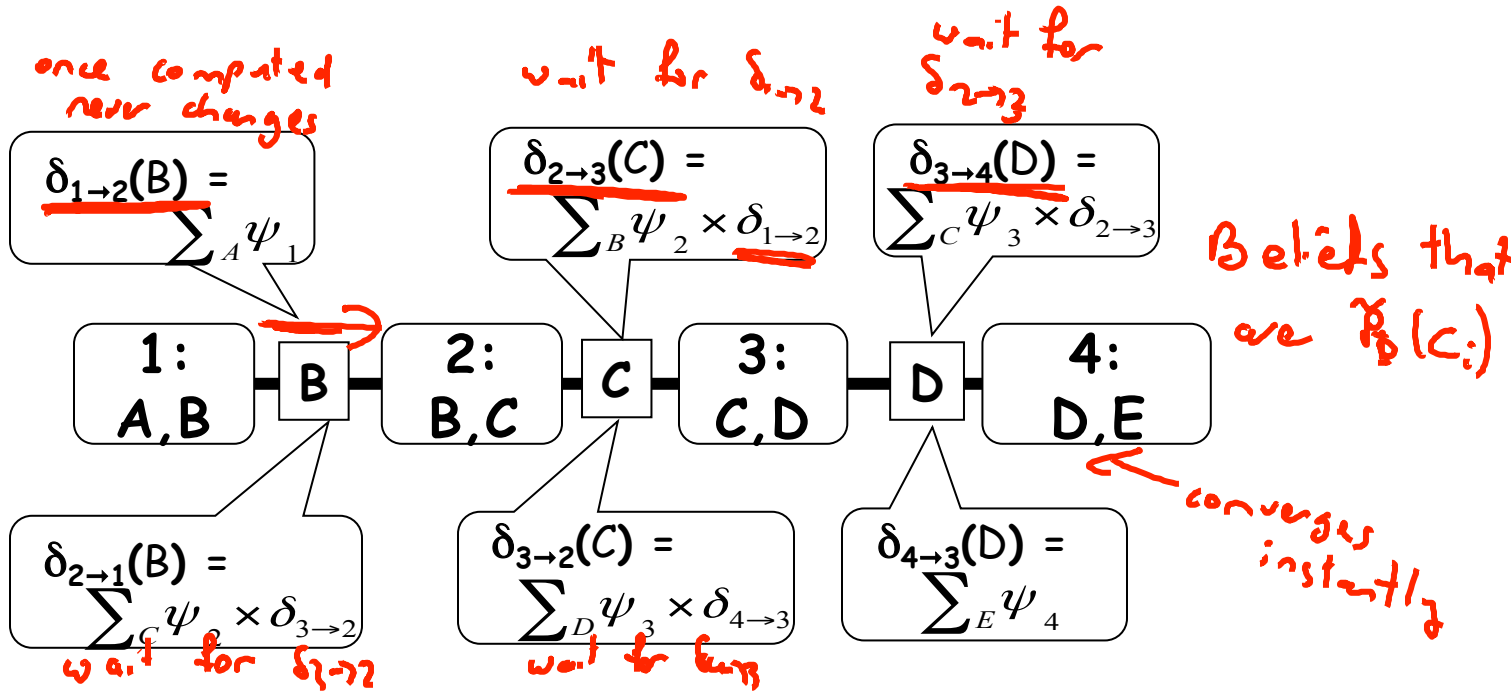
Inference

Message Passing

Clique Tree
Algorithm:
Computation

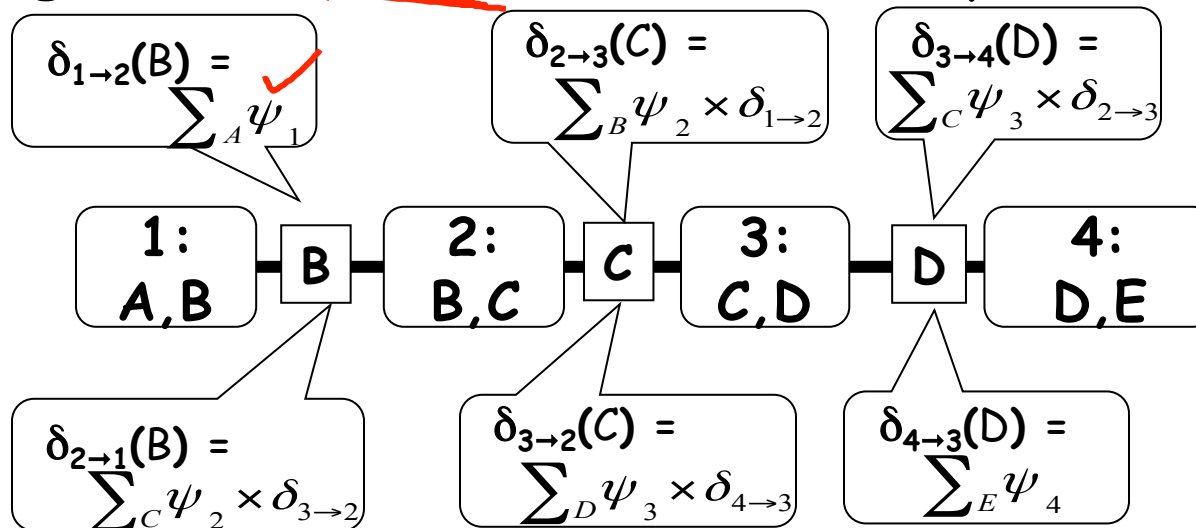
Message Passing in Trees

(Chains: forward-backward alg.)



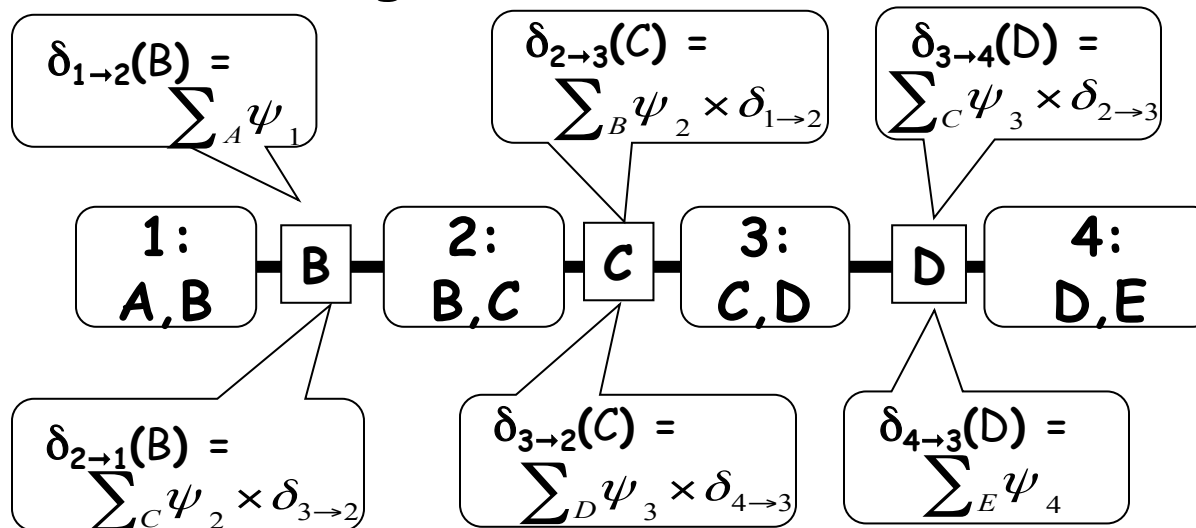
Convergence of Message Passing

- Once C_i receives a final message from all neighbors except C_j , then $\delta_{i \rightarrow j}$ is also final (will never change)
- Messages from leaves are immediately final



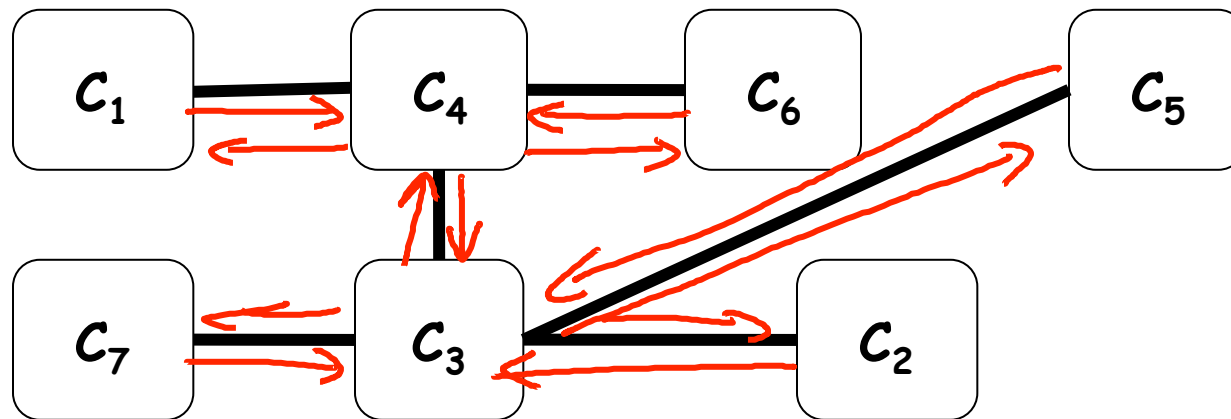
Convergence of Message Passing

- Can pass messages from leaves inward
- If messages are passed in the right order, only need to pass $2(K-1)$ messages

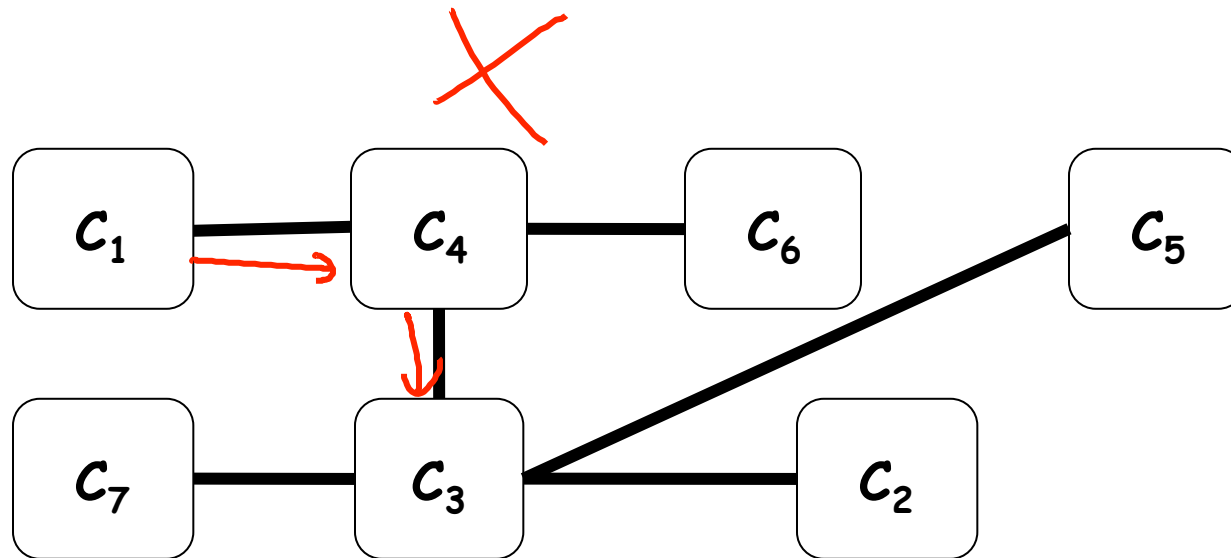


K total number of clusters.
 \Downarrow
 $k-1$ edges
 \Downarrow
 $2(k-1)$ msgs

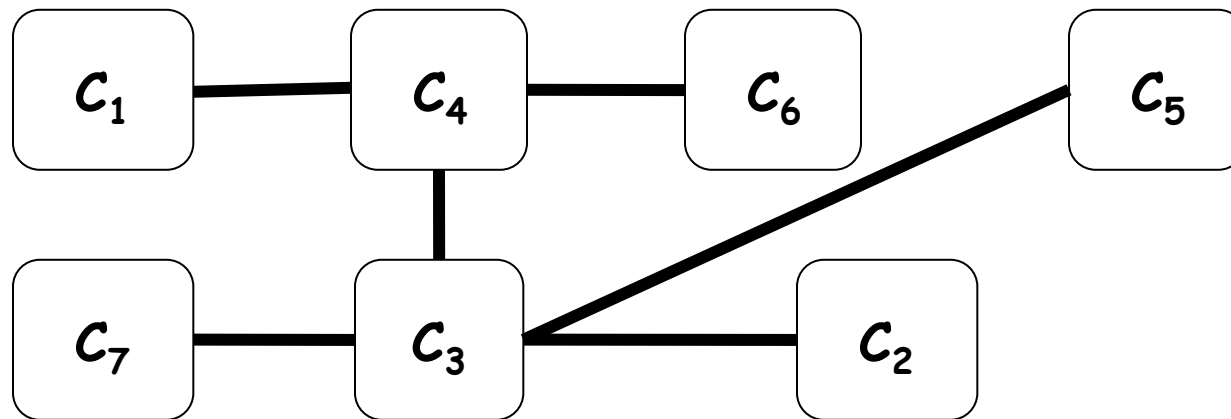
Message Passing Order I



Message Passing Order II



Message Passing Order III

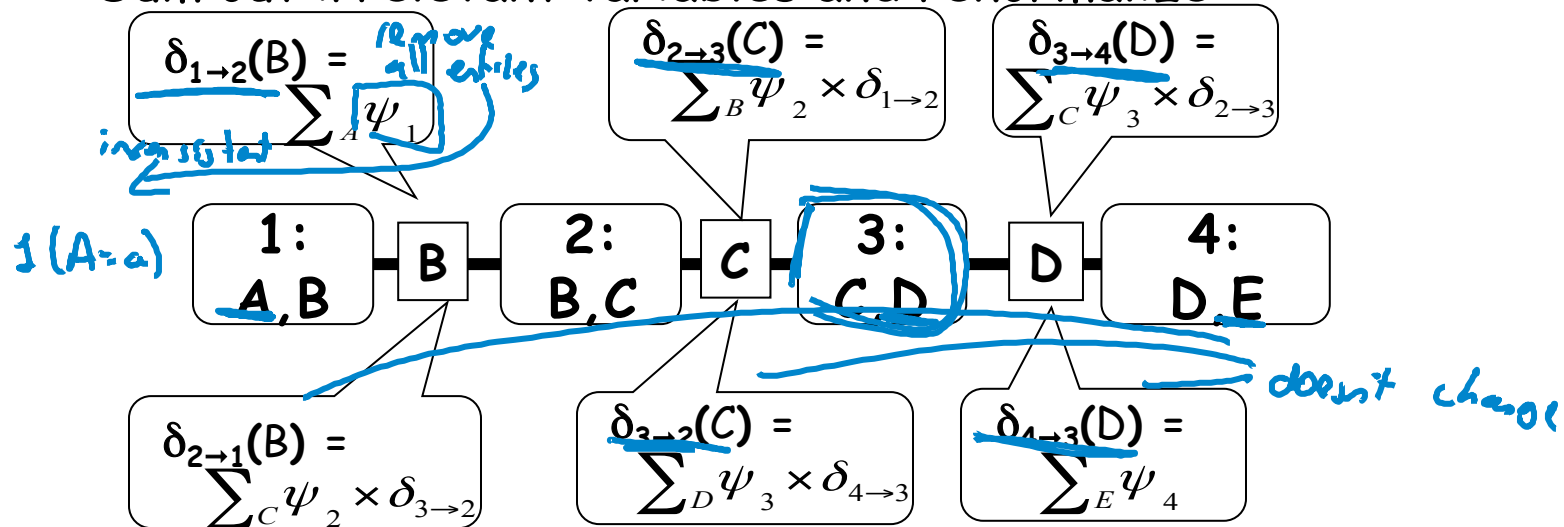


Answering Queries

- Posterior distribution queries on variables that appear together in clique
 - Sum out irrelevant variables from any clique containing those variables \mathcal{P}_ϕ renormalize
- Introducing new evidence $Z=z$ and querying X
 - If X appears in clique with Z Incremental inference
 - Multiply clique that contains X and Z with indicator function $1(Z=z)$ $\mathcal{P}_\phi(z, X)$ reduce clique $\mathcal{P}_\phi(z, X)$
 - Sum out irrelevant variables and renormalize $\mathcal{P}_\phi(z, X)$

And More Queries

- Introducing new evidence $Z=z$ and querying X if X does not share a clique with Z
 - Multiply $1(Z=z)$ into some clique containing Z *reduction of factor*
 - Propagate messages along path to clique containing X
 - Sum out irrelevant variables and renormalize



Summary

- In clique tree with K cliques, if messages are passed starting at leaves, $2(K-1)$ messages suffice to compute all beliefs
- Can compute marginals over all variables at only twice the cost of variable elimination
- By storing messages, inference can be reused in incremental queries