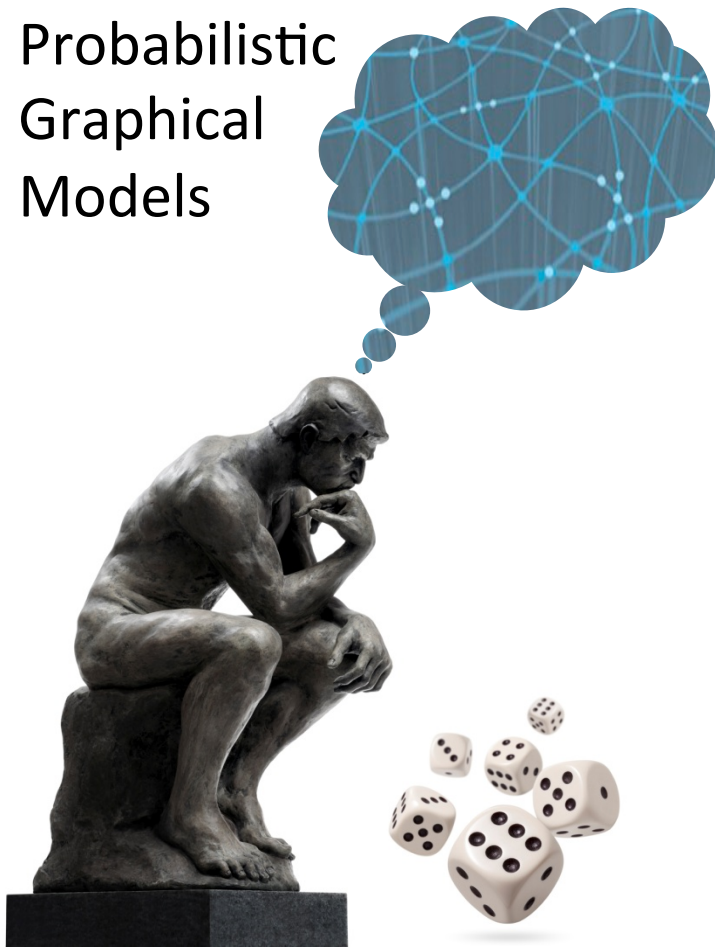


Probabilistic  
Graphical  
Models



Inference

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Message Passing

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Clique Tree &  
Independence

# RIP and Independence

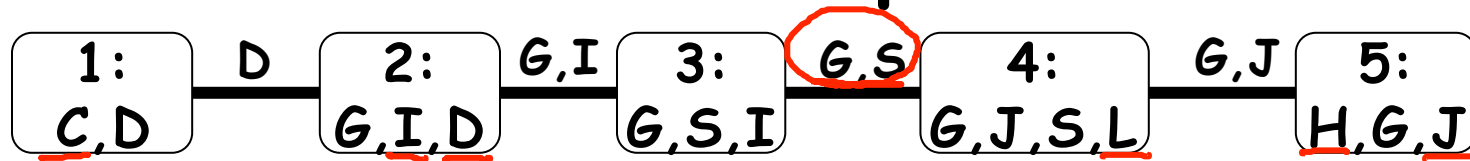
- For an edge  $(i,j)$  in  $T$ , let:



- $\mathbf{W}_{\langle(i,j)}$  = all variables that appear only on  $C_i$  side of  $T$
- $\mathbf{W}_{\langle(j,i)}$  = all variables that appear only on  $C_j$  side
- Variables on both sides are in the sepset  $S_{i,j}$

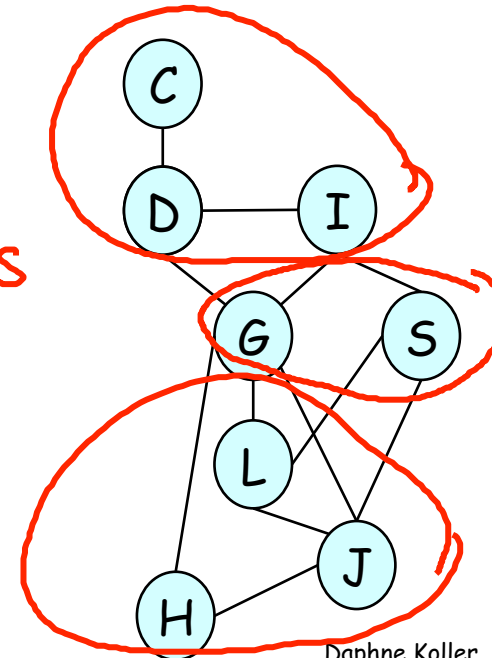
- **Theorem:**  $T$  satisfies RIP if and only if, for every  $(i,j)$   $P_{\Phi} \models (\mathbf{W}_{\langle(i,j)} \perp \mathbf{W}_{\langle(j,i)} \mid \mathbf{S}_{i,j})$

# RIP and Independence



$$P_{\Phi} \models (\{C, I, D\} \perp \{J, L, H\} \mid \{G, S\})$$

*C, I, D separated from H, L, J given G, S*

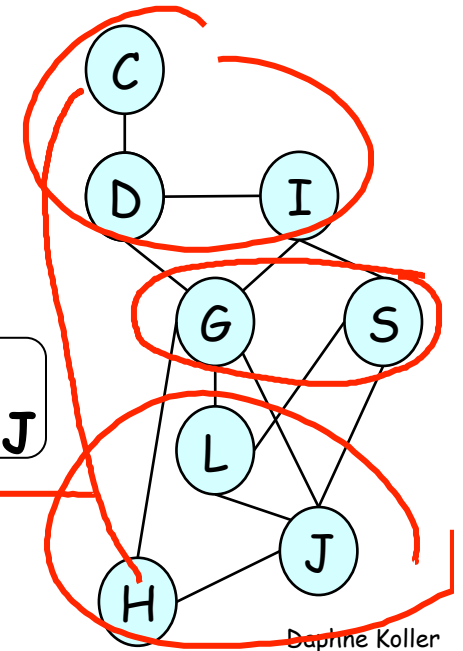


# RIP and Independence

- Theorem:**  $T$  satisfies RIP if and only if, for every edge  $(i,j)$   $P_{\Phi} \models (W_{\langle(i,j)} \perp W_{\langle(j,i)} \mid S_{i,j})$

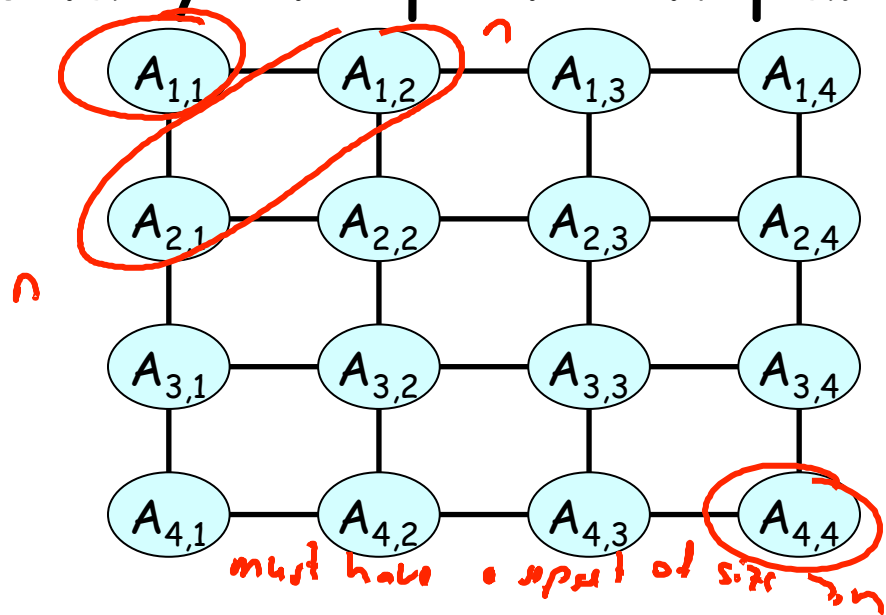
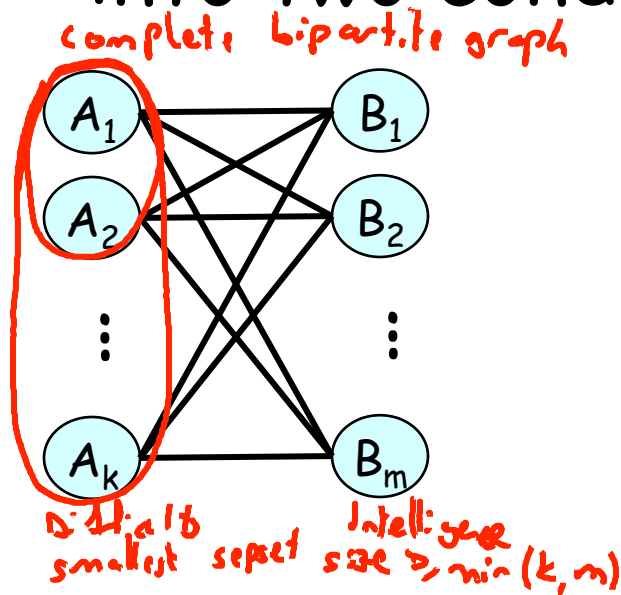
Assume otherwise  $\Rightarrow \exists$  path in induced Markov network between  $W_{\langle(i,j)}$  that goes through  $S_{i,j}$   $W_{\langle(j,i)}$  doesn't

Factor  $\phi(C, H)$



# Implications

- Each sepset needs to separate graph into two conditionally independent parts



# Summary

- Correctness of clique tree inference relies on running intersection property
- Running intersection property implies separation in original distribution
- Implies minimal complexity incurred by any clique tree: *separators*
  - Related to minimal induced width of graph *cliques*