Course Logistics

- Happy Election Day!
- We're still working on MP1...
- The MP was intractably difficult
- Instead, HW4 on BNs was posted on the web site yesterday; it is due a week from today.

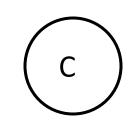
BN Construction

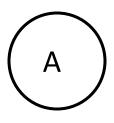
- Identify variables
- Order them*
- While there are variables to add
 - Pick the next in the ordering
 - Identify its parents in the net
 - Hold all others constant (in every configuration)
 - If net variable influences it, net var is a parent
 - Draw all arcs and add CPT
- * order matters a lot

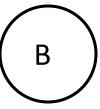
Dentist Example

3 Boolean Random Variables:

- C Patient has a cavity
- A Patient reports a toothache
- B Dentist's probe catches on tooth





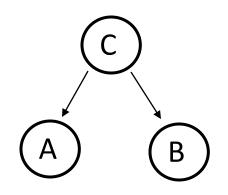


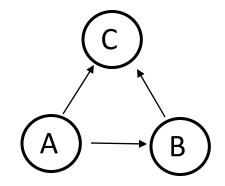
Effect of Order on BN structure

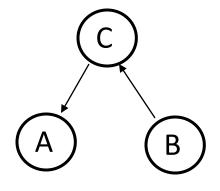
Order: C A B

Order: A B C

Order: B C A







5 parameters

7 parameters no savings over the joint

5 parameters

BN Structure

consider ordering M,J,E,B,A

B – a burglary is in progress

E – an earthquake is in progress

A – the alarm is sounding

J – John calls

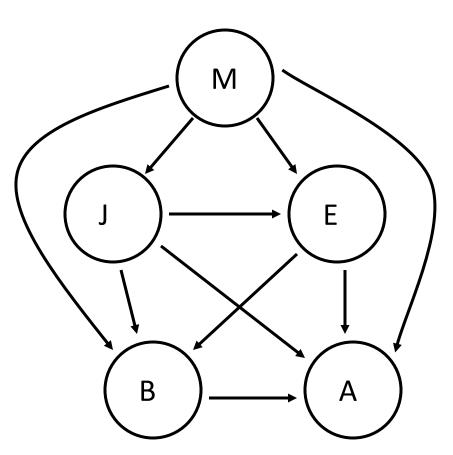
M – Mary calls

All variables are Boolean

How many numbers?

No savings over the Joint

What conditional independence assumptions does it make?



Ordering is Important

- Original net B E A J M
 - 10 parameters
 - Saves 21 over the joint
- Ordering M J E B A
 - 31 parameters
 - No savings over the joint
- "Causality" ordering is only heuristic
- Suppose we want to test all orderings of n random variables...

Proof of Adequacy for BNs

- Recover the joint
- Conditionalize on evidence
- Marginalize across unknowns
- Brute Force calculation of the Answer
- May be intractable for large BNs
- Joint has aⁿ for n RV's, "a" discrete values each
- Earthquake examples were all much simpler
 - Did not recover the full joint
 - Did not touch irrelevant values

Space Complexity of Distribution Data Structures

BNs win big if direct interdependencies (fan in) are bounded (limited by k)

Joint complexity:

O(aⁿ): 2ⁿ-1 for n Boolean variables

BN w/ at most k parents / node:

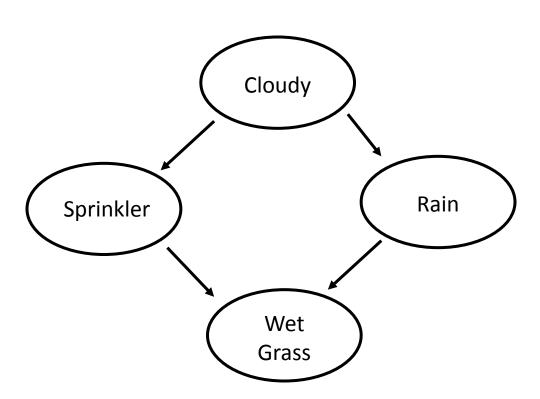
O(n): $\leq n \cdot 2^k$ for Boolean variables

Time Complexity of Inference

 Efficient exact algorithms for any query (evidence/query pairs) for polytrees*

* at most one undirected path between any two nodes

- Inference in non-polytrees can be exponential
 - Even approximate inference
 - Often works well in practice
 - Worst case is #P-hard (worse than NP complete)



Exact inference - clustering

- Join tree / junction tree algorithm
- Restore polytree property
- Identify clusters of nodes (super nodes)
- Collect all parents
- Exponential in cluster size (context grows as product of contexts)

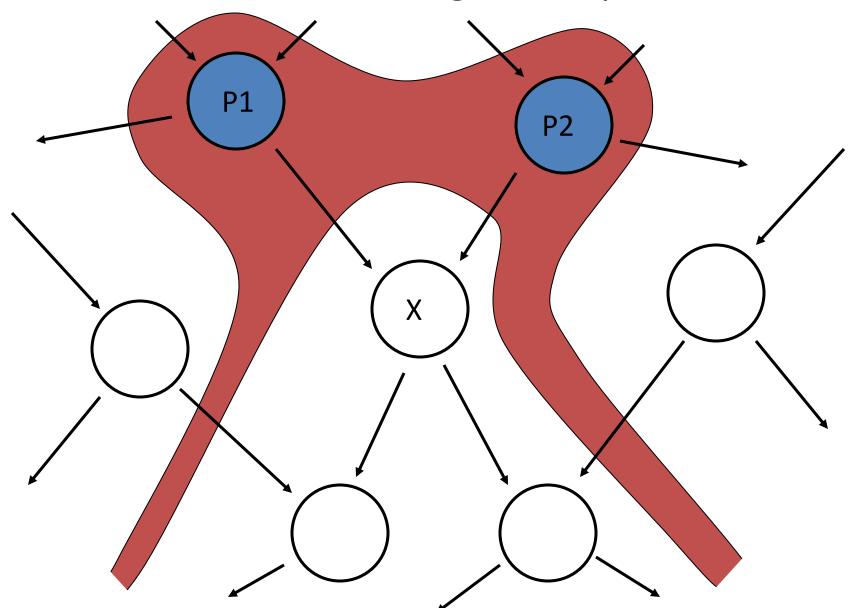
approximate inference

- Loopy belief propagation
 - Pretend it is a poly-tree
 - No guarantee of convergence / correctness
 - Surprisingly effective
- Sampling algorithms
 - Generate examples from the posterior
 - Given the evidence
 - Estimate the distribution by counting

Monte Carlo sampling algorithms Rejection sampling

- Start w/ all RV's unknown
- Choose a topological ordering
- Get random numbers to assign values
- Descendents are always unknown when assigning an RV
- Generate lots of samples
- Count but ignore samples violating evidence

X is conditionally independent of its non-descendants given its parents



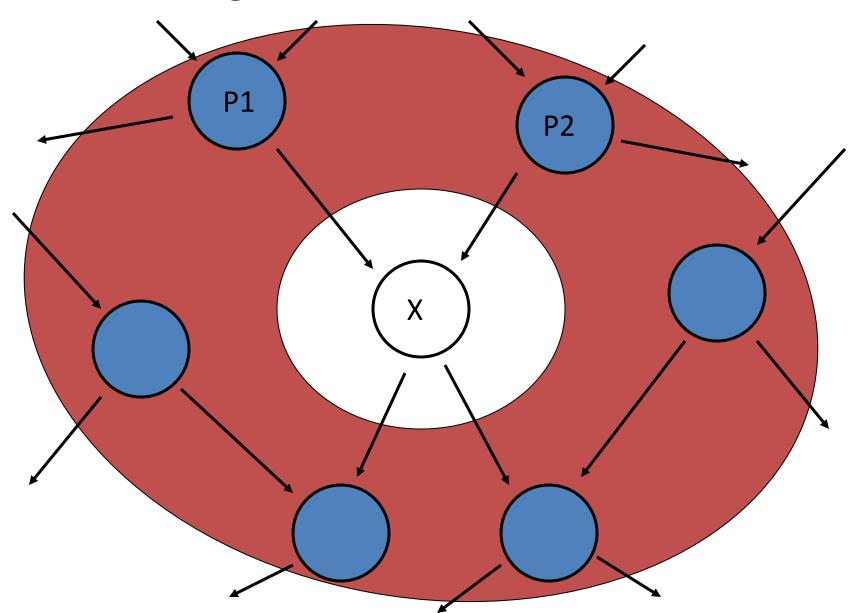
Monte Carlo sampling algorithms Likelihood weighting

- Fix evidence variables to their values
- Generate a sample randomly assigning other variables
- Generate lots of samples
- But count each weighted according to Pr(evidence | dice rolls)
- All samples are consistent w/ evidence
- Some count for very little

Monte Carlo sampling algorithms Gibbs sampling

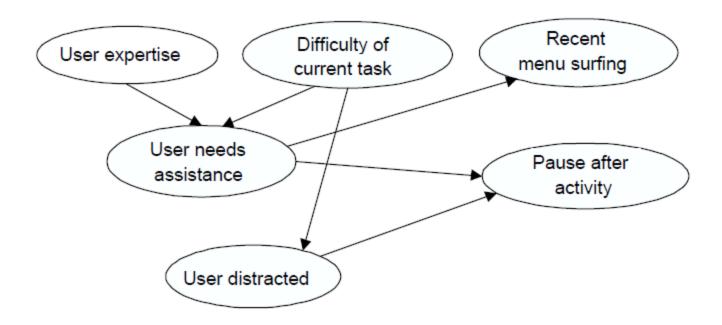
- Markov Chain Monte Carlo (MCMC)
- Fix evidence variables to their values
- Initialize others randomly
- Iterate through nodes
- Resample each given its Markov blanket
- After 'burn in,' accumulate each k'th sample
- Count

X is conditionally independent of everything else given its Markov blanket

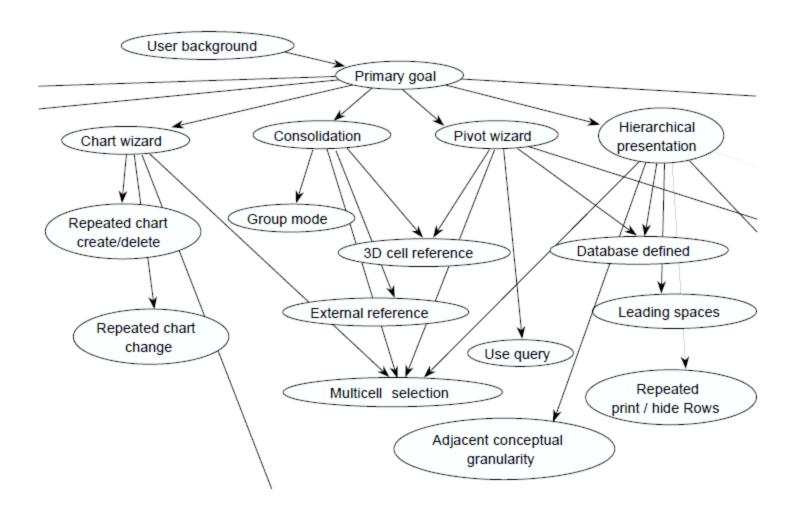


Example

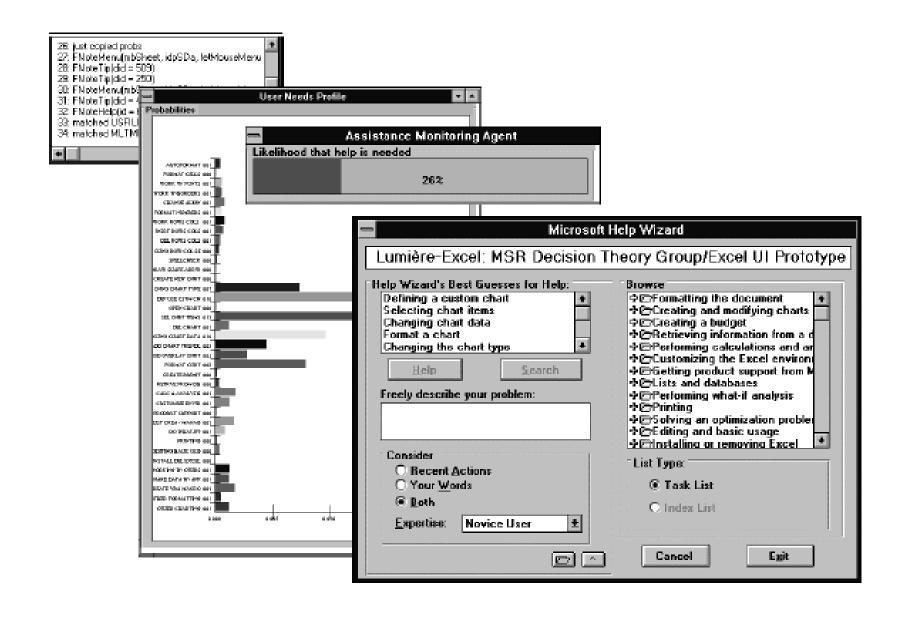




Early BN for when Clippy should pop up



Partial BN of Clippy's prior knowledge of MS Excel



Clippy's Bayesian reasoning: Event stream, Current probability distribution of important RVs, Pr(User wants Clippy), Prototype human interface