

# 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 can influence it, that net var is a parent
  - Draw all arcs and add CPT

\* order matters a lot

# Heuristic

Usually the most compact representation  
results when belief causality mirrors  
physical causality

# Dentist Example

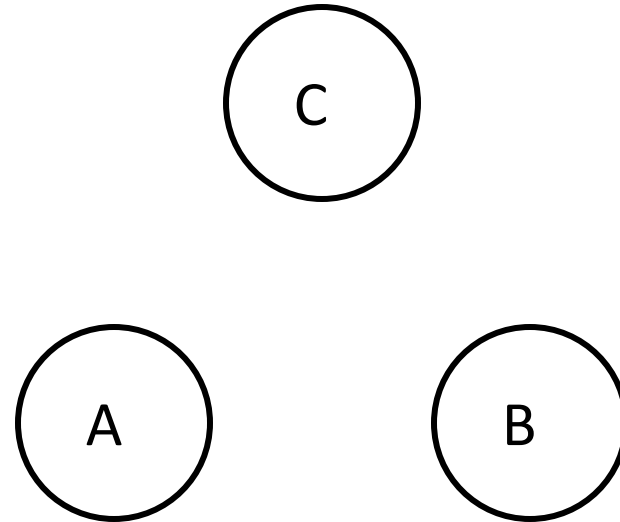
3 Boolean Random Variables:

C – Patient has a cavity

A – Patient reports a  
toothache

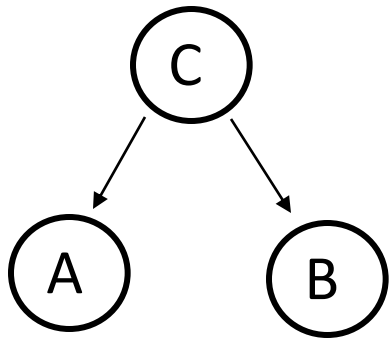
B – Dentist's probe catches  
on tooth

Joint: 7 parameters



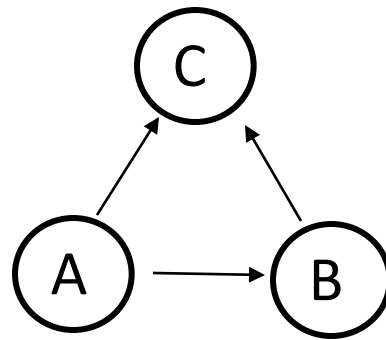
# Effect of Order on BN structure

Order: C A B



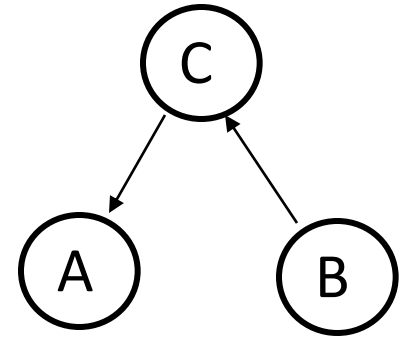
5 parameters

Order: A B C



7 parameters  
no savings over  
the joint

Order: B C A



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

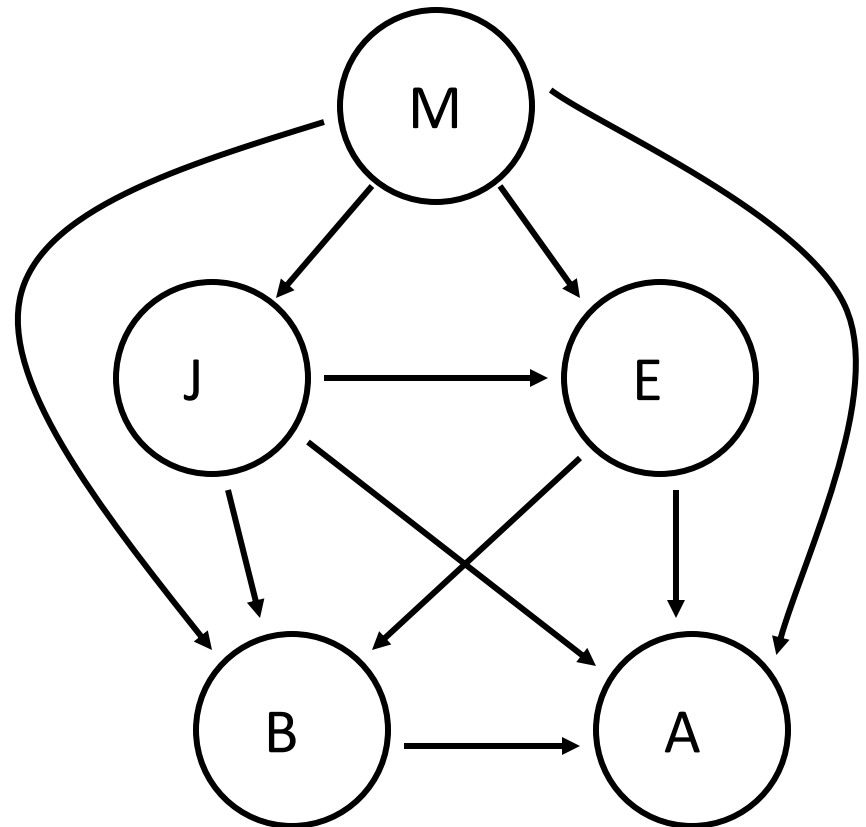
M – Mary calls

All variables are Boolean

How many numbers?

No savings over the Joint

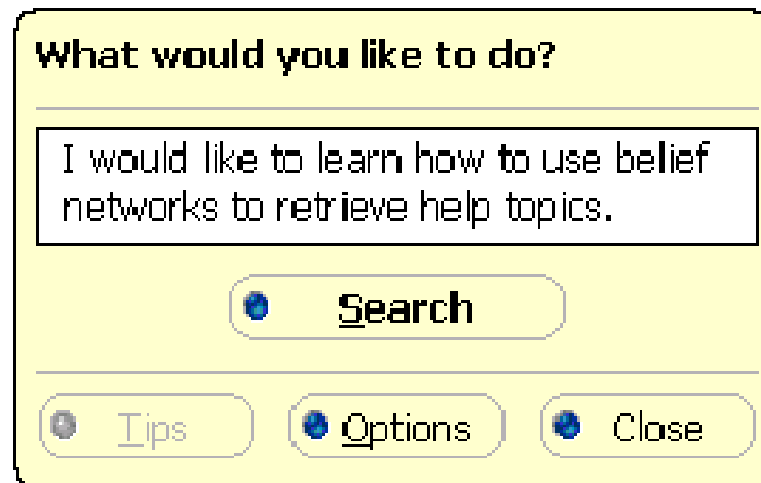
What conditional independence assumptions does it make?



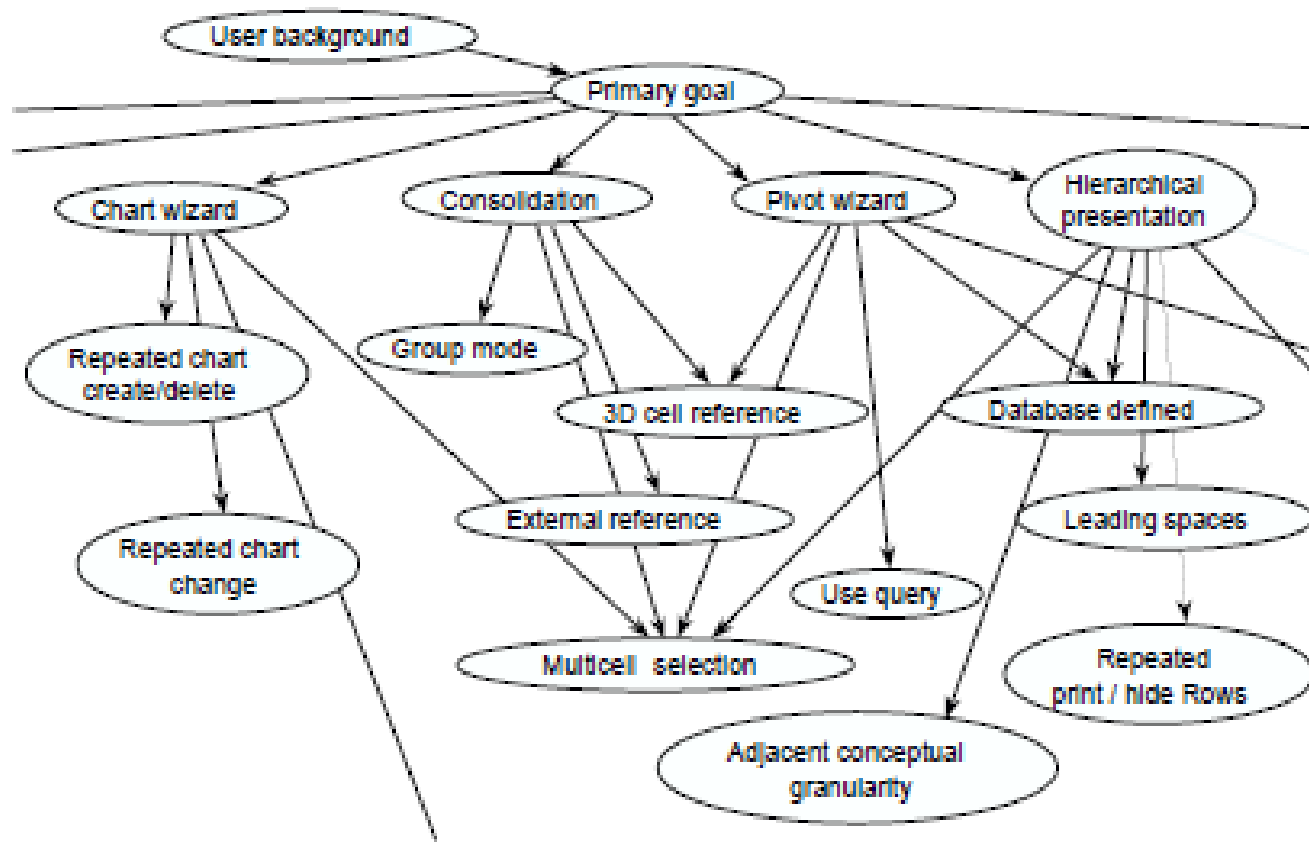
# Ordering is Important

- 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...

# Clippy (Lumiere Project): Another Real World(?) BN Example

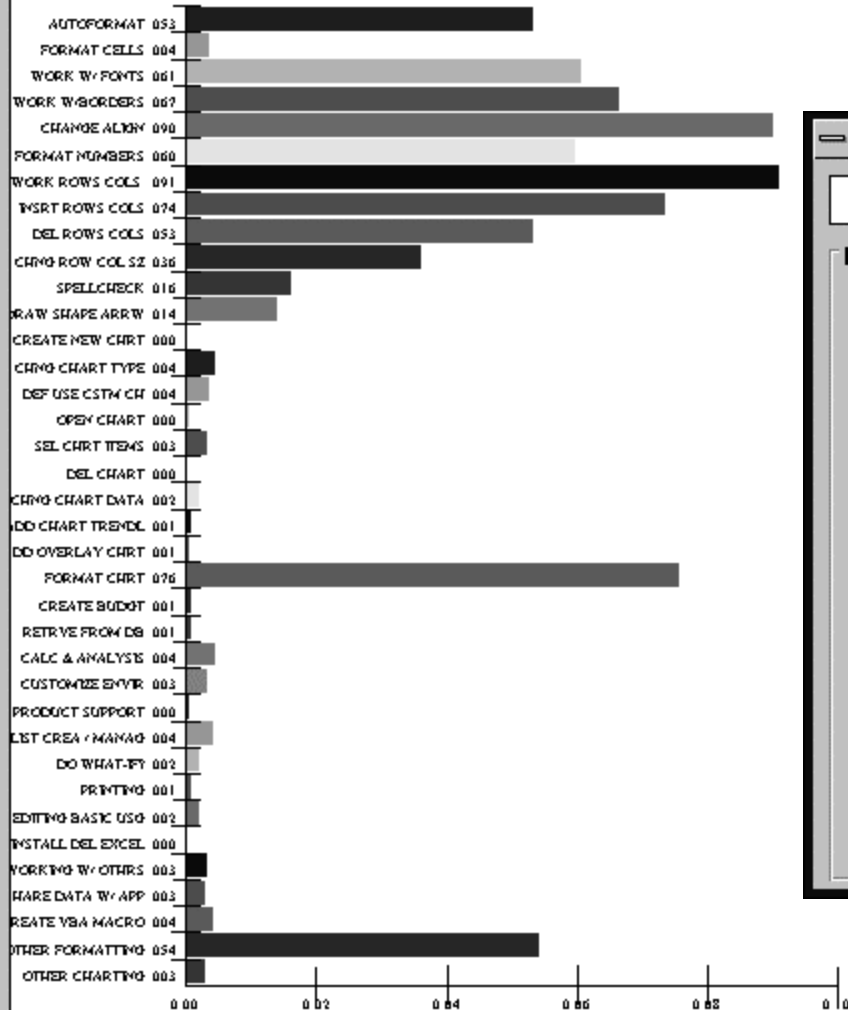


# Portion of Clippy's BN for Excel User Modeling



## User Needs Profile

Probabilities of relevant help topics given a request for help



## Microsoft Help Wizard

Lumière-Excel: MSR Decision Theory Group/Excel UI Prototype

Help Wizard's Best Guesses for Help:

Autoformatting my document  
Format a chart  
Changing alignment  
Working with borders  
Working with fonts

Help

Search

Freely describe your problem:

how do i make this look prettier?

Consider

Recent Actions

Your Words

Both

Expertise: Novice User

Browse

+ Formatting the document  
+ Creating and modifying charts  
+ Creating a budget  
+ Retrieving information from a d  
+ Performing calculations and ar  
+ Customizing the Excel environ  
+ Getting product support from M  
+ Lists and databases  
+ Performing what-if analysis  
+ Printing  
+ Solving an optimization problem  
+ Editing and basic usage  
+ Installing or removing Excel

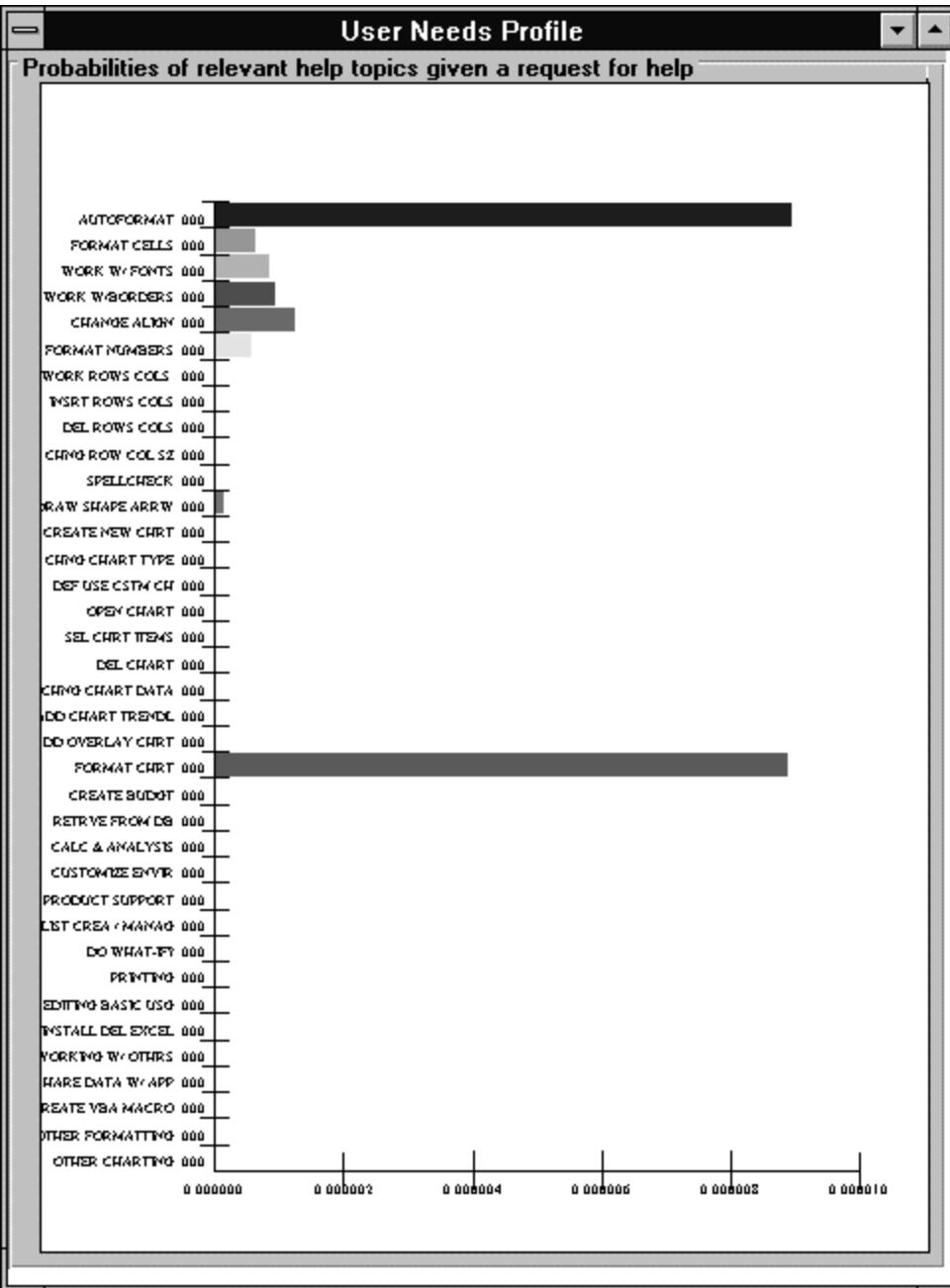
List Type:

Task List

Index List

Cancel

Exit



## Posterior

- Query nodes after
- Conditioning
- Marginalizing

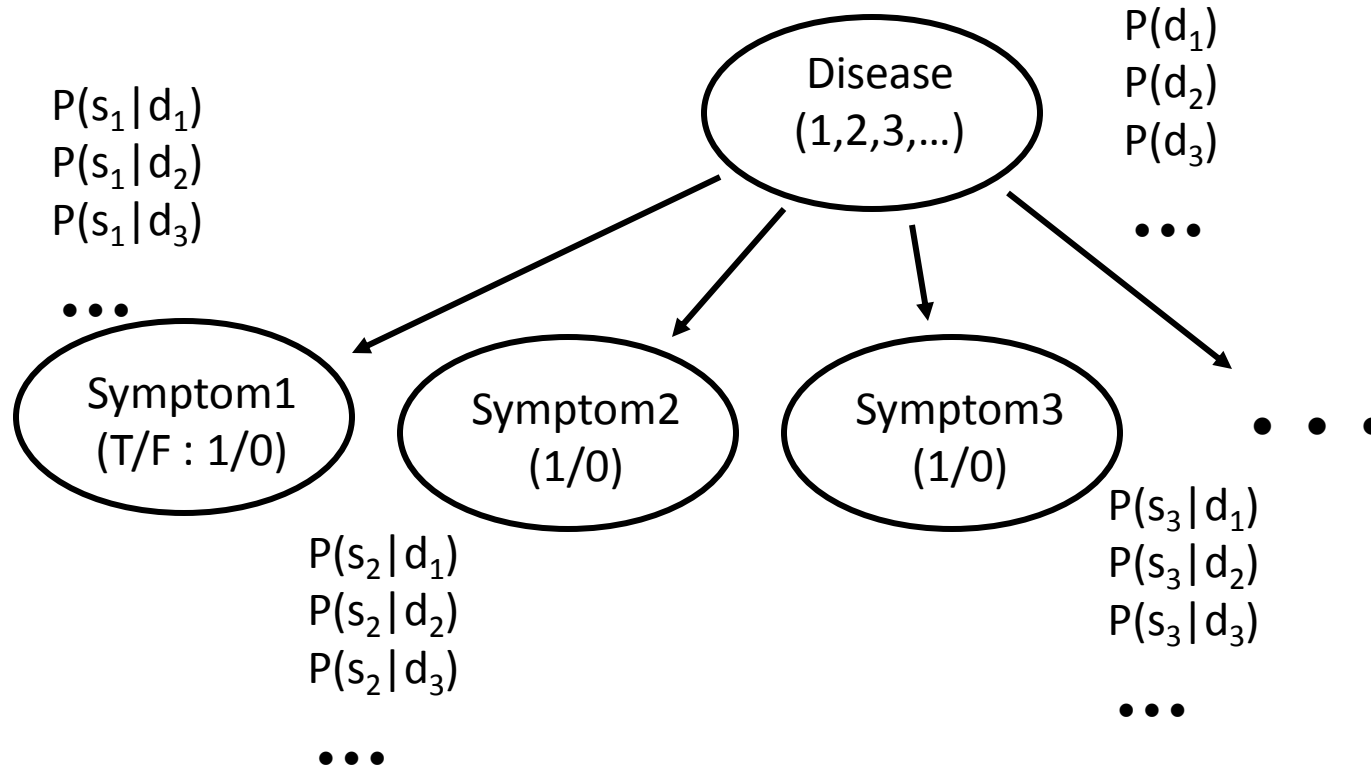
# Naïve Bayes: a Simple and Important BN type

- Assume Boolean Attributes (Symptoms) are conditionally independent of each other given the Classification (Disease)
- Impoverished expressiveness
- Works surprisingly (?) well in practice
- Common first choice

NOTE the use of intentionally impoverished model for computational convenience

What does a naïve BN look like?

# Naïve Bayes



Infer most likely disease:

$$\arg \max_{d_i} P(d_i | \bar{S})$$

# Naïve Bayes

Diagnose by inference with:

First three are  
probability models

|D| Functions, each assigns probability  
over the Boolean hypercube

If the probability models were correct  
we would no longer need to see  
the world

Fourth is not normalized (why does that work?)

Parameters are adjusted to best fit the world training samples

Fitting models to data is a kind of machine learning

$$\arg \max_{d_i} P(d_i | \bar{S})$$

$$\arg \max_{d_i} \frac{P(d_i) \cdot P(\bar{S} | d_i)}{P(\bar{S})}$$

$$\arg \max_{d_i} \frac{P(d_i) \cdot \prod_j P(s_j | d_i)}{P(\bar{S})}$$

$$\arg \max_{d_i} \left( P(d_i) \prod_j P(s_j | d_i) \right)$$

# Naïve Bayes

$$\arg \max_{d_i} \left( P(d_i) \prod_j P(s_j | d_i) \right)$$

- Suppose we always reason from observed symptoms to diseases & symptoms are Boolean
- Characterize the decision boundaries
- $\text{Log}(x)$  is monotonically increasing,

so:

$$\arg \max_{d_i} \text{Log} \left( P(d_i) \prod_j P(s_j | d_i) \right)$$

- Log of a product is...

# Naïve Bayes

$$\arg \max_{d_i} \text{Log} \left( P(d_i) \prod_j P(s_j | d_i) \right)$$

- We represent the symptoms  $S$  (evidence) is a Boolean vector:

$$\begin{aligned} s_k &= 1 \text{ if } k\text{'th symptom is present} \\ &= 0 \text{ if absent} \end{aligned}$$

- Becomes  $\arg \max_i$  of  $F_i(S)$   
where  $S$  is the Boolean vector of evidence

- $F_i(S) = \text{Log } P(d_i) + \sum \text{Log } P(s_k | d_i)$

- Form is  $F_i(S) = a_i + B_i \cdot S$

$$\begin{aligned} a_i &= \text{Log } P(d_i) + \sum \text{Log } P(s_k=0 | d_i) \\ b_{ik} &= \text{Log } P(s_k=1 | d_i) - \text{Log } P(s_k=0 | d_i) \end{aligned}$$

- The Log Prob fcns are...

hyperplanes cutting through the Boolean hypercube

# Naïve Bayes

$$\arg \max_{d_i} \text{Log} \left( P(d_i) \prod_j P(s_j | d_i) \right)$$

- Form  $F_i(S) = a_i + B_i \cdot S$   
Each is a linear polynomial in  $S$
- Diagnose using the highest valued function at a point  $S$
- What are the Naïve Bayes decision boundaries?
- (How do hyperplanes interact?)
- How do we determine  $a$ 's and  $B$ 's?
- These are *generative models* for the diseases