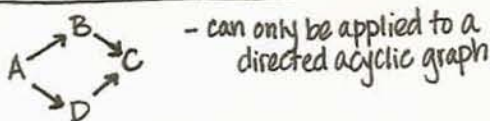
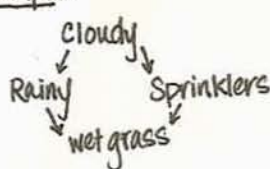


BAYESIAN NETWORK ANALYSIS



- can only be applied to a directed acyclic graph

Example.



P(R C)	T	F
	0.8	0.2
F	0.2	0.8

P(C)	T	F
	0.5	0.5

P(S C)	T	F
	0.1	0.9
F	0.5	0.5

P(W S,R)	T	F
	0.99	0.01
F	0.9	0.1
T	0.9	0.1
F	0.0	1.0

W=T
 The chain for probability
 $P(C,S,R,W) = P(C) \cdot P(S|C) \cdot P(R|S,C) \cdot P(W|R,S,C)$
 $= P(C) \cdot P(S|C) \cdot P(R|C) \cdot P(W|S,R)$
 Current model topology

BAYES THEOREM: $P(A,B) = P(A|B)P(B)$
 $P(A|B) = P(A)P(B|A) = P(B)P(A|B)$

$$P(S=T|W=T) = \frac{P(S=T, W=T)}{P(W=T)} = \frac{\sum_{C,R} P(C,S=T,R,W=T)}{P(W=T)} = 0.4298$$

$$P(R=T|W=T) = \frac{P(R=T, W=T)}{P(W=T)} = \frac{\sum_{C,S} P(C,S,R=T,W=T)}{P(W=T)} = 0.7079$$

$$P(W=T) = 0.6471$$

Experiments give $P(\text{Data})$
 Models can produce $P(\text{Data}|\text{Model})$
 What we want to compare $P(\text{Model}|\text{Data})$

$$P(\text{Model}|\text{Data}) = \frac{P(\text{Data}|\text{Model})P(\text{Model})}{P(\text{Data})}$$

$$\text{Score}_i = \log P(\text{Model}_i|\text{Data}) = \log P(\text{Data}|\text{Model}_i) + \log P(\text{Model}_i) - \log P(\text{Data})$$

← Prior knowledge about system is constant across all models

Sachs, Gifford, Jaakkola, Sorger, & Lauffenburger
<http://www.stke.org/cgi/content/full/sigtrans;2002/148/p.1>
 Stimulates (cue) - fibronectin (fn)
 (F) Focal Adhesion Kinase
 (E) Extracellular signal-related kinase
 Experiment was carried out twice

- M0: Cue (E) (F)
 M1: Cue → (F) → (E)
 M2: Cue → (E) → (F)
 M3: Cue → (E) (F)
 M4: Cue → (E) (F)

	score	Data Set I	Data Set II	(fold-change)
M0:	-50.5 (254)	-60.8	(1700)	
M1:	-42.8* (1)	-58.0	(104)	
M2:	-43.2 (3)	-57.6	(73)	
M3:	-44.3 (4)	-55.6	(10)	
M4:	-44.3 (4)	-53.4*	(1)	

Simulation of Reaction Kinetics:

Continuous vs Stochastic (ODE)

$$\frac{dx}{dt} = f(x, u)$$

Alternative framework

$$\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \leftarrow \begin{array}{l} \text{\# of molecules} \\ \text{of species } i \end{array}$$

Think about dynamics as:

- 1) System remains unchanged for time τ
- 2) System will change by a single chemical reaction, μ

Let $P(\tau, \mu) \delta\tau$ = probability that \vec{y} at time t will evolve such that no reaction occurs in $(t, t+\tau)$, but that the next reaction occurs in $(t+\tau, t+\tau+\delta\tau)$ and, it is Reaction μ = reaction probability density function

$$= P_0(\tau) \cdot a_\mu \delta\tau$$

probability that waiting time is τ probability that μ is the reaction

It is shown that

$$P(\tau, \mu) = a_\mu e^{-a_0 \tau} \text{ for } 0 \leq \tau < \infty$$

where $a_\mu \in \{1, \dots, M\}$

$$a_0 = \sum_{i=1}^M a_i$$

individual reaction propensities

$$a_\mu = h_\mu C_\mu$$

reaction rate # of collision types

DT Gillespie *J. Phys Chem* 81. 2340-61 (1977)
J. Phys Chem 115. 1716-33 (2001)

Pick 2 random numbers R_1, R_2 uniformly on $[0,1]$

$$\tau = \frac{1}{a_0} \ln \frac{1}{R_1}$$

select μ such that $\sum_{i=1}^M a_i < R_2 a_0 \leq \sum_{i=1}^M a_i$

