

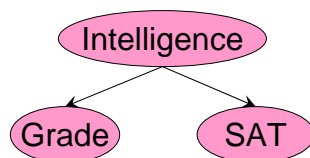
Mathematical Tools: Probabilistic Models

Lecture 3

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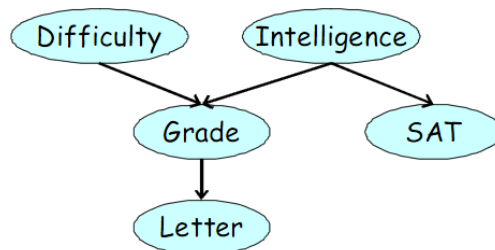
Bayesian network

- Directed acyclic graph (DAG)
 - Node: a random variable
 - Edge: direct influence of one node on another
- The *Student* example revisited
 - Intelligence (I), SAT (S), Grade (G)
 - Val (I) = $\{i^1, i^0\}$, Val (S) = $\{s^1, s^0\}$, Val (G) = $\{g^1, g^2, g^3\}$



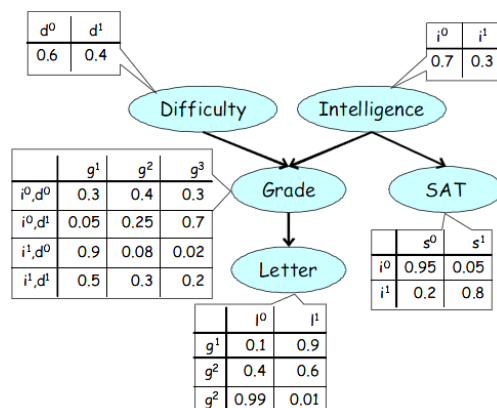
The *Student* example revisited

- More complex scenario
 - Course difficulty (D), quality of the recommendation letter (L), Intelligence (I), SAT (S), Grade (G)
 - $\text{Val}(D) = \{\text{easy}, \text{hard}\}$, $\text{Val}(L) = \{\text{strong}, \text{weak}\}$,
 $\text{Val}(I) = \{i^1, i^0\}$, $\text{Val}(S) = \{s^1, s^0\}$, $\text{Val}(G) = \{g^1, g^2, g^3\}$
 - Joint distribution requires 48 entries



The *Student* Bayesian network

- Joint distribution
 - $P(I, D, G, S, L) = P(I)P(D)P(G|I, D)P(S|I)P(L|G)$



Basic independencies

- Given G, L is independent of I, D, S.
- Given I, S is independent of D, G, L.
- I and D are independent
- D is independent of I and S
- :
- The intuition
 - The parents of a variable “shield” it from probabilistic influence that is causal in nature.

Bayesian network semantics

- **Definition:** A Bayesian network structure G is a directed acyclic graph whose nodes represent random variables X_1, \dots, X_n .
Pa X_i : parents of X_i in G
NonDescendants X_i : the variables in G that are not descendants of X_i .

Then, G encodes the following set of conditional independence assumptions, called the local Markov assumptions, and denoted by $I_L(G)$:

For each variable X_i :

Bayesian network joint distribution

- **Definition:** Let G be a Bayesian network graph over the variables X_1, \dots, X_n . We say that a distribution P over the sample space factorizes according to G if P can be expressed as a product:

- **Definition:** A Bayesian network is a pair (G, P) where P factorizes over G , and where P is specified as a set of CPDs associated with G 's nodes.

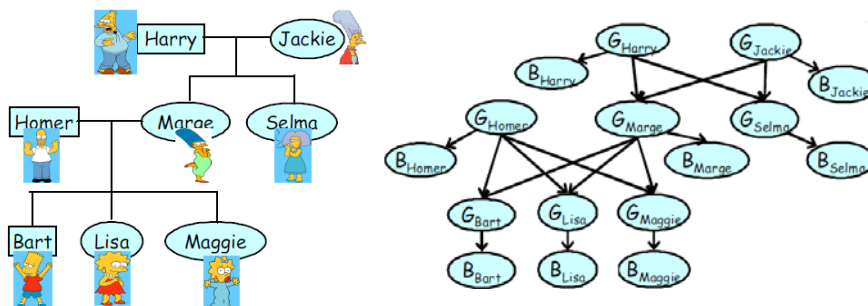
[Koller and Friedman]

Computational Molecular Biology and Genomics, Spring 2009

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The *Genetics* example

- Variables
 - B: blood type (a phenotype)
 - G: genotype of the gene that encodes a person's blood type; $\langle A, A \rangle$, $\langle A, B \rangle$, $\langle A, O \rangle$, $\langle B, B \rangle$, $\langle B, O \rangle$, ...



[Koller and Friedman]

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The *Genetics* example

- CPDs
 - The penetrance model $P(B(c) | C(c))$
 - The transmission model $P(G(c) | G(p), G(m))$
 - Genotype priors $P(G(c))$

Acknowledgement

- This set of slides is based on the following materials:
 - “Probabilistic Graphical Models: Principles and Techniques” by Daphne Koller and Nir Friedman