

CS440/ECE448: Intro to Artificial Intelligence

Lecture 18:

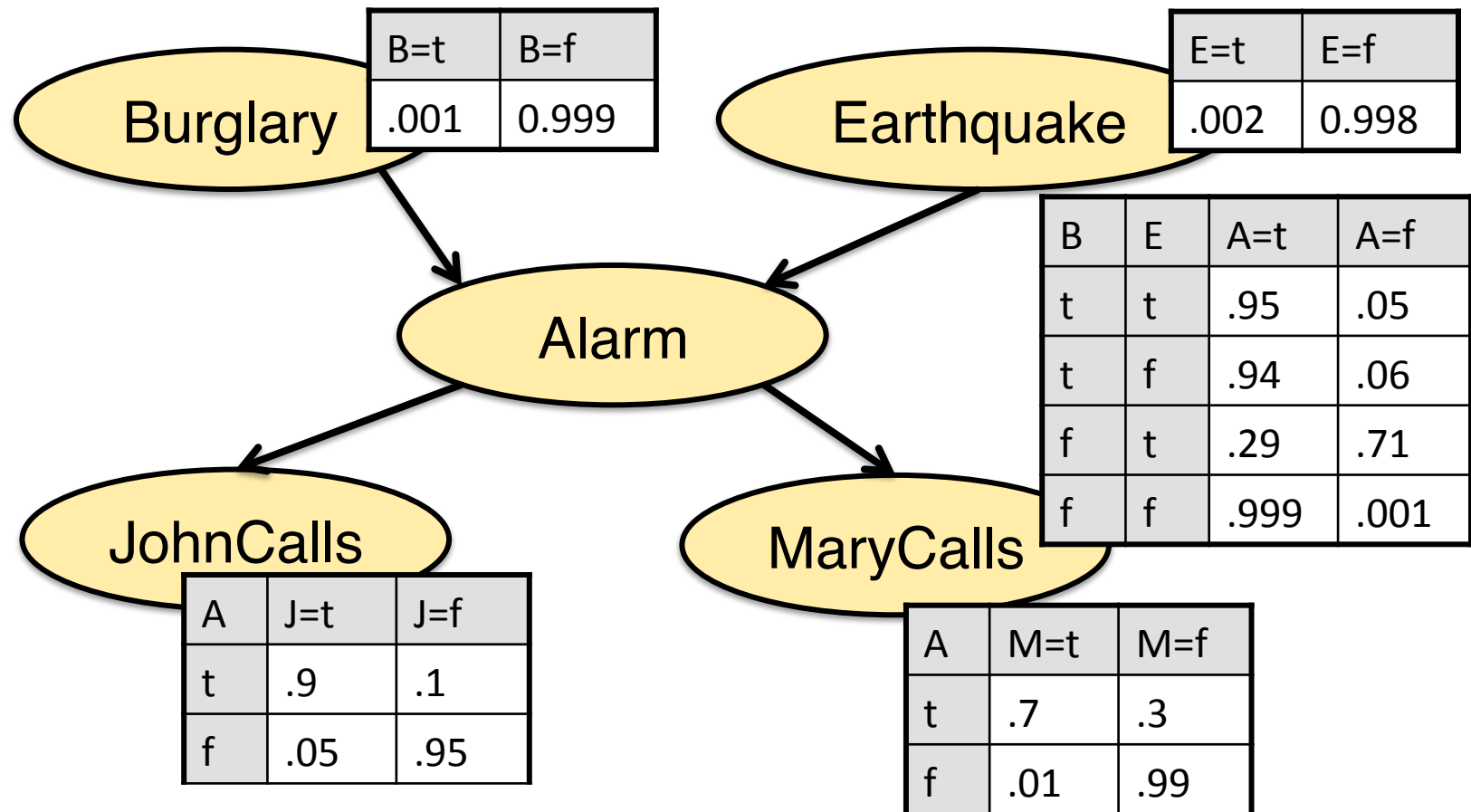
Bayesian Networks

continued

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<http://cs.illinois.edu/fa11/cs440>

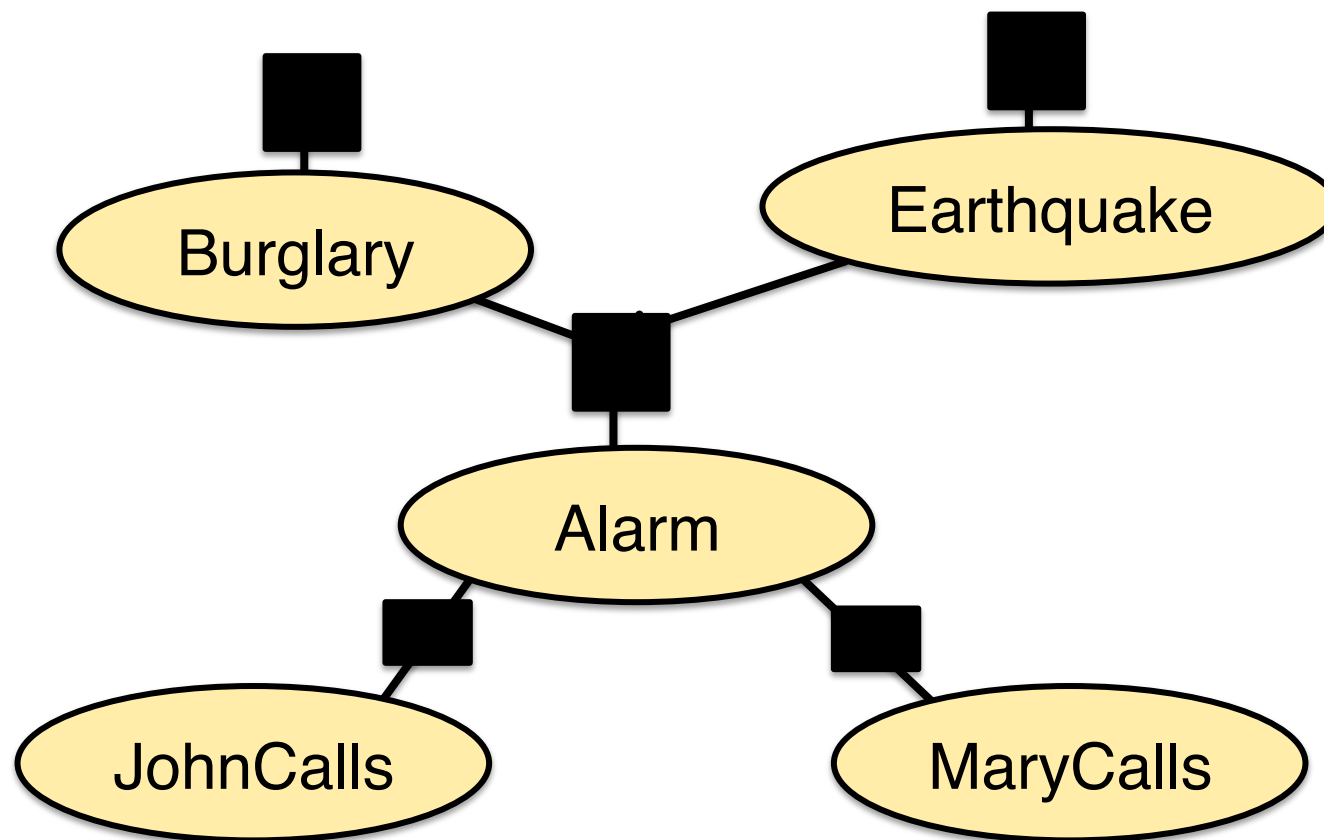
The *Burglary* example

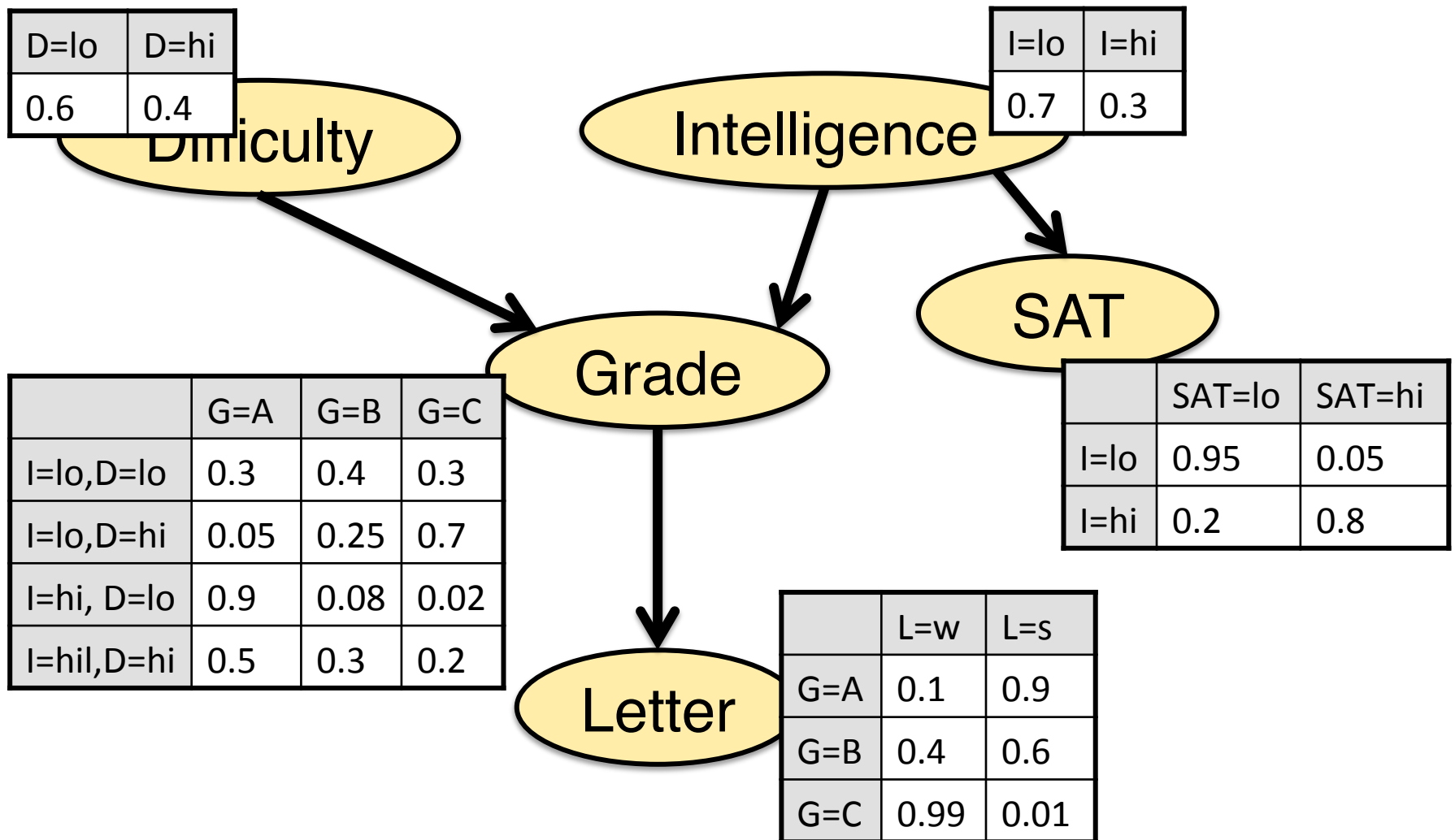


What is the probability of a burglary if John and Mary call?

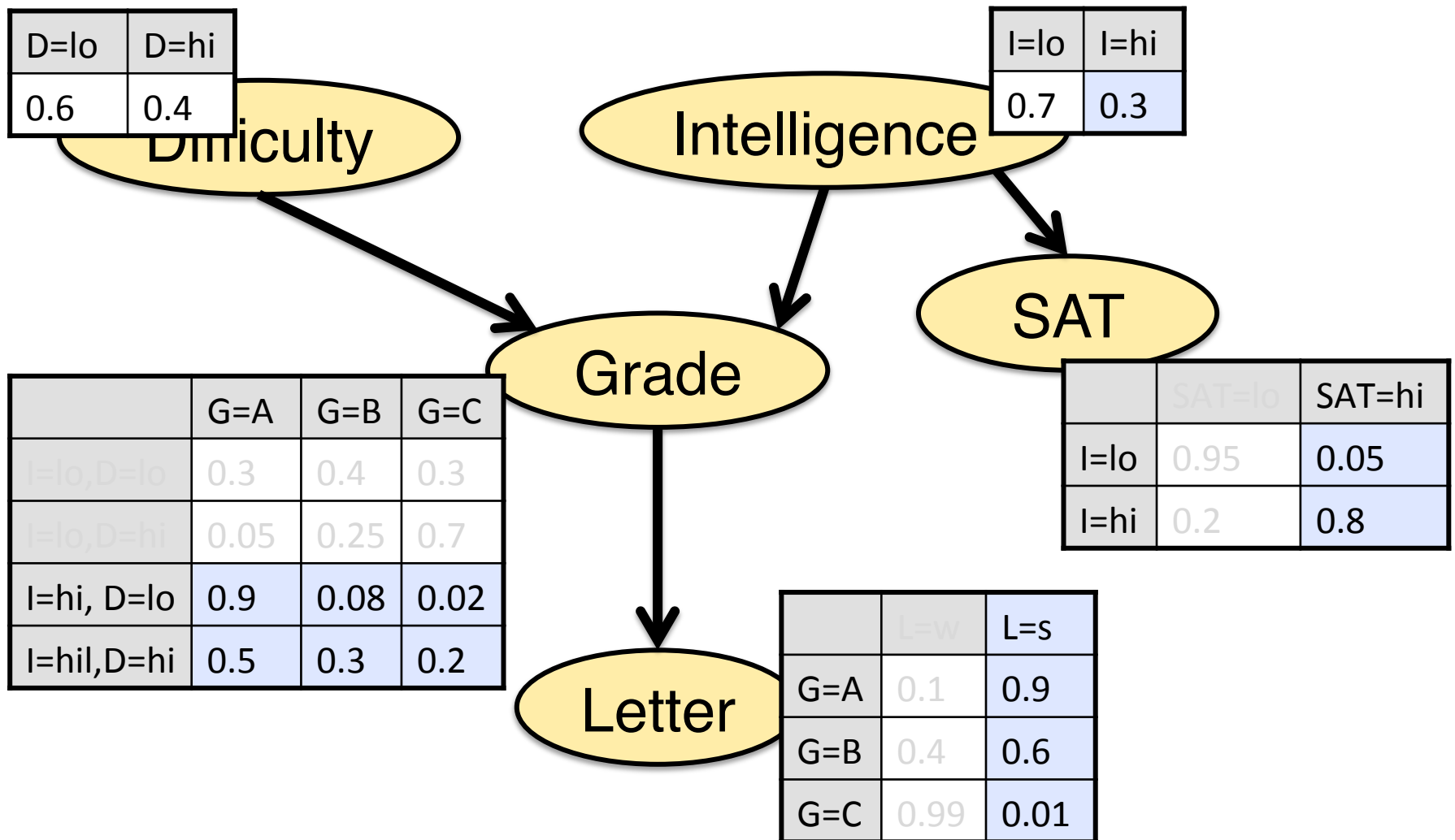
Factor graphs

Each CPT is one factor (= black box)

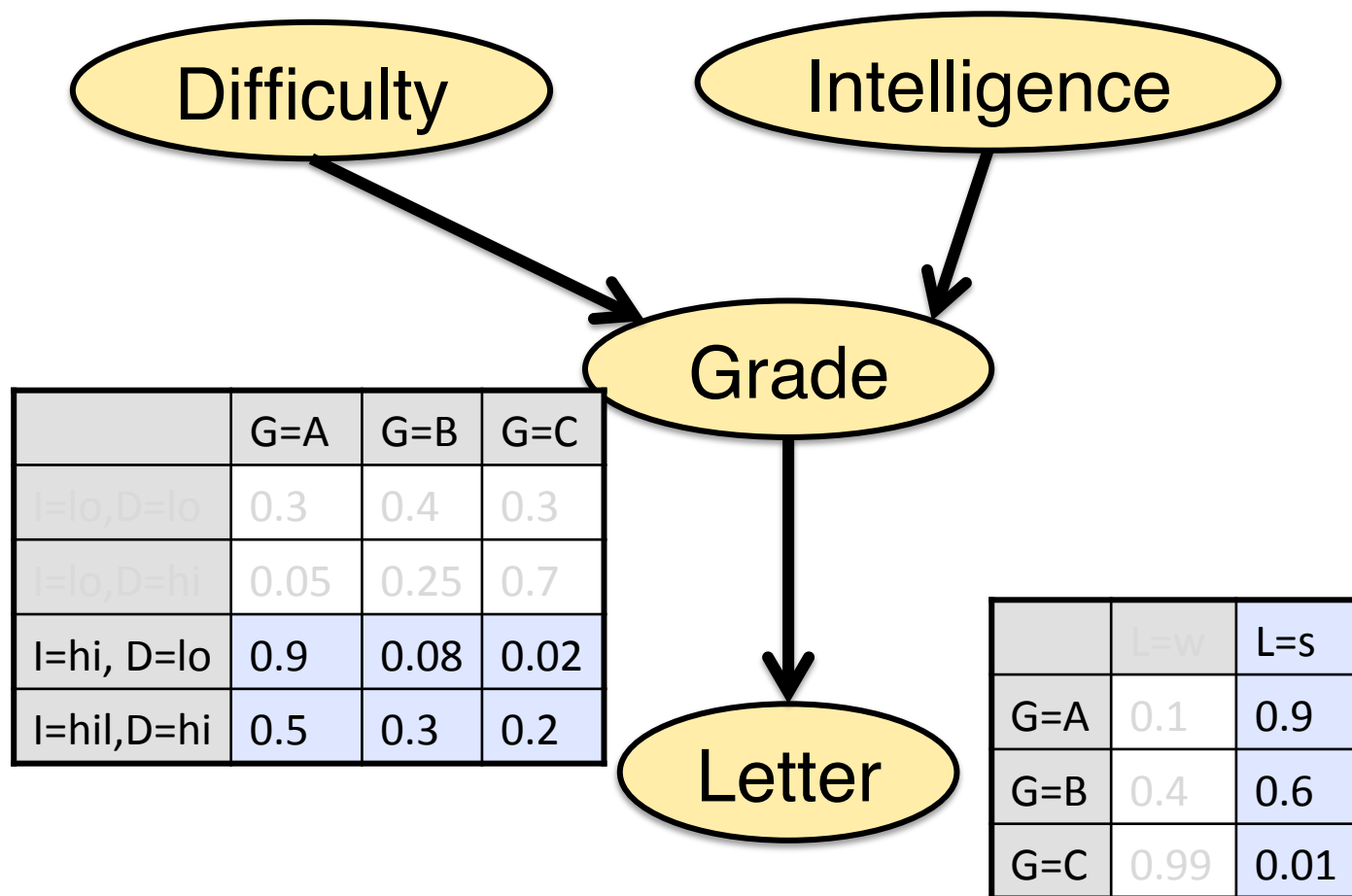




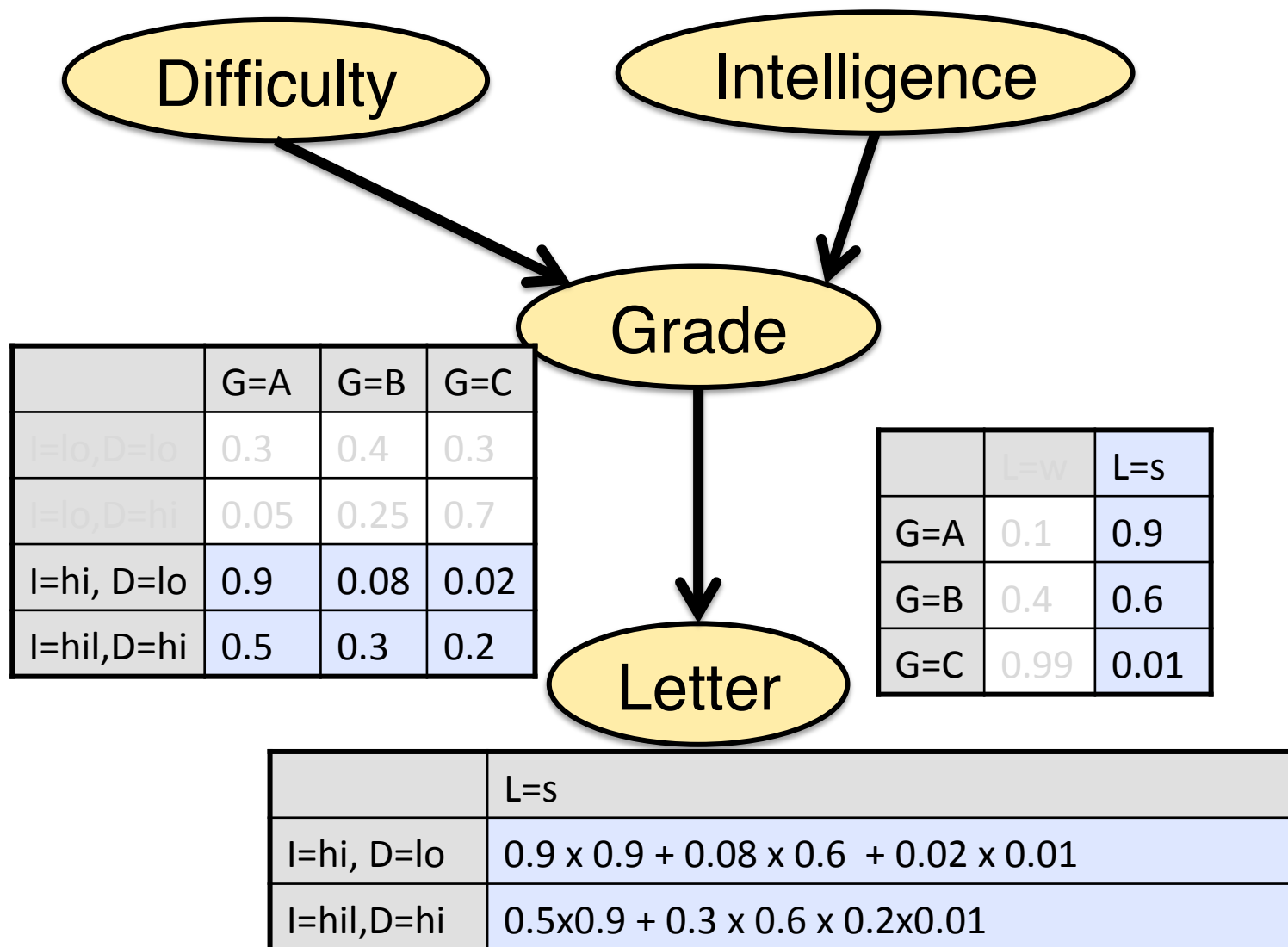
What's the probability that the student is intelligent if he got a good letter and a high SAT test?

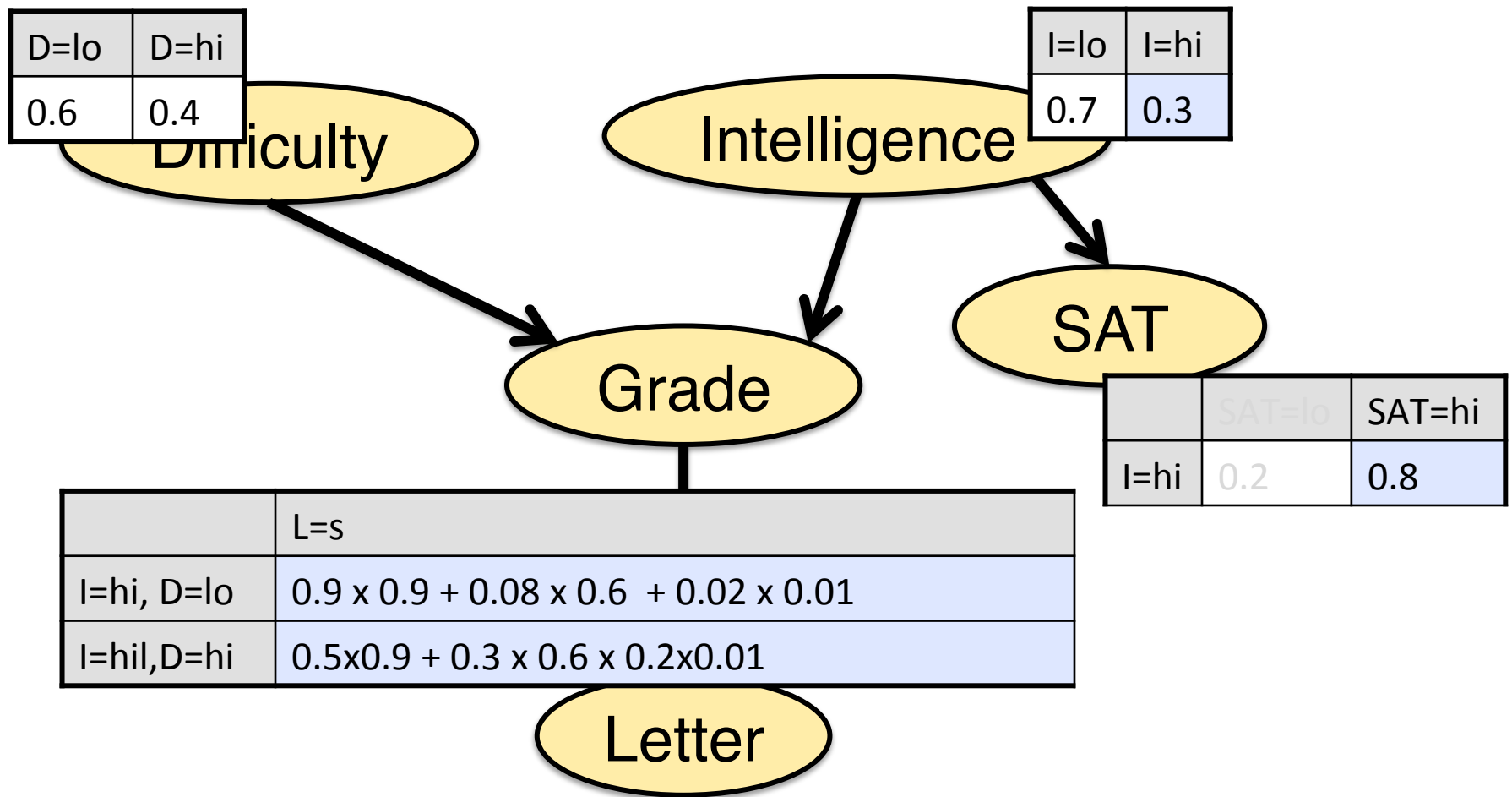


What's the probability that the student is intelligent if he got a good letter and a high SAT test?

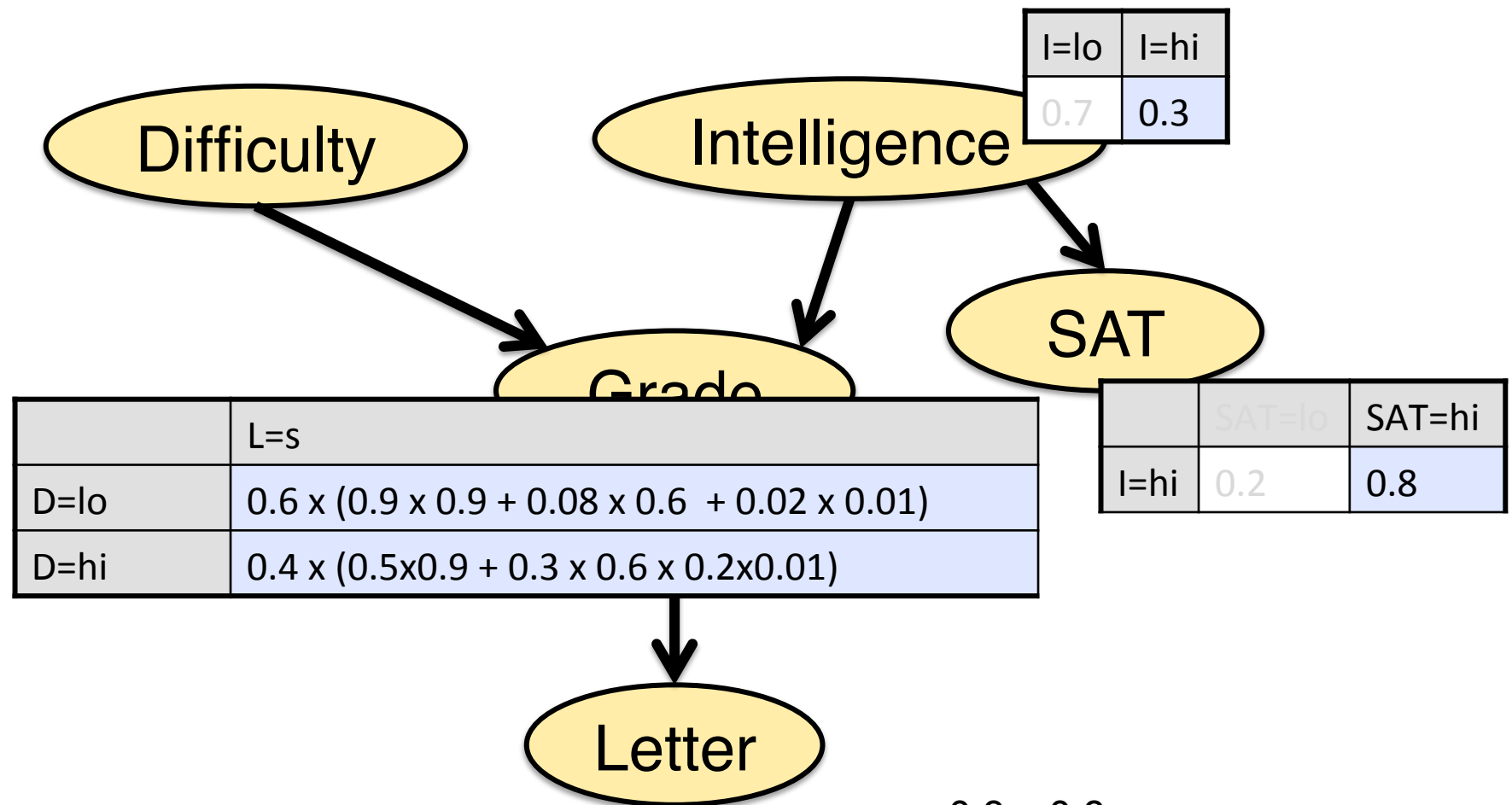


What's the probability that the student is intelligent if he got a good letter and a high SAT test?





What's the probability that the student is intelligent if he got a good letter and a high SAT test?



What's the probability that the student is intelligent if he got a good letter and a high SAT test?

$$0.3 \times 0.8 \times [0.6 \times (0.9 \times 0.9 + 0.08 \times 0.6 + 0.02 \times 0.01) + 0.4 \times (0.5 \times 0.9 + 0.3 \times 0.6 \times 0.2 \times 0.01)]$$

Complexity of variable elimination

Polytree networks:

only one (undirected) path between each pair of nodes.

Complexity is linear in the size of the network (number of CPT entries)

Complexity of variable elimination

Multiply connected networks:

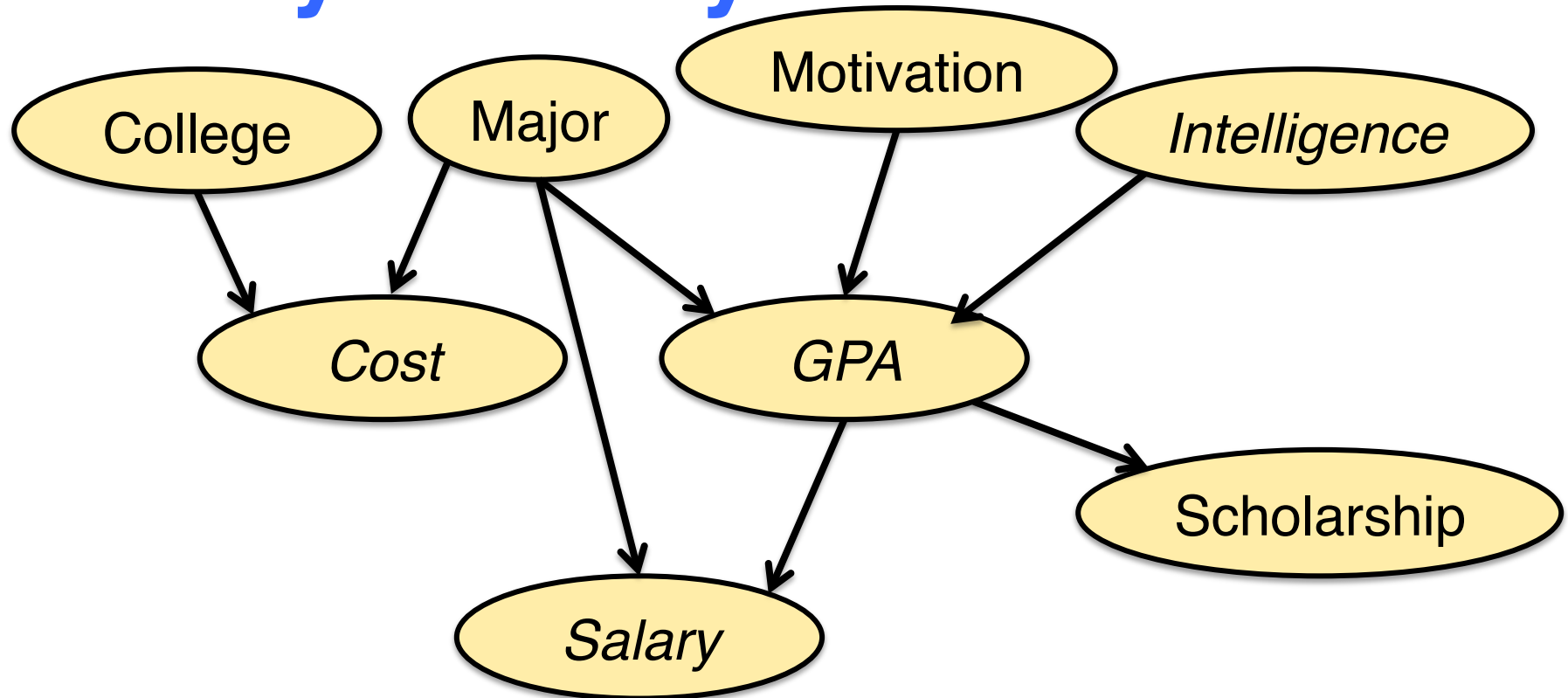
multiple (undirected) paths between some pair of nodes.

Complexity is **exponential** in the size of the network (number of CPT entries)

- (Similar to finding number of models in propositional logic)

Continuous variables

A hybrid Bayesian Network



The cost of your degree (tuition and living) depends on your choice of college and major. Your GPA depends on your motivation (high/normal/low), your IQ and your major.

Whether you get a scholarship depends on your GPA.

Your starting salary after you graduate depends on your major and your GPA.

Continuous variables

Continuous R.V.s (without parents)

e.g. *Temperature* (may not depend on anything)

Continuous R.V.s with discrete parents

e.g. *Tip* might depend on *CustomerIsHappy*

Continuous R.V.s with continuous parents

e.g. *SATscore* might depend on *IQ**

**(it's easier to treat integers like reals)*

Discrete R.V.s with continuous parents

e.g. *CustomerIsHappy* might depend on *Temperature*

R.V.s with mixed (discrete&continuous) parents

Continuous random variables

Many quantities of interest are numerical:

- intelligence: IQ score
- SAT: SAT score
- Temperature,...

We can quantize them:

- intelligence: low = $IQ < 85$; avg: $85 < IQ < 115$

Or we can deal with them directly

- if we are talking about a scale, it's often easier to treat them as real numbers

Continuous random variables

If X is a real-valued random variable, what is the probability that $X=1.2, 1.21, 1.211, \dots$?

We can't assign probabilities to individual outcomes anymore.

But we can assign probabilities to the event that X falls into a specific interval:

What is the probability that $1.20 < X < 1.21$?

Probability density functions

A **probability density function** $P(X)$ (**PDF**) for a real-valued random variable X is a function such that

- $P(x) \geq 0$ for all $-\infty \leq x \leq +\infty$ (P is non-negative)
- $\int P(x)dx = 1$
- For any interval A : $P(x \in A) = \int_A P(x)dx$

The **cumulative distribution function** $F(n)$ is the probability that $X \leq n$: $F(n) = P(X \leq n) = \int^n P(x)dx$

Expectation (Mean)

The **expectation** $E(X)$ of a real-valued random variable X with PDF $P(X)$ is defined as

$$E(X) = \int xP(x)dx$$

$E(X)$ is also called the **mean** μ

The **expectation** $E(f(X))$ of a function of X with PDF $P(X)$ is defined as:

$$E(X) = \int f(x)P(x)dx$$

Variance

The **variance** $Var(X) = \sigma^2$ is defined as

$$Var(X) = E[(X - \mu)^2]$$

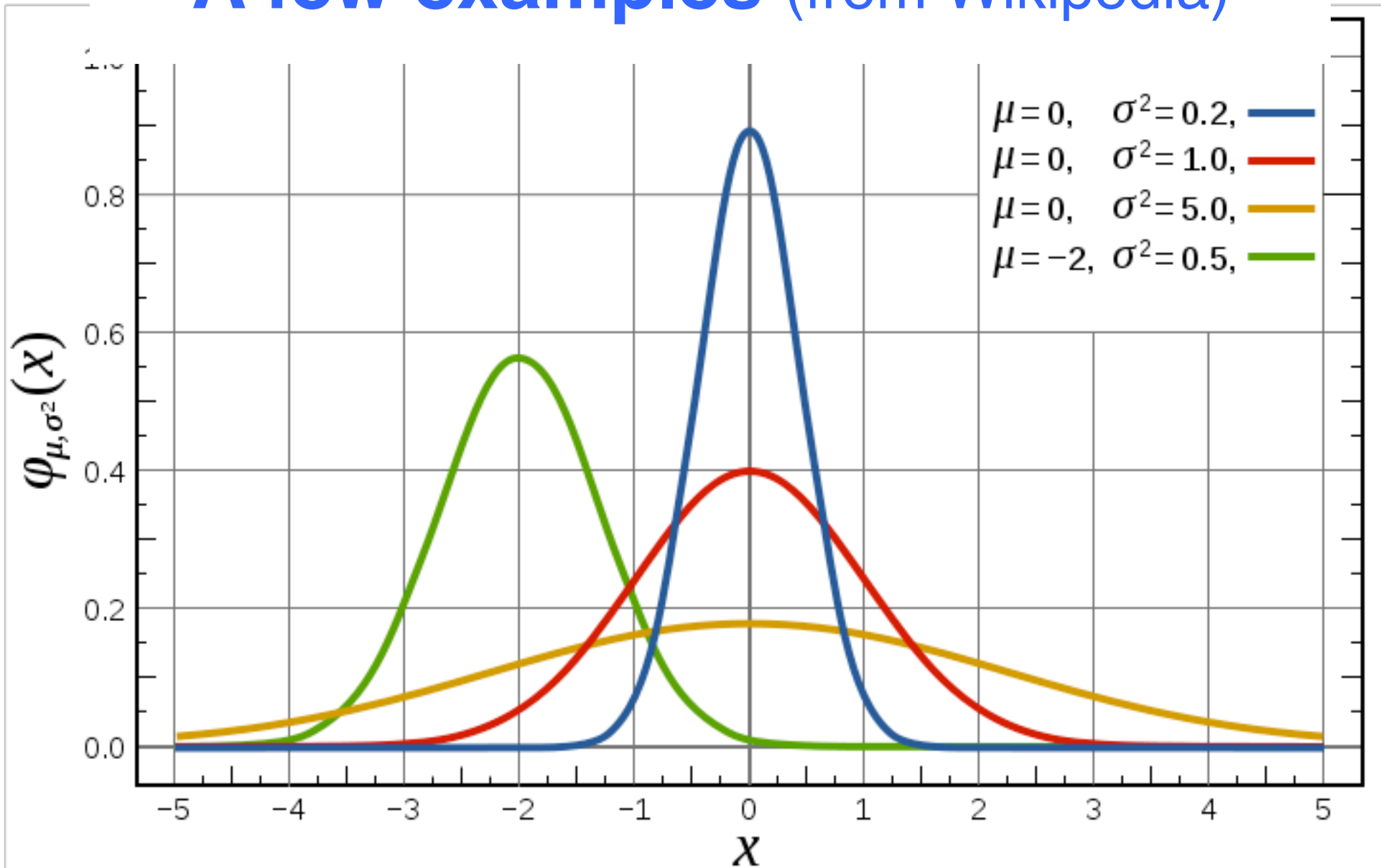
The **standard deviation** σ is the square root of the variance.

The normal (Gaussian) distribution

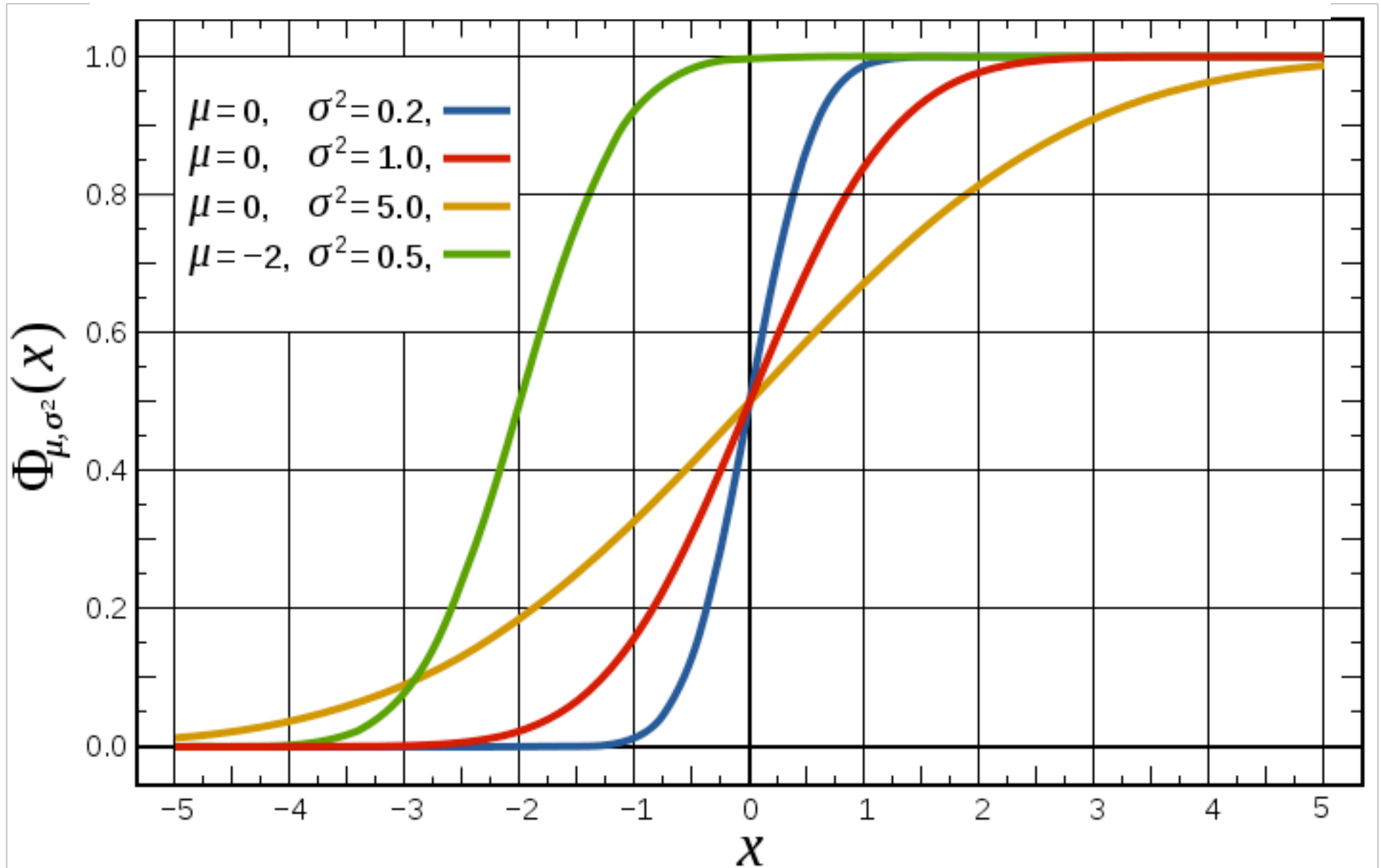
Gaussian (Normal) distribution with mean μ
and variance σ^2

$$P(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

A few examples (from Wikipedia)

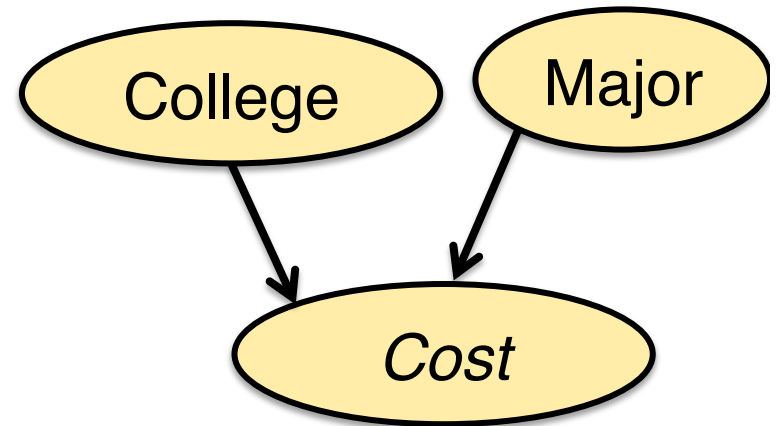


The CDFs (also from Wikipedia)



Continuous R.V. with discrete parents: a set of Gaussians

Each possible outcome of the set of discrete parents = one distribution



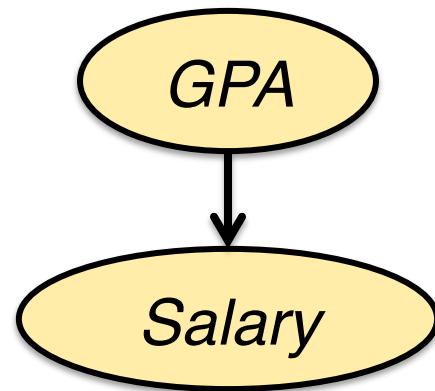
College	Major	Cost
U of I	CS	$N(\mu_{UofI,CS}, \sigma^2_{UofICS})$
U of I	English	$N(\mu_{UofI,E}, \sigma^2_{UofIE})$
UChicago	CS	$N(\mu_{UC,CS}, \sigma^2_{UC,CS})$

Continuous R.V.s with continuous parents: linear Gaussians

Salary is a Gaussian whose mean depends (linearly) on *GPA*: $\mu_{Salary} = a \times GPA + b$

We assume a fixed variance σ^2 .

$$P(Salary \mid GPA) = N(a \times GPA + b, \sigma^2)$$



$$P(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

Discrete (binary) R.V.s with continuous parents

The probability of getting a scholarship depends on your GPA:

- If your GPA is very low, it is close to zero.
- If your GPA is very high, it is close to one (you'd hope).
- Inbetween, the probability varies smoothly.

We need a threshold function

- Probit: CDF of Standard Normal (Gaussian with mean 0 and variance 1)
- Logit: uses logistic function $1/(1+e^{-x})$