

# COMP3702/7702 Artificial Intelligence Probability Review

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# How to quantify uncertainty ?

## Probability to the rescue...

Bertsekas & Tsitsiklis,  
Introduction to Probability.

- ▶ FATHER(F): Nurse, what is the probability that the drug will work?
- ▶ NURSE (N): I hope it works, we'll know tomorrow.
- ▶ F: Yes, but what is the probability that it will?
- ▶ N: Each case is different, we have to wait.
- ▶ F: But let's see, out of a hundred patients that are treated under similar conditions, how many times would you expect it to work?
- ▶ N (somewhat annoyed): I told you, every person is different, for some it works, for some it doesn't.
- ▶ F (insisting): Then tell me, if you had to bet whether it will work or not, which side of the bet would you take?
- ▶ N (cheering up for a moment): I'd bet it will work.
- ▶ F (somewhat relieved): OK, now, would you be willing to lose two dollars if it doesn't work, and gain one dollar if it does?
- ▶ N (exasperated): What a sick thought! You are wasting my time!



# Probabilistic Modeling

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- ▶ **View:**

- ▶ Experiments with random outcome.
- ▶ Quantifiable properties of the outcome.

- ▶ **Three components:**

- ▶ Sample space: Set of all possible outcomes.
- ▶ Events: Subsets of sample space.
- ▶ Probability: Quantify how likely an event occurs.



# Probability

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- ▶ Probability: A function that maps events to real numbers satisfying these axioms:
  1. Non-negativity:  $P(E) \geq 0$ , where  $E$  is an event.
  2. Normalization:  $P(S) = 1$ , where  $S$  is the sample space.
  3. Additivity of finite / countably infinite events.

$$P\left(\bigcup_{i=1}^{\infty/n} E_i\right) = \sum_{i=1}^{\infty/n} P(E_i),$$

where  $E_i$  are disjoint / mutually exclusive,  $i$ : natural number.



# Use of Probability Axioms

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- ▶ Linda is 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student, she was deeply concerned with issues of discrimination and social justice, and also participated in anti-nuclear demonstrations.  
Which is more probable?

1. Linda is a bank teller.
2. Linda is a bank teller and is active in the feminist movement.

Tversky & Kahneman.



# Conditional Probability

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- ▶ Sometimes, knowing something makes a difference.
- ▶ Model/reason about the outcome of an experiment, based on partial information.
- ▶ Given that event  $A$  occurs.
- ▶ The probability that event  $B$  also occurs:

$$P(B \mid A) = \frac{P(A \cap B)}{P(A)}$$

- ▶ Extendable to knowing multiple events.



# Chain Rule

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- ▶ Probability that two events occur:

$$P(A \cap B) = P(B | A)P(A)$$

- ▶ In general,

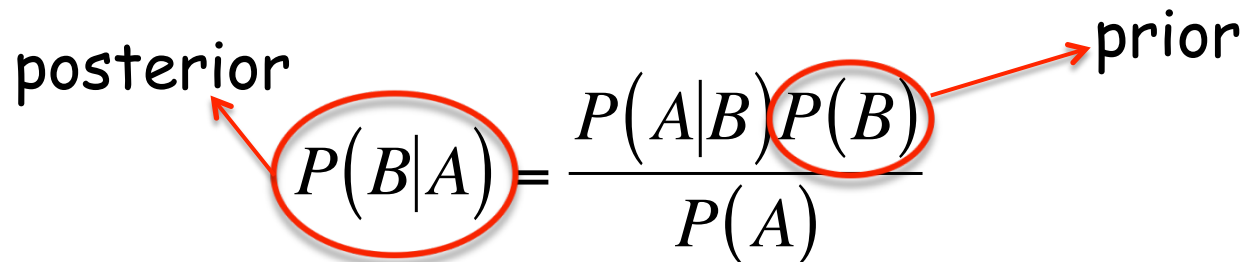
$$\begin{aligned} P\left(\bigcap_{i=1}^n A_i\right) &= P\left(A_1 \middle| \bigcap_{j=2}^n A_j\right) P\left(\bigcap_{j=2}^n A_j\right) \\ &= P\left(A_1 \middle| \bigcap_{j=2}^n A_j\right) P\left(A_2 \middle| \bigcap_{k=3}^n A_k\right) \dots P\left(A_{n-1} \middle| A_n\right) P\left(A_n\right) \end{aligned}$$



# Bayes Rule

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- ▶ Knowing  $P(A|B)$  maybe easier than knowing  $P(B|A)$ , e.g.,
  - ▶ Knowing symptoms of a disease is easier than figuring out the disease given the symptoms.



The diagram shows the Bayes' Rule formula:  $P(B|A) = \frac{P(A|B)P(B)}{P(A)}$ . The term  $P(B|A)$  is circled in red, with a red arrow pointing to it from the word "posterior" on the left. The term  $P(B)$  in the numerator is also circled in red, with a red arrow pointing to it from the word "prior" on the right.

$$\text{posterior } P(B|A) = \frac{P(A|B)P(B)}{P(A)} \text{ prior}$$





# Example

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- ▶ A cab was involved in a hit & run accident at night.
- ▶ Only 2 cabs company operate in the city, the Blue & the Green.
- ▶ 85% of the cabs in the city are Green.
- ▶ The court tested the reliability of the witness under the same circumstances that existed on the night of the evidence & concluded that the witness correctly identify the color of the taxi 80% of the time.
- ▶ A witness identified the cab is Blue.
- ▶ Is Blue cab more likely to be the one involved in the accident ?

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Tversky & Kahneman.



B : Blue cab was involved in the accident.

G : Green cab was involved in the accident.

$W_B$  : The witness says blue is the one involved in the accident.

$W_G$  : The witness says green is the one involved in the accident.

$$P(B) = 0.15 \quad ; \quad P(G) = 0.85$$

$$P(W_B|B) = P(W_G|G) = 0.8 \quad ; \quad P(W_B|G) = P(W_G|B) = 0.2$$

$$\begin{aligned} P(B|W_B) &= \frac{P(W_B|B)P(B)}{P(W_B)} \\ &= \frac{P(W_B|B)P(B)}{P(W_B \cap B) + P(W_B \cap G)} \\ &= \frac{P(W_B|B)P(B)}{P(W_B|B)P(B) + P(W_B|G)P(G)} \end{aligned}$$

The same derivation applies to  $P(G|W_B)$ . The answer to the question is blue if  $P(B|W_B) > P(G|W_B)$  and vice versa.

# Independence

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- ▶ When knowing something doesn't make a difference.
- ▶ Knowing event A occurs does not change the probability that event B occurs:

$$P(B|A) = P(B) \rightarrow P(A \cap B) = P(A)P(B)$$

- ▶ Chain rule becomes:

$$P\left(\bigcap_{i=1}^n A_i\right) = \prod_{i=1}^n P(A_i)$$



# Random Variables

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- ▶ Interest is on numerical values associated w. samples, e.g.:
  - ▶ Sample 50 students enrolled in COMP3702/7702, the number of students from mechatronics, IT, s/w eng., bioinf.
  - ▶ Roll a fair dice, get \$5 if the outcome is even, & loose \$5 if the outcome is odd.
- ▶ Random variable  $X$  is a function  $X : S \rightarrow Num$ .
  - ▶ Num: countable set (e.g., integer)  $\rightarrow$  discrete random variable.
  - ▶ Num: uncountable set (e.g., real)  $\rightarrow$  continuous random variable.



# Characterizing Random Variables

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- ▶ Cumulative distribution function (cdf)

$$F_X(x) = P(X \leq x) = P\left(\{s | X(s) \leq x, s \in S\}\right)$$

- ▶ Discrete: Probability mass function (pmf)

$$f_X[x] = P(X = x)$$

- ▶ Continuous: Probability density function/probability distribution function (pdf)

$$f_X(x) = \frac{dF_X(x)}{dx} \quad ; \quad P(a \leq X \leq b) = \int_a^b f_X(x) dx$$



# More Compact Characterization of Random Variables

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- Expectation: Weighted average of possible values of  $X$ , weight: probability.

$$E[X] = \sum_x x f_X[x] \quad ; \quad E[X] = \int_{-\infty}^{\infty} x f_X(x) dx$$

$$E[g(X)] = \sum_x g(x) f_X[x] \quad ; \quad E[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

$$\text{Linearity of Expectation: } E[aX + b] = aE[X] + b$$



# More Compact Characterization of Random Variables

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- ▶ Variance: A measure of dispersion around the mean.

$$\text{var}(X) = E\left[(X - E[X])^2\right] = E[X^2] - (E[X])^2$$

$$\text{var}(g(X)) = \sum_x g(X) f_X[x] \quad ; \quad g(X) = (X - E[X])^2$$

When  $g$  is linear:  $\text{var}(aX + b) = a^2 \text{var}(X)$

- ▶ Standard deviation:

$$\sigma(X) = \sqrt{\text{var}(X)}$$



# For more on probability...

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- ▶ Khan Academy (linked from <http://robotics.itee.uq.edu.au/~ai/doku.php?id=ai:resources> )
- ▶ STAT2202/STAT2203.
- ▶ ENGG7302.
- ▶ Introduction to Probability by Bertsekas & Tsitsiklis.

