

Probabilities

Alice Gao
Lecture 12

Based on work by K. Leyton-Brown, K. Larson, and P. van Beek

Outline

Learning Goals

Introduction to Probability Theory

Inferences Using the Joint Distribution

- The Sum Rule

- The Product Rule

Inferences using Prior and Conditional Probabilities

- The Chain Rule

- Bayes' Rule

Revisiting the Learning goals

Learning Goals

By the end of the lecture, you should be able to

- ▶ Calculate prior, posterior, and joint probabilities using the sum rule, the product rule, the chain rule and Bayes' rule.

Why handle uncertainty?

Why does an agent need to handle uncertainty?

- ▶ An agent may not observe everything in the world.
- ▶ An action may not have its intended consequences.

An agent needs to

- ▶ Reason about its uncertainty.
- ▶ Make a decision based on their uncertainty.

Probability

- ▶ Probability is the formal measure of uncertainty.
- ▶ There are two camps: Frequentists and Bayesians.
- ▶ **Frequentists' view of probability:**
 - ▶ Frequentists view probability as something *objective*.
 - ▶ Compute probabilities by counting the frequencies of events.
- ▶ **Bayesians' view of probability:**
 - ▶ Bayesians view probability as something *subjective*.
 - ▶ Probabilities are degrees of belief.
 - ▶ We start with **prior** beliefs and **update** beliefs based on new evidence.

Random variable

A random variable

- ▶ Has a **domain** of possible values
- ▶ Has an associated **probability distribution**, which is a function from the domain of the random variable to $[0, 1]$.

Example:

- ▶ random variable: The alarm is going.
- ▶ domain: {true, false}
- ▶ $P(\text{The alarm is going} = \text{true}) = 0.1$
 $P(\text{The alarm is going} = \text{false}) = 0.9$

Shorthand notation

Let A and B be Boolean random variables.

- ▶ $P(A)$ denotes $P(A = \textit{true})$.
- ▶ $P(\neg A)$ denotes $P(A = \textit{false})$.

Axioms of Probability

Let A and B be Boolean random variables.

- ▶ Every probability is between 0 and 1.

$$0 \leq P(A) \leq 1$$

- ▶ Necessarily true propositions have prob 1. Necessarily false propositions have probability 0.

$$P(\text{true}) = 1, P(\text{false}) = 0$$

- ▶ The inclusion-exclusion principle:

$$P(A \vee B) = P(A) + P(B) - P(A \wedge B)$$

These axioms limit the functions that can be considered as probability functions.

Joint Probability Distribution

- ▶ A **probabilistic model** contains a set of random variables.
- ▶ An **atomic event** assigns a value to every random variable in the model.
- ▶ A **joint probability distribution** assigns a probability to every atomic event.

Prior and Posterior Probabilities

$P(X)$:

- ▶ **prior** or **unconditional** probability
- ▶ Likelihood of X in the absence of any other information
- ▶ Based on the background information

$P(X|Y)$

- ▶ **posterior** or **conditional** probability
- ▶ Likelihood of X given Y .
- ▶ Based on Y as evidence

The Holmes Scenario

Mr. Holmes lives in a high crime area and therefore has installed a burglar alarm. He relies on his neighbors to phone him when they hear the alarm sound. Mr. Holmes has two neighbors, Dr. Watson and Mrs. Gibbon.

Unfortunately, his neighbors are not entirely reliable. Dr. Watson is known to be a tasteless practical joker and Mrs. Gibbon, while more reliable in general, has occasional drinking problems.

Mr. Holmes also knows from reading the instruction manual of his alarm system that the device is sensitive to earthquakes and can be triggered by one accidentally. He realizes that if an earthquake has occurred, it would surely be on the radio news.

Modeling the Holmes Scenario

What are the random variables?

How many probabilities are there in the joint probability distribution?

Learning Goals

Introduction to Probability Theory

Inferences Using the Joint Distribution

The Sum Rule

The Product Rule

Inferences using Prior and Conditional Probabilities

Revisiting the Learning goals

The Joint Distribution

	A		$\neg A$		
	G	$\neg G$		G	$\neg G$
W	0.032	0.048	W	0.036	0.324
$\neg W$	0.008	0.012	$\neg W$	0.054	0.486

The Sum Rule

Given a joint probability distribution, we can compute the probability over a subset of the variables.

CQ: Applying the sum rule

CQ: What is probability that
the alarm is **NOT** going and Dr. Watson is calling?

- (A) 0.36
- (B) 0.46
- (C) 0.56
- (D) 0.66
- (E) 0.76

CQ: Applying the sum rule

CQ: What is probability that
the alarm is going and Mrs. Gibbon is NOT calling?

- (A) 0.05
- (B) 0.06
- (C) 0.07
- (D) 0.08
- (E) 0.09

CQ: Applying the sum rule

CQ: What is probability that **the alarm is NOT going**?

(A) 0.1

(B) 0.3

(C) 0.5

(D) 0.7

(E) 0.9

The Product Rule

$$\forall x, y, P(X = x|Y = y) = \frac{P(X = x \wedge Y = y)}{P(Y = y)} \text{ whenever } P(Y = y) > 0$$

CQ: Calculating a conditional probability

CQ: What is probability that

Dr. Watson is calling given that the alarm is NOT going?

- (A) 0.2
- (B) 0.4
- (C) 0.6
- (D) 0.8
- (E) 1.0

CQ: Calculating a conditional probability

CQ: What is probability that

Mrs. Gibbon is NOT calling given that the alarm is going?

- (A) 0.2
- (B) 0.4
- (C) 0.6
- (D) 0.8
- (E) 1.0

Learning Goals

Introduction to Probability Theory

Inferences Using the Joint Distribution

Inferences using Prior and Conditional Probabilities

The Chain Rule

Bayes' Rule

Revisiting the Learning goals

The Prior and Conditional Probabilities

The prior probabilities:

$$P(A) = 0.1$$

The conditional probabilities

$$P(W|A) = 0.9$$

$$P(W|\neg A) = 0.4$$

$$P(W|A \wedge G) = 0.9$$

$$P(W|A \wedge \neg G) = 0.9$$

$$P(W|\neg A \wedge G) = 0.4$$

$$P(W|\neg A \wedge \neg G) = 0.4$$

$$P(G|A) = 0.3$$

$$P(G|\neg A) = 0.1$$

$$P(G|A \wedge W) = 0.3$$

$$P(G|A \wedge \neg W) = 0.3$$

$$P(G|\neg A \wedge W) = 0.1$$

$$P(G|\neg A \wedge \neg W) = 0.1$$

The Chain Rule

The chain rule for two variables (a.k.a. the product rule):

$$P(A \wedge B) = P(A|B) * P(B)$$

The chain rule for three variables:

$$P(A \wedge B \wedge C) = P(A|B \wedge C) * P(B|C) * P(C)$$

The chain rule can be generalized to any number of variables.

$$\begin{aligned} &P(X_n \wedge X_{n-1} \wedge \cdots \wedge X_2 \wedge X_1) \\ &= \prod_{i=1}^n P(X_i | X_{i-1} \wedge \cdots \wedge X_1) \\ &= P(X_n | X_{n-1} \wedge \cdots \wedge X_2 \wedge X_1) * \dots * P(X_2 | X_1) * P(X_1) \end{aligned}$$

CQ: Calculating the joint probability

CQ: What is probability that **the alarm is going, Dr. Watson is calling and Mrs. Gibbon is NOT calling?**

- (A) 0.060
- (B) 0.061
- (C) 0.062
- (D) 0.063
- (E) 0.064

CQ: Calculating the joint probability

CQ: What is probability that **the alarm is NOT going, Dr. Watson is NOT calling and Mrs. Gibbon is NOT calling?**

- (A) 0.486
- (B) 0.586
- (C) 0.686
- (D) 0.786
- (E) 0.886

Bayes' Rule

Definition (Bayes' rule)

$$P(X|Y) = \frac{P(Y|X) * P(X)}{P(Y)}.$$

Why is Bayes' rule useful?

Often you have causal knowledge:

- ▶ $P(\textit{symptom} \mid \textit{disease})$
- ▶ $P(\textit{alarm} \mid \textit{fire})$

...and you want to do evidential reasoning:

- ▶ $P(\textit{disease} \mid \textit{symptom})$
- ▶ $P(\textit{fire} \mid \textit{alarm})$.

CQ Applying the Bayes' rule

CQ: What is the probability that the alarm is **NOT** going given that Dr. Watson is calling?

- (A) 0.6
- (B) 0.7
- (C) 0.8
- (D) 0.9
- (E) 1.0

CQ Applying the Bayes' rule

CQ: What is the probability that the alarm is going given that Mrs. Gibbon is NOT calling?

- (A) 0.04
- (B) 0.05
- (C) 0.06
- (D) 0.07
- (E) 0.08

Revisiting the Learning Goals

By the end of the lecture, you should be able to

- ▶ Calculate prior, posterior, and joint probabilities using the sum rule, the product rule, the chain rule and Bayes' rule.