

# Reasoning Under Uncertainty: Introduction to Probability

CPSC 322 Lecture 23

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Textbook §9

# Lecture Overview

Recap

Probability Introduction

Syntax and Semantics of Probability

# Objects and Relations

- ▶ It is useful to view the world as consisting of **objects** and **relationships** between these objects.
- ▶ Often the propositions we spoke about before can be condensed into a much smaller number of propositions if they are allowed to express relationships between objects and/or functions of objects.
- ▶ Thus, reasoning in terms of objects and relationships can be simpler than reasoning in terms of features, as you can express more general knowledge using logical variables.

# Syntax of Datalog

- ▶ **variable** starts with upper-case letter.
- ▶ **constant** starts with lower-case letter or is a sequence of digits (numeral).
- ▶ **predicate symbol** starts with lower-case letter.
- ▶ **term** is either a variable or a constant.
- ▶ **atomic symbol** (atom) is of the form  $p$  or  $p(t_1, \dots, t_n)$  where  $p$  is a predicate symbol and  $t_i$  are terms.

# Syntax of Datalog (cont)

- ▶ **definite clause** is either an atomic symbol (a fact) or of the form:

$$\underbrace{a}_{\text{head}} \leftarrow \underbrace{b_1 \wedge \dots \wedge b_m}_{\text{body}}$$

where  $a$  and  $b_i$  are atomic symbols.

- ▶ **query** is of the form  $?b_1 \wedge \dots \wedge b_m$ .
- ▶ **knowledge base** is a set of definite clauses.

# Formal Semantics

An **interpretation** is a triple  $I = \langle D, \phi, \pi \rangle$ , where

- ▶  $D$ , the **domain**, is a nonempty set. Elements of  $D$  are **individuals**.
- ▶  $\phi$  is a mapping that assigns to each constant an element of  $D$ . Constant  $c$  **denotes** individual  $\phi(c)$ .
- ▶  $\pi$  is a mapping that assigns to each  $n$ -ary predicate symbol a relation: a function from  $D^n$  into  $\{TRUE, FALSE\}$ .

# Truth in an interpretation

A constant  $c$  **denotes in  $I$**  the individual  $\phi(c)$ .

Ground (variable-free) atom  $p(t_1, \dots, t_n)$  is

- ▶ **true in interpretation  $I$**  if  $\pi(p)(t'_1, \dots, t'_n) = \text{TRUE}$ , where  $t_i$  denotes  $t'_i$  in interpretation  $I$  and
- ▶ **false in interpretation  $I$**  if  $\pi(p)(t'_1, \dots, t'_n) = \text{FALSE}$ .

Ground clause  $h \leftarrow b_1 \wedge \dots \wedge b_m$  is **false in interpretation  $I$**  if  $h$  is false in  $I$  and each  $b_i$  is true in  $I$ , and is **true in interpretation  $I$**  otherwise.

# Variables

- ▶ Variables are **universally quantified** in the scope of a clause.
- ▶ A **variable assignment** is a function from variables into the domain.
- ▶ Given an interpretation and a variable assignment, each term denotes an individual and each clause is either true or false.
- ▶ A clause containing variables is true in an interpretation if it is true **for all** variable assignments.



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# Using Uncertain Knowledge

- ▶ Agents don't have complete knowledge about the world.
- ▶ Agents need to make decisions based on their uncertainty.
- ▶ It isn't enough to assume what the world is like.  
**Example:** wearing a seat belt.
- ▶ An agent needs to reason about its uncertainty.
- ▶ When an agent makes an action under uncertainty, it is gambling  $\implies$  probability.

# Probability

- ▶ Probability is formal measure of uncertainty. There are two camps:
- ▶ **Frequentists**: believe that probability represents something *objective*, and compute probabilities by counting the frequencies of different events
- ▶ **Bayesians**: believe that probability represents something *subjective*, and understand probabilities as degrees of belief.
  - ▶ They compute probabilities by starting with **prior beliefs**, and then **updating** beliefs when they get new data.
  - ▶ **Example**: Your degree of belief that a bird can fly is your measure of belief in the flying ability of an individual based only on the knowledge that the individual is a bird.
  - ▶ Other agents may have different probabilities, as they may have had different experiences with birds or different knowledge about this particular bird.
  - ▶ An agent's belief in a bird's flying ability is affected by what the agent knows about that bird.

# Numerical Measures of Belief

- ▶ Belief in proposition,  $f$ , can be measured in terms of a number between 0 and 1 — this is the **probability of  $f$** .
  - ▶ The probability  $f$  is 0 means that  $f$  is believed to be definitely false.
  - ▶ The probability  $f$  is 1 means that  $f$  is believed to be definitely true.
- ▶ Using 0 and 1 is purely a convention.
- ▶  $f$  has a probability between 0 and 1, doesn't mean  $f$  is true to some degree, but means you are ignorant of its truth value. Probability is a measure of your ignorance.

# Random Variables

- ▶ A **random variable** is a term in a language that can take one of a number of different values.
- ▶ The **domain** of a variable  $X$ , written  $dom(X)$ , is the set of values  $X$  can take.
- ▶ A tuple of random variables  $\langle X_1, \dots, X_n \rangle$  is a complex random variable with domain  $dom(X_1) \times \dots \times dom(X_n)$ . Often the tuple is written as  $X_1, \dots, X_n$ .
- ▶ Assignment  $X = x$  means variable  $X$  has value  $x$ .
- ▶ A **proposition** is a Boolean formula made from assignments of values to variables.

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# Possible World Semantics

- ▶ A **possible world** specifies an assignment of one value to each random variable.
- ▶  $w \models X = x$  means variable  $X$  is assigned value  $x$  in world  $w$ .
- ▶ Logical connectives have their standard meaning:

$$w \models \alpha \wedge \beta \text{ if } w \models \alpha \text{ and } w \models \beta$$

$$w \models \alpha \vee \beta \text{ if } w \models \alpha \text{ or } w \models \beta$$

$$w \models \neg\alpha \text{ if } w \not\models \alpha$$

- ▶ Let  $\Omega$  be the set of all possible worlds.

# Semantics of Probability: finite case

For a finite number of possible worlds:

- ▶ Define a nonnegative measure  $\mu(w)$  to each world  $w$  so that the measures of the possible worlds sum to 1.
- ▶ The measure specifies how much you think the world  $w$  is like the real world.
- ▶ The **probability** of proposition  $f$  is defined by:

$$P(f) = \sum_{w \models f} \mu(w).$$



# Axioms of Probability: finite case

Four axioms define what follows from a set of probabilities:

**Axiom 1**  $P(f) = P(g)$  if  $f \leftrightarrow g$  is a tautology. That is, logically equivalent formulae have the same probability.

**Axiom 2**  $0 \leq P(f)$  for any formula  $f$ .

**Axiom 3**  $P(\tau) = 1$  if  $\tau$  is a tautology.

**Axiom 4**  $P(f \vee g) = P(f) + P(g)$  if  $\neg(f \wedge g)$  is a tautology.

- ▶ These axioms are sound and complete with respect to the semantics.

# Semantics of Probability: general case

In the general case we have a measure on sets of possible worlds, satisfying:

- ▶  $\mu(S) \geq 0$  for  $S \subseteq \Omega$
- ▶  $\mu(\Omega) = 1$
- ▶  $\mu(S_1 \cup S_2) = \mu(S_1) + \mu(S_2)$  if  $S_1 \cap S_2 = \{\}$ .