

# Quantifying Uncertainty

## Belief States

### Definition

#### **Belief State**

A representation of all possible world states with likelihoods for all of them

- Expensive to compute
- Only likely to be used for small systems

## Decision Making

- Sometimes there is no "correct" answer or path
- Sometimes there is a **qualification problem** where there are factors hard to explain
- Decisions must still be made by the system based on rationality

## Summarizing Uncertainty

### Def

#### **Probability Theory**

A way to represent likelihoods of current states from a limited amount of information about the environment

- Explaining an effect from a list of causes can also be impossible
  - Must be represented with rules of uncertainty
  - Failure to correctly identify an effect could stem from
    - Laziness
    - Theoretical Ignorance
    - Practical Ignorance
- **Degrees of belief** must be used to represent likely effects from causes using probability theory

## Rational Decisions

### Def

#### **Utility Theory**

A way to represent preferences based on their **relative** usefulness to each other.

## Decision Theory

Probability Theory + Utility Theory

- Rational Decisions are ones use the **Maximum expected utility** of a state.

## Probability Lingo

### Def

#### Marginal Probability

Probability of a particular trait across all possible other trait combinations

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#### Marginalization

Summing individual probabilities to get a marginal probability

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#### Conditioning

Representing  $P(X \cap Y)$  as  $P(X|Y)P(Y)$

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#### Independence

When  $P(X|Y) = P(X)$

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#### Bayes' Rule

$$P(Y|X \cap Z) = \frac{P(X|Y \cap Z)P(Y|Z)}{P(X|Z)}$$

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#### Conditional Independence

In  $A$  and  $B$  where

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

but not necessarily

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

Said to be  $A$  and  $B$  are independent, only given  $C$

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## Naive Bayes' Rule

Assuming conditional independence across all other variables given one of them.

$$P(X_1, \dots, X_n) = P(X_1) \prod_{i=2}^n P(X_i | X_1)$$

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## Posterior, Likelihood, Prior, and Marginal Probabilities

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

Posterior:  $P(A|B)$

Likelihood:  $P(B|A)$

Prior:  $P(A)$

Marginal:  $P(B)$