



Probability Theory

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The lecture is based on the slides developed by Prof. Yu Zhang
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Objectives



Objective

Define
Probability
Space and
Conditional
Probability



Objective

Discuss Bayes
Rule

Uncertainty



| General situation:

- **Observed variables (evidence):** Agent knows certain things about the state of the world (e.g., sensor readings or symptoms)
- **Unobserved variables:** Agent needs to reason about other aspects (e.g., where an object is or what disease is present)
- **Model:** Agent knows something about how the known variables relate to the unknown variables

| Probabilistic reasoning gives us a framework for managing our beliefs and knowledge.

Probability Space (1/2)

| A probability space is a triplet (Ω, \mathcal{B}, P) that is used to model a process or an experiment with random outcomes.

| Ω : The sample space Ω is the set of all possible outcomes of an experiment.

- Consider two different experiments:
 - (1) Tossing a coin
 - (2) Tossing a die

Probability Space (2/2)

\mathcal{B} : a σ -algebra (or Borel field), or, informally, a collection of events to consider.

- **Event**: A subset of Ω , subject to some constraints (e.g., containing the empty set, being closed under complements and countable union)

P : A measure called "probability" defined on \mathcal{B} that satisfies these conditions:

- $P(A) \geq 0$ for all $A \in \mathcal{B}$
- $P(\Omega) = 1$
- If $A_1, A_2, \dots \in \mathcal{B}$ are pairwise disjoint, then $P(\cup A_i) = \sum P(A_i)$

Conditional Probability

| Let (Ω, \mathcal{B}, P) be a probability space, and let $H \in \mathcal{B}$, with $P(H) > 0$.

– For any $B \in \mathcal{B}$, we define:

$$P(B \mid H) = P(B \cap H) / P(H)$$

and call $P(B \mid H)$ the **conditional probability** of B , given H .

The Total Probability Rule

| Let (Ω, \mathcal{B}, P) be a probability space, and let $\{H_j\}$ be pairwise disjoint events in \mathcal{B} (i.e., let $H_j \cap H_k = \emptyset$, $\forall j \neq k$), and $\bigcup_{j=1, \dots, n} H_j = \Omega$. Let $A \in \mathcal{B}$.

- Such $\{H_j\}$ is called a **partition of Ω** and is finite or countably infinite.
- Suppose $P(H_j) > 0$, $\forall j$, then the total probability rule states:

$$P(A) = \sum_{j=1, \dots, n} P(A \cap H_j)$$

$$P(A) = \sum_{j=1, \dots, n} P(A \mid H_j) P(H_j)$$

The Product Rule

| Sometimes we have conditional distributions but we might want the joint distribution.

– Note: $P(x, y) = P(x \cap y)$

$$P(y) P(x \mid y) = P(x, y)$$

$$P(x \mid y) = P(x, y) / P(y)$$

The Chain Rule

| More generally, we can always write any joint distribution as an incremental product of conditional distributions.

$$P(x_1, x_2, x_3) = P(x_1)P(x_2 | x_1)P(x_3 | x_1, x_2)$$

$$P(x_1, x_2, \dots, x_n) = \prod_i P(x_i | x_1, \dots, x_{i-1})$$