



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

The Bayes Rule

| Let (Ω, \mathcal{B}, P) be a probability space, and let $\{H_j\}$ be pairwise disjoint events in \mathcal{B} , with $\bigcup_{j=1, \dots, n} H_j = \Omega$ and $P(H_j) > 0, \forall j$.

– Then $\forall B \in \mathcal{B}$ where $P(B) > 0$,

$$P(H_j | B) = P(H_j)P(B | H_j) / \sum_{i=1, \dots, n} P(H_i)P(B | H_i), \forall j$$

Bayes' Theorem



| Given events A and B , where $P(B) \neq 0$,
 $P(A|B) = P(B | A)P(A) / P(B)$.

The Bayes Rule



| Why is this at all helpful?

- Lets us build one conditional from its reverse
- Often one conditional is tricky but the other one is simple
- Foundation of many systems (e.g., tracking, localization)

| In the running for most important AI equation!

Independence of Events

| Let (Ω, \mathcal{B}, P) be a probability space, $\forall A, B \in \mathcal{B}$, we say A and B are independent if $P(A, B) = P(A)P(B)$.

- Splits the joint distribution factors into a product of two simple ones.
- Usually, variables are not independent! But we can use independence as a **modelling assumption**:
 - Independence can be a simplifying assumption.
 - Empirical joint distributions: at best "close" to independent.

Conditional Independence

| $P(\text{Toothache}, \text{Cavity}, \text{Catch})$

| If I have a cavity, the probability that the probe catches it does *not* depend on whether I have a toothache:

- $P(+\text{catch} \mid +\text{toothache}, +\text{cavity}) = P(+\text{catch} \mid +\text{cavity})$

| The same independence holds if I do not have a cavity:

- $P(+\text{catch} \mid +\text{toothache}, -\text{cavity}) = P(+\text{catch} \mid -\text{cavity})$

| "Catch" is conditionally independent of "Toothache" given "Cavity":

- $P(\text{Catch} \mid \text{Toothache}, \text{Cavity}) = P(\text{Catch} \mid \text{Cavity})$

Conditional Independence

| Unconditional (absolute) independence very rare (why?).

| Conditional independence is our most basic and robust form of knowledge about uncertain environments.

– X is conditionally independent of Y given Z (i.e., $X \perp\!\!\!\perp Y|Z$)

if and only if $\forall x,y,z: P(x,y | z) = P(x | z)P(y | z)$,

or, equivalently, if and only if $\forall x,y,z: P(x | z,y) = P(x | z)P(y | z)$.