

Probability

CS 470 Introduction To Artificial Intelligence

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Outline

- 1 Probability
 - Definition
 - Discrete variable
 - Continuous variable
- 2 Multiple variables
 - Joint distribution
 - Marginal distribution
 - Conditional distribution
- 3 Bayes
 - Independence
 - Bayes
 - Maximum likelihood
 - Maximum a posteriori



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Probability space

Probability space

- Ω - world physical states
- defining P
 - $P : 2^{\Omega} \implies [0, 1]$
 - for all $A \subseteq \Omega$, $P(A) \in [0, 1]$
 - $P(\Omega) = 1$

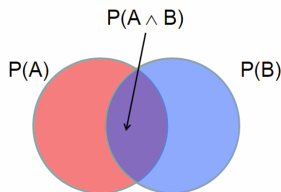


Probability space

Kolmogorvo's axioms

For any set of propositions (events) A , B

- $0 \leq P(A) \leq 1$
- $P(\text{True}) = 1$ and $P(\text{False}) = 0$
- $P(A \vee B) = P(A) + P(B) - P(A \wedge B)$

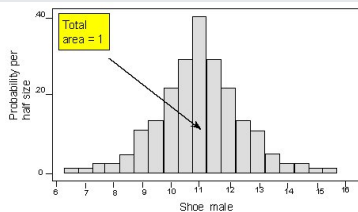




Type

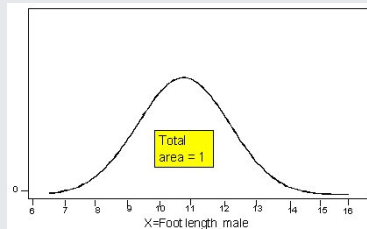
Discrete

Discrete random variable



Continuous

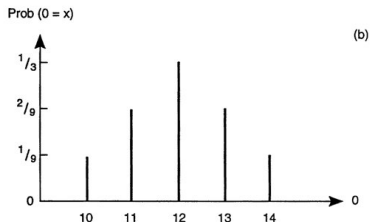
Continuous random variable





Probability mass function

- $\forall x \in S, P(X = x) = f(x) \geq 0$
- $\sum_{x \in S} f(x) = 1$
- $P(A) = \sum_{x \in A} f(x)$

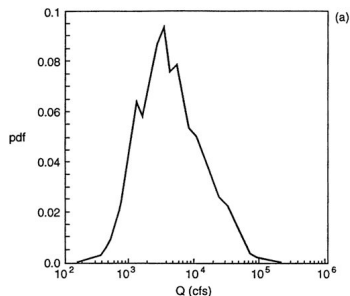




Continuous random variable

Probability density function

- $\forall x \in S, P(X = x) = f(x) \leq 0$
- $\int_{x \in S} f(x) = 1$
- $P(A) = \int_{x \in A} f(x)$

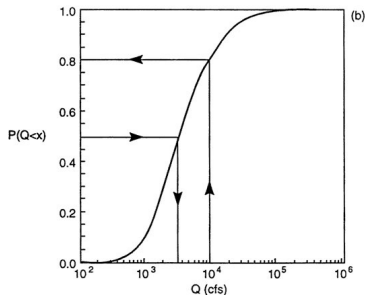




Continuous random variable

Cumulative density function

- $F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$
- $P(a < X \leq b) = \int_a^b f(t) dt = F(b) - F(a)$



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Joint distribution

Joint probability of $P(X_1, X_2, \dots, X_n)$

- $P(X_1 = x_1, X_2 = x_2, \dots, X_n = x_n) \in [0, 1]$
- $\sum P(X_1, X_2, \dots, X_n) = 1$

Marginal distribution

Relative to Joint distribution

Example:

$P(X)$ and $P(Y)$ are marginal distributions to the joint distribution $P(X, Y)$

discrete variable

$$P(X) = \sum_Y P(X, y)$$

continuous variable

$$P(X) = \int_Y P(X, y)$$

mixing discrete variable and continuous variable?



Marginal distribution

Example

P(Cavity, Toothache)	
<i>Cavity = false \wedge Toothache = false</i>	0.8
<i>Cavity = false \wedge Toothache = true</i>	0.1
<i>Cavity = true \wedge Toothache = false</i>	0.05
<i>Cavity = true \wedge Toothache = true</i>	0.05

P(Cavity)	
<i>Cavity = false</i>	?
<i>Cavity = true</i>	?

P(Toothache)	
<i>Toothache = false</i>	?
<i>Toothache = true</i>	?



Conditional distribution

- $P(A \mid B)$ is the probability of A given B

-

$$P(A \mid B) = \frac{P(A \wedge B)}{P(B)} = \frac{P(A, B)}{P(B)}$$

- B provides information \longrightarrow uncertainty reduce
- $P(A) \leq P(A \mid B)$



Conditional distribution

Example

P(Cavity, Toothache)	
<i>Cavity = false \wedge Toothache = false</i>	0.8
<i>Cavity = false \wedge Toothache = true</i>	0.1
<i>Cavity = true \wedge Toothache = false</i>	0.05
<i>Cavity = true \wedge Toothache = true</i>	0.05

P(Cavity)	
<i>Cavity = false</i>	0.9
<i>Cavity = true</i>	0.1

P(Toothache)	
<i>Toothache = false</i>	0.85
<i>Toothache = true</i>	0.15

- $P(\text{Cavity} = \text{true} \mid \text{Toothache} = \text{false})$
- $P(\text{Cavity} = \text{false} \mid \text{Toothache} = \text{true})$



Conditional distribution

discrete variable

$$\sum_X P(x | y) = 1$$

continuous variable

$$\int_X P(x | y) = 1$$

P(Cavity, Toothache)	
<i>Cavity = false ∧ Toothache = false</i>	0.8
<i>Cavity = false ∧ Toothache = true</i>	0.1
<i>Cavity = true ∧ Toothache = false</i>	0.05
<i>Cavity = true ∧ Toothache = true</i>	0.05



Select

Toothache, Cavity = false	
<i>Toothache = false</i>	0.8
<i>Toothache = true</i>	0.1



Renormalize

P(Toothache Cavity = false)	
<i>Toothache = false</i>	0.889
<i>Toothache = true</i>	0.111



Chain rule

From conditional probability

$$\begin{aligned} &P(A_1, A_2, \dots, A_n) \\ &= P(A_1)P(A_2 \mid A_1) \cdots P(A_n \mid A_1, \dots, A_{n-1}) \\ &= \prod_{i=1}^n P(A_i \mid A_1, \dots, A_{i-1}) \end{aligned}$$



Example : Toothache

	<i>toothache</i>		\neg <i>toothache</i>	
	<i>catch</i>	\neg <i>catch</i>	<i>catch</i>	\neg <i>catch</i>
<i>cavity</i>	.108	.012	.072	.008
\neg <i>cavity</i>	.016	.064	.144	.576

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



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Independence

- $P(A \wedge B) = P(A)P(B)$
- $\forall a \in A, b \in B, P(a \wedge b) = P(a)P(b)$
- Equivalence
 - $P(A | B) = P(A)$
 - $P(B | A) = P(B)$
- Coin flip



Independence

Example:

 $P_1(T, W)$

T	W	P
hot	sun	0.4
hot	rain	0.1
cold	sun	0.2
cold	rain	0.3

 $P(T)$

T	P
hot	0.5
cold	0.5

 $P_2(T, W)$

T	W	P
hot	sun	0.3
hot	rain	0.2
cold	sun	0.3
cold	rain	0.2

 $P(W)$

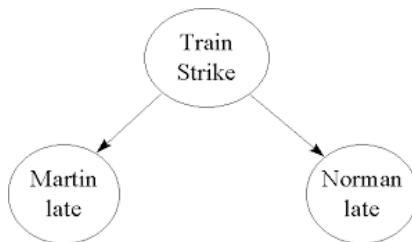
W	P
sun	0.6
rain	0.4



Conditional independence

- $P(A \wedge B \mid c) = P(A \mid c)P(B \mid c)$
- $\forall a \in A, b \in B, P(a \wedge b \mid c) = P(a \mid c)P(b \mid c)$

Example:





Bayes theorem

Bayes theorem

$$P(H | E) = \frac{P(E | H)P(H)}{P(E)}$$

- E - Evidence
- H - Hypothesis
- $P(H | E)$ - Posterior
- $P(H)$ - Priori
- $P(E | H)$ - Likelihood



ML estimation

Maximum likelihood estimate

ML

$$x^* = \arg \max_x P(e \mid x)$$

Example:

- $P(\text{slim} \mid \text{action} = \text{diet}) = 1$
- $P(\text{slim} \mid \text{action} = \text{drug}) = 0.5$

What should I do to be slim?



MAP estimation

Maximum a posteriori estimate

MAP

$$x^* = \arg \max_x P(x \mid e)$$

Example:

- $P(\text{slim} \mid \text{action} = \text{diet}) = 1$
- $P(\text{slim} \mid \text{action} = \text{drug}) = 0.5$
- $P(\text{action} = \text{diet}) = 0.1$
- $P(\text{action} = \text{drug}) = 0.9$
- $P(\text{slim}) = 0.5$

What should I do to be slim?



MAP estimation

Maximum a posteriori estimate

MAP

$$\begin{aligned}x^* &= \arg \max_X P(e | X)P(X)/P(e) \\ &= \arg \max_X P(e | X)P(X)\end{aligned}$$

Example:

- $P(\text{slim} \mid \text{action} = \text{diet}) = 1$
- $P(\text{slim} \mid \text{action} = \text{drug}) = 0.5$
- $P(\text{action} = \text{diet}) = 0.1$
- $P(\text{action} = \text{drug}) = 0.9$

What should I do to be slim?