

# PAI. Probability Tutorial

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# Probability

## Today:

- Conditional Probability
- Sum Rule and Product Rule
- Conditional independence
- Conditional and marginal distributions

## Conditional Probability

- Let  $X$  and  $Y$  - random variable with  $p(x)$  and  $p(y)$
- In general,  $p(x, y) \neq p(x)p(y)$   
If  $X$  and  $Y$  are **independent**,  $p(x, y) = p(x)p(y)$ .
- Conditional probability  $p(x|y) = \frac{p(x, y)}{p(y)}$   
Meaning: how the fact  $Y = y$  affects the distribution of  $X$ .  
Note, that  $\int p(x|y)dx \equiv 1$ , but  $\int p(x|y)dy$  is not essentially 1, because for  $y$  it is not a density function, but **likelihood function**.

## Sum Rule

- All the operations over probabilities are based on **sum rule** and **product rule**.
- Sum rule: Let  $X_1, \dots, X_k$  – are mutually excluding, one of which always happens. Then:

$$P(X_i \cup X_j) = P(X_i) + P(X_j) \quad \sum_{i=1}^k P(X_i) = 1$$

- In discrete case:  $\forall Y : \sum_{i=1}^k P(X_i|Y) = 1$ , then

$$\sum_{i=1}^k \frac{P(Y|X_i)P(X_i)}{P(Y)} = 1 \quad P(Y) = \sum_{i=1}^k P(Y|X_i)P(X_i)$$

- In continuous case:

$$p(y) = \int p(y, x) dx = \int p(y|x)p(x) dx$$

## Product Rule

- Product rule: any joint distribution can be represented as a product

$$p(x, y) = p(x|y)p(y) \quad P(X, Y) = P(X|Y)P(Y)$$

- Equivalent form:  $p(x|y)p(y) = p(x, y) = p(y|x)p(x)$
- For multidimensional joint distributions:

$$p(x_1, \dots, x_n) = p(x_1|x_2, \dots, x_n) \dots p(x_{n-1}|x_n)p(x_n)$$

- For a pdf of  $n$  variables, i.e.,  $p(x_1, x_2, \dots, x_n)$ , how many different factorizations exist?
- If the variables are all independent, how many different factorizations exist?

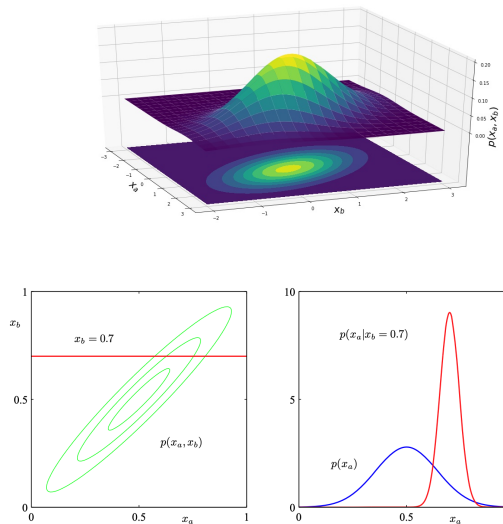
## Conditional and marginal distributions

- We are interested in distribution of  $x_a$  of a multivariate random variable  $(x_a, x_b)$ . How to do so?
- We know values  $p(x_b)$ : **conditional** distribution

$$p(x_a|x_b) = \frac{p(x_a, x_b)}{p(x_b)}$$

- We don't know values  $p(x_b)$ : **marginal** distribution

$$p(x_a) = \int p(x_a, x_b) dx_b$$



**Figure:** Upper: Joint distribution  $p(x_a, x_b)$  in 3D.  
 Lower: Joint distribution  $p(x_a, x_b)$ , conditional  $p(x_a | x_b)$ , marginal  $p(x_a)$   
 (Image: Bayesian methods, HSE, Vetrov)

## Bayes Rule

- $P(A|B) = \frac{P(B|A)P(A)}{P(B)}$
- 1% of women at age forty who participate in routine screening have breast cancer. 80% of women with breast cancer will get positive mammography. 9.6% of women without breast cancer will also get positive mammography.  
A women in this age group had a positive mammography in a routine screening. What is the probability that she has breast cancer?



## Bayes Rule

### Data

- $P(BC = T) = 1\%$ ,  $P(BC = F) = 99\%$ .
- $P(+|BC = T) = 80\%$ ,  $P(+|BC = F) = 9.6\%$ .

$$\begin{aligned}P(BC = T|+) &= \frac{P(BC = T, +)}{P(+)} \\&= \frac{P(BC = T, +)}{\sum_{BC=\{T,F\}} P(BC, +)} \\&= \frac{P(+|BC = T)P(BC = T)}{\sum_{BC=\{T,F\}} P(+|BC)P(BC)} \\&= \frac{80\% \times 1\%}{80\% \times 1\% + 9.6\% \times 99\%} = 7.76\%\end{aligned}$$

## Conditional Independence

Prove or disprove (by counterexample):

(a)  $X \perp Y|Z \Rightarrow X \perp Y$

False. Assume that:  $P(X = 1) = 0.1$ ,  $P(Y = 1) = 0.5$ ,  
 $P(Z = 1) = 0.5$ , and:

$P(\cdot Z = z)$	$Z = 0$	$Z = 1$
$X = 1$	0.4	0.3
$Y = 1$	0.6	0.2

$$P(X, Y) = \sum_Z P(X, Y, Z) = \sum_Z P(X|Z)P(Y|Z)P(Z)$$

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

$$P(X = 1, Y = 1) = 0.4 \times 0.6 \times 0.5 + 0.3 \times 0.2 \times 0.5 = 0.15$$

$$P(X = 1|Y = 1) = P(X = 1, Y = 1)/P(Y = 1) = 0.3 > P(X = 1)$$

## Conditional Independence

Prove or disprove (by counterexample):

$$(b) (X \perp Y|Z) \& (X \perp Z|Y) \Rightarrow X \perp (Y, Z)$$

$$\begin{aligned} P(X, Y, Z) &= P(X, Y|Z)P(Z) = P(X|Z)P(Y|Z)P(Z) \\ &= P(X|Z)P(Y, Z) \\ &= P(X, Z|Y)P(Y) = P(X|Y)P(Z|Y)P(Y) \\ &= P(X|Y)P(Y, Z) \\ &\Rightarrow P(X|Z) = P(X|Y) \end{aligned}$$

$$\begin{aligned} P(X|Z)P(Z)P(Y) &= P(X, Z)P(Y) = \\ P(X|Y)P(Y)P(Z) &= P(X, Y)P(Z) \end{aligned}$$

$$\underbrace{\sum_z P(X, Z = z) P(Y)}_{=P(X)} = P(X, Y) \underbrace{\sum_z P(Z = z)}_{=1} \Rightarrow X \perp Y$$

## Conditional Independence

- Does conditional independence implies independence?
- Does independence implies conditional independence?

## Conditional Independence

- Does conditional independence imply independence? (proved in the tutorial)
- Does independence imply conditional independence? (hw exercise)