

# Lecture 10. Conditional probability and expectation

Alex Avdiushenko

Neapolis University Paphos

Bonus-2023



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## Definition

Let  $\xi$  and  $\eta$  be two discrete random variables. Then

$$P(\xi = a \mid \eta = b) = \begin{cases} \frac{P(\xi=a, \eta=b)}{P(\eta=b)} \\ 0, \text{ if } P(\eta=b) = 0 \end{cases}$$

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Moreover

$$\sum_b \mathbb{E}(\xi \mid \eta = b) P(\eta = b) = \sum_{a,b} a P(\xi = a, \eta = b) = \sum_a a P(\xi = a) = \mathbb{E}\xi$$

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$$\begin{aligned} \mathbb{E}\xi &= \int_{\mathbb{R}} x p_{\xi, \eta}(x, y) dx dy = \int_{\mathbb{R}} dy \left( \int_{\mathbb{R}} x p_{\xi, \eta}(x, y) dx \right) = \int_{\mathbb{R}} \mathbb{E}(\xi | \eta = y) p_{\eta}(y) dy = \\ &= \mathbb{E}_{\eta}(\mathbb{E}(\xi | \eta = y)) \end{aligned}$$

# Example

Let  $(\Omega, P)$  is probability space and  $\xi : \Omega \rightarrow \mathbb{R}$ ,  $\chi_A : \Omega \rightarrow \mathbb{R}$  — indicator of  $A \subseteq \Omega$ . Find  $\mathbb{E}(\xi | \chi_A = y)$ .

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So

$$\mathbb{E}(\xi | \chi_A = y) = \frac{\mathbb{E}(\xi, \chi_A = y)}{P(\chi_A = y)}$$

# Continuous Version of Bayes Formula

## Definition

Let  $(\xi, \eta) \sim p_{\xi, \eta}(x, y)$ . Then conditional density

$$p_{\xi|\eta=y}(x) = p(x|y) = \frac{p(y|x)p(x)}{p(y)}$$

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

For  $\xi_1, \dots, \xi_n$  we have density  $p(x_1, \dots, x_n)$  and

$$p(x_1 \ x_2 \mid x_3 \ x_4) = \frac{p(x_1 \ x_2 \ x_3 \ x_4)}{p(x_3 \ x_4)}$$

and hence

$$p(x_1 \ x_2 \ x_3) = p(x_1 \mid x_2 \ x_3)p(x_2 \mid x_3)p(x_3)$$

# Task 1

Suppose we have a dice with  $n$  faces, results are uniformly distribute. We roll the dice until the sum of all the rolled results is greater than  $n$ . So

$$\Omega = \left\{ (a_1, \dots, a_k) \mid \sum_{i=1}^k a_i \geq n, \sum_{i=1}^{k-1} a_i < n \right\}$$

$$P(a_1, \dots, a_k) = \frac{1}{n^k}$$

$$\xi(a_1, \dots, a_k) = k \text{ — number of dice rolls, } \mathbb{E}\xi = ?$$

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Consider new random variable  $\eta_1$  = result of first dice roll

$$\mathbb{E}\xi = \mathbb{E}_{\eta_1}(\mathbb{E}(\xi \mid \eta_1 = a)) = \sum_{k=1}^n \mathbb{E}(\xi \mid \eta_1 = k) P(\eta_1 = k)$$



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Now consider random variables

$\xi_m$  = number of dice rolls to get  $m$  in sum, so  $\xi = \xi_n$

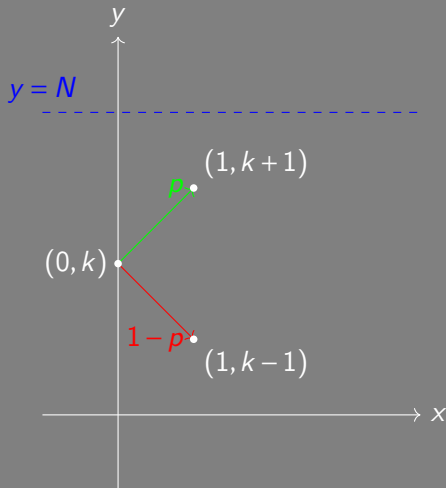
$$\begin{aligned}
 \mathbb{E}(\xi_n \mid \eta_1 = k) &= \mathbb{E}(1 + \xi_{n-k} \mid \eta_1 = k) = \{\text{as } \xi_{n-k} \perp\!\!\!\perp \eta_1\} = \mathbb{E}(1 + \xi_{n-k}) \\
 \Rightarrow \mathbb{E}\xi_m &= \frac{\mathbb{E}(1 + \xi_{m-1}) + \mathbb{E}(1 + \xi_{m-2}) + \cdots + \mathbb{E}(1 + \xi_{m-n})}{n} = \\
 &= \frac{n + \mathbb{E}\xi_{m-1} \cdots + \mathbb{E}\xi_1}{n}
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\end{aligned}$$

$$\text{If } a_k = \mathbb{E}\xi_k, \quad a_1 = \mathbb{E}\xi_1 = 1$$

$$a_k = 1 + \frac{a_1 + \cdots + a_{k-1}}{n} \Rightarrow a_k - a_{k-1} = \frac{a_{k-1}}{n} \Rightarrow a_k = \left(1 + \frac{1}{n}\right) a_{k-1} = \left(1 + \frac{1}{n}\right)^{k-1}$$

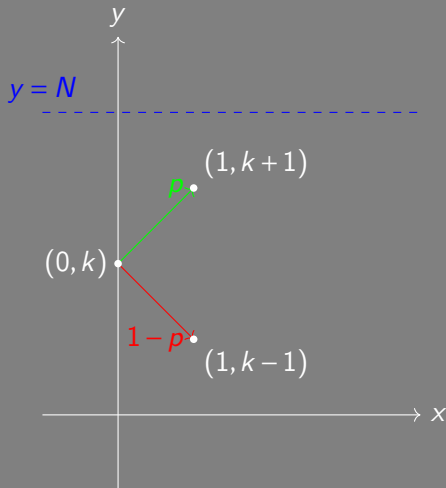
## Task 2: casino



If you are lucky to score  $N$ , then you win and leave. If you lose all the money to zero, then the game also ends.

What is  $\Omega$  here? All the paths!

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### Question

Find the average length of game  $\mathbb{E}\xi_k$ .

$$\mathbb{E}\xi_0 = \mathbb{E}\xi_N = 0$$

$$\begin{aligned}\mathbb{E}\xi_k &= \mathbb{E}(\xi_k \mid \text{win})P(\text{win}) + \mathbb{E}(\xi_k \mid \text{lose})P(\text{lose}) = \\ &= \mathbb{E}(1 + \xi_{k+1} \mid \text{win})p + \mathbb{E}(1 + \xi_{k-1} \mid \text{lose})q = \\ &= 1 + p\mathbb{E}(\xi_{k+1}) + q\mathbb{E}(\xi_{k-1})\end{aligned}$$

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Let column  $x = (x_0, \dots, x_N) = (\mathbb{E}\xi_0, \dots, \mathbb{E}\xi_N)$

$$\begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ \vdots \\ x_{N-1} \\ x_N \end{bmatrix} = \begin{bmatrix} 0 & 0 & \dots & \dots & 0 & 0 \\ q & 0 & p & 0 & \dots & 0 \\ 0 & q & 0 & p & \dots & 0 \\ 0 & 0 & \ddots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & q & 0 & p \\ 0 & 0 & \dots & \dots & 0 & 0 \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ \vdots \\ x_{N-1} \\ x_N \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ \vdots \\ \vdots \\ 1 \\ 0 \end{bmatrix}$$

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$$x = Ax + b \Rightarrow \boxed{x = (I - A)^{-1}b}$$



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We can find it only for independent random variables  $\xi \perp\!\!\!\perp \eta$ , or we must know joint probability distribution.

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In discrete case

$$P(\xi + \eta = a) = \sum_{u+v=a} P(\xi = u)P(\eta = v)$$

And for integer values we have convolution

$$P(\xi + \eta = k) = \sum_{i \in \mathbb{Z}} p_i q_{k-i}$$

For continuous case we have  $\xi \sim p(x)$ ,  $\eta \sim q(y)$

$$\begin{aligned} F_{\xi+\eta}(t) &= P(\xi + \eta \leq t) = \int_{x+y \leq t} p(x)q(y) dx dy = \\ &= \{x + y = u, y = v\} = \int_{u \leq t} p(u-v)q(v) \det \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} du dv = \\ &= \int_{-\infty}^t du \int_{\mathbb{R}} p(u-v)q(v) dv \end{aligned}$$

$$F'_{\xi+\eta}(t) = p_{\xi+\eta}(t) = \int_{\mathbb{R}} p(t-v)q(v) dv = \int_{\mathbb{R}} p(u)q(t-u) du$$