

## Lecture 02: Mathematical Basics (Probability)

# Probability Basics

- Sample Space:  $\Omega$  is a set of outcomes (it can either be finite or infinite)
- Random Variable:  $\mathbb{X}$  is a random variable that assigns probabilities to outcomes

Example: Let  $\Omega = \{\text{Heads}, \text{Tails}\}$ . Let  $\mathbb{X}$  be a random variable that outputs Heads with probability  $1/3$  and outputs Tails with probability  $2/3$

- The probability that  $\mathbb{X}$  assigns to the outcome  $x$  is represented by

$$\mathbb{P}[\mathbb{X} = x]$$

Example: In the ongoing example  $\mathbb{P}[\mathbb{X} = \text{Heads}] = 1/3$ .

# Function of a Random Variable

- Let  $f: \Omega \rightarrow \Omega'$  be a function
- Let  $\mathbb{X}$  be a random variable over the sample space  $\mathbb{X}$
- We define a new random variable  $f(\mathbb{X})$  is over  $\Omega'$  as follows

$$\mathbb{P} [f(\mathbb{X}) = y] = \sum_{x \in \Omega: f(x)=y} \mathbb{P} [\mathbb{X} = x]$$

# Joint Distribution and Marginal Distributions I

- Suppose  $(\mathbb{X}_1, \mathbb{X}_2)$  is a random variable over  $\Omega_1 \times \Omega_2$ .
  - Intuitively, the random variable  $(\mathbb{X}_1, \mathbb{X}_2)$  takes values of the form  $(x_1, x_2)$ , where the first coordinate lies in  $\Omega_1$ , and the second coordinate lies in  $\Omega_2$

For example, let  $(\mathbb{X}_1, \mathbb{X}_2)$  represent the temperatures of West Lafayette and Lafayette. Their sample space is  $\mathbb{Z} \times \mathbb{Z}$ . Note that these two outcomes can be correlated with each other.

## Joint Distribution and Marginal Distributions II

- Let  $P_1: \Omega_1 \times \Omega_2 \rightarrow \Omega_1$  be the function  $P_1(x_1, x_2) = x_1$  (the projection operator)
- So, the random variable  $P_1(\mathbb{X}_1, \mathbb{X}_2)$  is a probability distribution over the sample space  $\Omega_1$
- This is represented simply as  $\mathbb{X}_1$ , the marginal distribution of the first coordinate
- Similarly, we can define  $\mathbb{X}_2$

# Conditional Distribution

- Let  $(\mathbb{X}_1, \mathbb{X}_2)$  be a joint distribution over the sample space  $\Omega_1 \times \Omega_2$
- We can define the distribution  $(\mathbb{X}_1 \mid \mathbb{X}_2 = x_2)$  as follows
  - This random variable is a distribution over the sample space  $\Omega_1$
  - The probability distribution is defined as follows

$$\mathbb{P}[\mathbb{X}_1 = x_1 \mid \mathbb{X}_2 = x_2] = \frac{\mathbb{P}[\mathbb{X}_1 = x_1, \mathbb{X}_2 = x_2]}{\sum_{x \in \Omega_1} \mathbb{P}[\mathbb{X}_1 = x, \mathbb{X}_2 = x_2]}$$

For example, conditioned on the temperature at Lafayette being 0, what is the conditional probability distribution of the temperature in West Lafayette?

## Theorem (Bayes' Rule)

Let  $(\mathbb{X}_1, \mathbb{X}_2)$  be a joint distribution over the sample space  $(\Omega_1, \Omega_2)$ .  
Let  $x_1 \in \Omega_1$  and  $x_2 \in \Omega_2$  be such that  $\mathbb{P}[\mathbb{X}_1 = x_1, \mathbb{X}_2 = x_2] > 0$ .  
Then, the following holds.

$$\mathbb{P}[\mathbb{X}_1 = x_1 \mid \mathbb{X}_2 = x_2] = \frac{\mathbb{P}[\mathbb{X}_1 = x_1, \mathbb{X}_2 = x_2]}{\mathbb{P}[\mathbb{X}_2 = x_2]}$$

The random variables  $\mathbb{X}_1$  and  $\mathbb{X}_2$  are independent of each other if the distribution  $(\mathbb{X}_1 \mid \mathbb{X}_2 = x_2)$  is identical to the random variable  $\mathbb{X}_1$ , for all  $x_2 \in \Omega_2$  such that  $\mathbb{P}[\mathbb{X}_2 = x_2] > 0$

We can generalize the Bayes' Rule as follows.

## Theorem (Chain Rule)

*Let  $(\mathbb{X}_1, \mathbb{X}_2, \dots, \mathbb{X}_n)$  be a joint distribution over the sample space  $\Omega_1 \times \Omega_2 \times \dots \times \Omega_n$ . For any  $(x_1, \dots, x_n) \in \Omega_1 \times \dots \times \Omega_n$  we have*

$$\mathbb{P}[\mathbb{X}_1 = x_1, \dots, \mathbb{X}_n = x_n] = \prod_{i=1}^n \mathbb{P}[\mathbb{X}_i = x_i \mid \mathbb{X}_{i-1} = x_{i-1} \dots, \mathbb{X}_1 = x_1]$$



## Important: Why use Bayes' Rule I

In which context do we foresee to use the Bayes' Rule to compute joint probability?

- Sometimes, the problem at hand will clearly state how to sample  $\mathbb{X}_1$  and then, conditioned on the fact that  $\mathbb{X}_1 = x_1$ , it will state how to sample  $\mathbb{X}_2$ . In such cases, we shall use the Bayes' rule to calculate

$$\mathbb{P}[\mathbb{X}_1 = x_1, \mathbb{X}_2 = x_2] = \mathbb{P}[\mathbb{X}_1 = x_1] \mathbb{P}[\mathbb{X}_2 = x_2 | \mathbb{X}_1 = x_1]$$

- Let us consider an example.
  - Suppose  $\mathbb{X}_1$  is a random variable over  $\Omega_1 = \{0, 1\}$  such that  $\mathbb{P}[\mathbb{X}_1 = 0] = 1/2$ . Next, the random variable  $\mathbb{X}_2$  is over  $\Omega_2 = \{0, 1\}$  such that  $\mathbb{P}[\mathbb{X}_2 = x_1 | \mathbb{X}_1 = x_1] = 2/3$ . Note that  $\mathbb{X}_2$  is biased towards the outcome of  $\mathbb{X}_1$ .
  - What is the probability that we get  $\mathbb{P}[\mathbb{X}_1 = 0, \mathbb{X}_2 = 1]$ ?

## Important: Why use Bayes' Rule II

- To compute this probability, we shall use the Bayes' rule.

$$\mathbb{P}[\mathbb{X}_1 = 0] = 1/2$$

Next, we know that

$$\mathbb{P}[\mathbb{X}_2 = 0 | \mathbb{X}_1 = 0] = 2/3$$

Therefore, we have  $\mathbb{P}[\mathbb{X}_2 = 1 | \mathbb{X}_1 = 0] = 1/3$ . So, we get

$$\begin{aligned}\mathbb{P}[\mathbb{X}_1 = 0, \mathbb{X}_2 = 1] &= \mathbb{P}[\mathbb{X}_1 = 0] \mathbb{P}[\mathbb{X}_2 = 1 | \mathbb{X}_1 = 0] \\ &= (1/2) \cdot (1/3) = 1/6\end{aligned}$$

# Independence of Random Variables

- Consider a joint distribution  $(X_1, X_2)$  over the sample space  $\Omega_1 \times \Omega_2$
- The marginal distributions  $X_1$  and  $X_2$  are independent of each other, if for all  $x_1 \in \Omega_1$  and  $x_2 \in \Omega_2$  we have: If  $\mathbb{P}[X_2 = x_2] > 0$  then

$$\mathbb{P}[X_1 = x_1] = \mathbb{P}[X_1 = x_1 | X_2 = x_2].$$

- Equivalently, the following condition is satisfied

$$\mathbb{P}[X_1 = x_1] \cdot \mathbb{P}[X_2 = x_2] = \mathbb{P}[X_1 = x_1, X_2 = x_2].$$

# Probability: First Example I

- Let  $\mathbb{S}$  be the random variable representing whether I studied for my exam. This random variable has sample space  $\Omega_1 = \{Y, N\}$
- Let  $\mathbb{P}$  be the random variable representing whether I passed my exam. This random variable has sample space  $\Omega_2 = \{Y, N\}$
- Our sample space is  $\Omega = \Omega_1 \times \Omega_2$
- The joint distribution  $(\mathbb{S}, \mathbb{P})$  is represented in the next page

# Probability: First Example II

$s$	$p$	$\mathbb{P}[S = s, P = p]$
Y	Y	$1/2$
Y	N	$1/4$
N	Y	$0$
N	N	$1/4$

## Probability: First Example III

Here are some interesting probability computations  
The probability that I pass.

$$\begin{aligned}\mathbb{P}[\mathbb{P} = \mathbb{Y}] &= \mathbb{P}[\mathbb{S} = \mathbb{Y}, \mathbb{P} = \mathbb{Y}] + \mathbb{P}[\mathbb{S} = \mathbb{N}, \mathbb{P} = \mathbb{Y}] \\ &= 1/2 + 0 = 1/2\end{aligned}$$

## Probability: First Example IV

The probability that I study.

$$\begin{aligned}\mathbb{P}[S = Y] &= \mathbb{P}[S = Y, P = Y] + \mathbb{P}[S = Y, P = N] \\ &= 1/2 + 1/4 = 3/4\end{aligned}$$

## Probability: First Example V

The probability that I pass conditioned on the fact that I studied.

$$\begin{aligned}\mathbb{P}[P = Y \mid S = Y] &= \frac{\mathbb{P}[P = Y, S = Y]}{\mathbb{P}[S = Y]} \\ &= \frac{1/2}{3/4} = \frac{2}{3}\end{aligned}$$



## Probability: Second Example I

- Let  $\mathbb{T}$  be the time of the day that I wake up. The random variable  $\mathbb{T}$  has sample space  $\Omega_1 = \{4, 5, 6, 7, 8, 9, 10\}$
- Let  $\mathbb{B}$  represent whether I have breakfast or not. The random variable  $\mathbb{B}$  has sample space  $\Omega_2 = \{T, F\}$
- Our sample space is  $\Omega = \Omega_1 \times \Omega_2$
- The joint distribution of  $(\mathbb{T}, \mathbb{B})$  is presented on the next page

## Probability: Second Example II

$t$	$b$	$\mathbb{P}[T = t, \mathbb{B} = b]$
4	T	0.03
4	F	0
5	T	0.02
5	F	0
6	T	0.30
6	F	0.05
7	T	0.20
7	F	0.10
8	T	0.10
8	F	0.08
9	T	0.05
9	F	0.05
10	T	0
10	F	0.02

- What is the probability that I have breakfast conditioned on the fact that I wake up at or before 7?

Formally, what is  $\mathbb{P} [\mathbb{B} = \mathbb{T} \mid \mathbb{T} \leq 7]$ ?