

**Slides**  
**Probability Theory**  
**Chapter 3**

Ch. 3: Conditional Probability  
and Independence

# Overview Basic Probability in Ch. 1+2

**Axioms of Kolmogorov: 1.  $P(A) \geq 0$**

**2.  $P(S) = 1$**

**3.  $P(\cup_i A_i) = \sum_i P(A_i)$ ,**

if the  $A_i$ 's are mutually exclusive.

**Complement rule:  $P(\bar{A}) = 1 - P(A)$**

**Addition Rule:  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$**

For disjoint events we have:  $P(A \cap B) = 0$

**Symmetric probability space (Laplace):  $P(A) = \frac{N(A)}{N(S)}$**

**number of permutations of  $n$  out of  $N$  is  $\frac{N!}{(N-n)!}$**

**number of combinations of  $n$  out of  $N$  is  $\binom{N}{n} = \frac{N!}{n!(N-n)!}$**

# Conditional probability

“Probabilities within a part  $A$  of the sample space  $S$ ”

**Definition:** 
$$P(B|A) = \frac{P(A \cap B)}{P(A)},$$
  
if  $P(A) > 0$

From this definition the **Product rule** follows:

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

or: 
$$P(A \cap B) = P(A|B)P(B)$$

“**Conditioning w.r.t.  $A$  and  $B$ , respectively**”

Product rule for  $n$  events:

$$P(A_1 A_2 \dots A_n) = P(A_1)P(A_2|A_1) \dots P(A_n|A_1 \dots A_{n-1})$$

$(S, P(. | A))$  is a probability space (Kolmogorov)

# An application of Bayes' Rule

Success proportions at the Dutch HBO-colleges	<b>Previous education</b>	<b>MBO</b>	<b>HAVO</b>	<b>VWO</b>
	Proportion students	<b>20%</b>	<b>50%</b>	<b>30%</b>
	Success proportion	<b>80%</b>	<b>60%</b>	<b>90%</b>

Answer the following questions:

1. What is the overall success proportion at the HBO?
2. Determine the probability that a HBO-graduate had his previous training at VWO

# Solution

- **Define:**  $A$  = “student succeeds”  
 $S_1$  = “previous training is MBO”  
 $S_2$  and  $S_3$  similar for HAVO and VWO resp.
- **Probabilities:**  $P(S_1)=0.20$ ,  $P(S_2)=0.50$ ,  $P(S_3)=0.30$ ,  
 $P(A/S_1) = 0.80$ ,  $P(A/S_2) = 0.60$  and  $P(A/S_3) = 0.90$
- **Apply the rules** of total probability and Bayes:

$$\begin{aligned} 1. \quad P(A) &= P(A \cap S_1) + P(A \cap S_2) + P(A \cap S_3) \\ &= P(A|S_1)P(S_1) + P(A|S_2)P(S_2) + P(A|S_3)P(S_3) \\ &= 0.80 \times 0.20 + 0.60 \times 0.50 + 0.90 \times 0.30 = \underline{\underline{0.73}} \end{aligned}$$

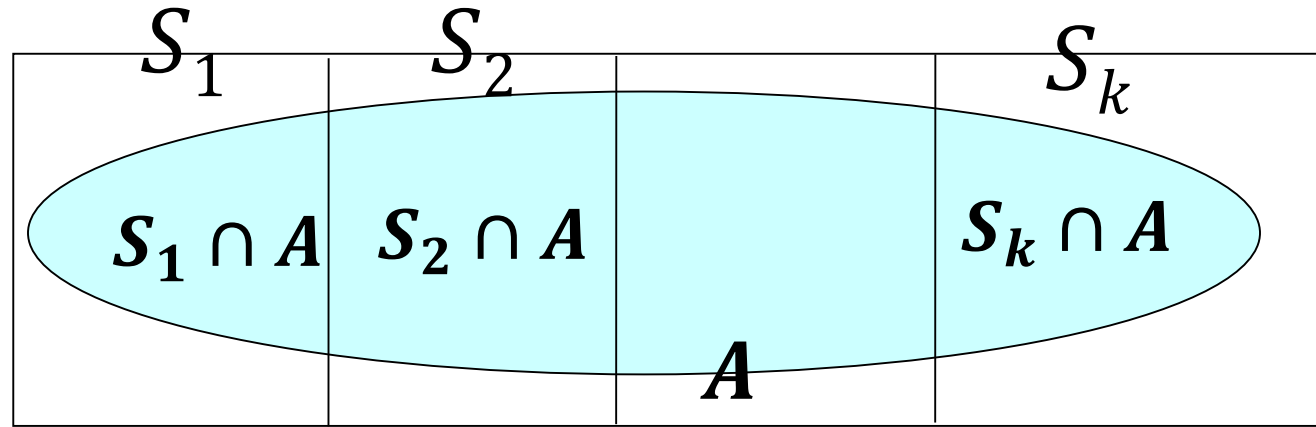
$$2. \quad P(S_3|A) = \frac{P(S_3 \cap A)}{P(A)} = \frac{P(A|S_3)P(S_3)}{P(A)} = \frac{0.9 \times 0.3}{0.73} \approx \underline{\underline{0.37}}$$

# Rules of total probability and Bayes

Given:

- $P(S_i)$  and  $P(A|S_i)$

• Then:



$$P(A) = P(A \cap S_1) + \dots + P(A \cap S_k)$$

$$= P(A|S_1)P(S_1) + \dots + P(A|S_k)P(S_k)$$

Law of total probability

And:

$$P(S_1|A) = \frac{P(S_1 \cap A)}{P(A)}$$

**Bayes' rule**

$$= \frac{P(A|S_1)P(S_1)}{P(A|S_1)P(S_1) + \dots + P(A|S_k)P(S_k)}$$

# Independence of events $A$ and $B$ :

“ $B$  does not give information about (the probability of)  $A$ ”

$$P(A|B) = P(A) \Leftrightarrow \frac{P(A \cap B)}{P(B)} = P(A) \Leftrightarrow P(B|A) = P(B)$$

**Definition:**  $A$  and  $B$  are **independent**,

$$\text{if } P(A \cap B) = P(A) \cdot P(B)$$

$A$ ,  $B$  and  $C$  are independent if  $P(A \cap B) = P(A) \cdot P(B)$

$$P(A \cap C) = P(A) \cdot P(C)$$

$$P(B \cap C) = P(B) \cdot P(C)$$

$$\text{and } P(A \cap B \cap C) = P(A) \cdot P(B) \cdot P(C)$$

$A_1, \dots, A_n$  are **independent** if the equality holds for any subsequence  $A_{i_1}, \dots, A_{i_k}$  ( $k = 2, \dots, n$ )

# Independent experiments

Experiments are said to be independent if events related to different experiments are independent.

Usually experiments are **assumed** to be independent.

E.g. **Bernoulli experiments** or **Bernoulli trials**:

1. A sequence of **independent** experiments.  
(Repetitions of the same experiment)
2. **Two possible outcomes**: “Success” and “Failure”
3. Constant “Success”-probability  $p$ .

## Application 1 (**geometric** formula)

$X$  = the required number of Bernoulli trials until the first success occurs.

$$P(X = k) = (1 - p)^{k-1} p, \text{ where } k = 1, 2, \dots$$

## Application 2 (**binomial** formula)

$X$  = the number of successes in  $n$  Bernoulli trials

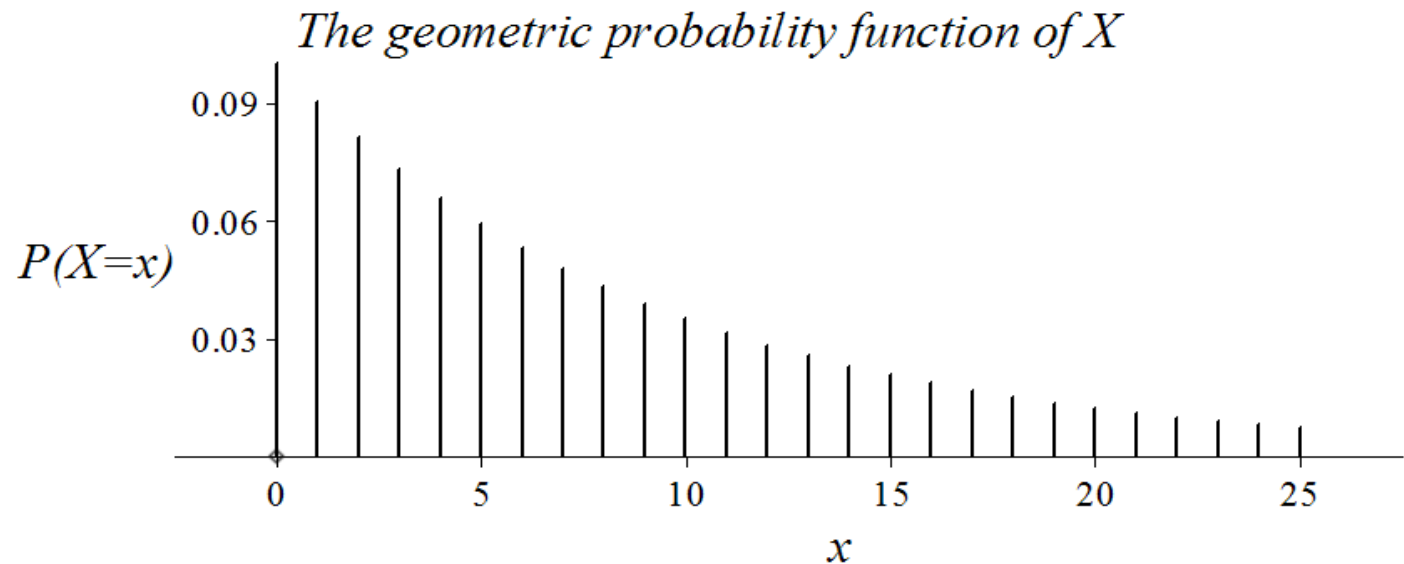
$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}, k = 0, 1, \dots, n$$

The numerical variables  $X$  are called **random variables** → chapter 4 .

# Example geometric distribution

Suppose that 10% of the passing cars are Mercedes.  
 $X$  = “the number van the first passing Mercedes”

$$P(X = k) = 0.9^{k-1} \cdot 0.1, \text{ with } k = 1, 2, 3, \dots$$



We can reason that on the expected value is

$$E(X) = \frac{1}{0.1} = 10 \text{ and } P(X > 10) = 0.9^{10} \approx 34.9\%$$

# The binomial distribution – 2 examples

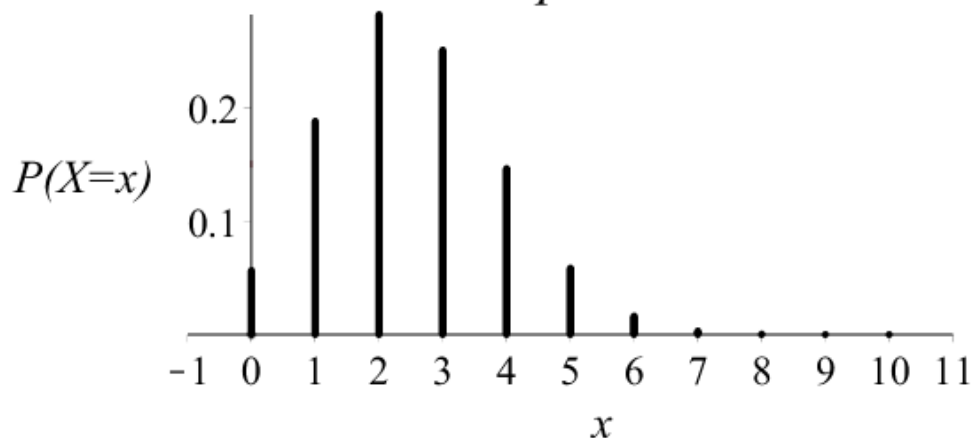
Applicable if we have  $n$  (ind.) Bernoulli trials with success rate  $p$

Example:  $X$  = the number of correct random answers to 10 MC-items:

here  $n = 10$  and  $p = \frac{1}{4}$

$$P(X = x) = \binom{10}{x} \left(\frac{1}{4}\right)^x \left(\frac{3}{4}\right)^{10-x}, \quad x = 0, \dots, 10$$

$n=10$  and  $p=0.25$



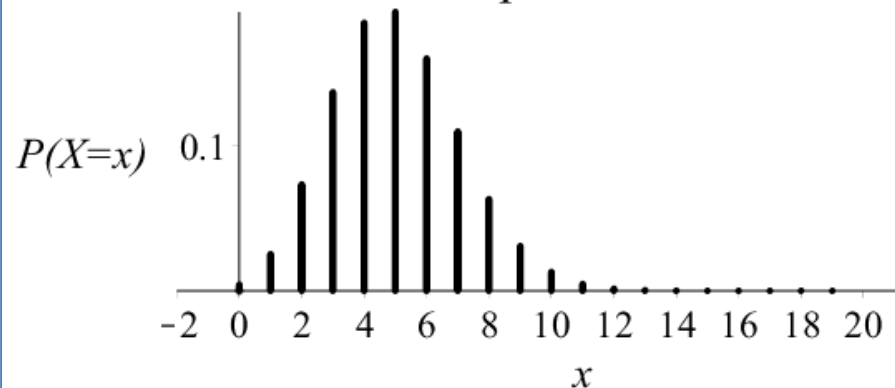
Example:

$X$  = The number of sixes in 30 rolls of a dice.

So  $n = 30$  and  $p = \frac{1}{6}$

$$P(X = x) = \binom{30}{x} \left(\frac{1}{6}\right)^x \left(\frac{5}{6}\right)^{30-x}, \quad x = 0, \dots, 30$$

$n = 30$  and  $p = 0.167$



What is the expected number of correct answers and resp. sixes?

# Summary Chapter 3

- **Conditional Probability:**  $P(A|B) = \frac{P(A \cap B)}{P(B)}$
- **General product rule:**  $P(A \cap B) = P(A|B)P(B)$
- **Product rule in case of independence:**  
$$P(A \cap B) = P(A)P(B)$$

- **Law of Total Probability:**

$$P(A) = P(A|S_1)P(S_1) + \cdots + P(A|S_n)P(S_n)$$

- **Bayes` rule:**

$$P(S_1|A) = \frac{P(S_1 \cap A)}{P(A)} = \frac{P(A|S_1)P(S_1)}{P(A|S_1)P(S_1) + \cdots + P(A|S_k)P(S_k)}$$