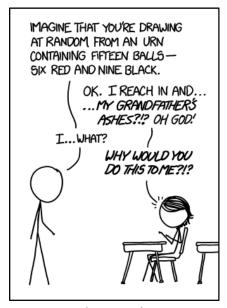


# **Applied Statistics**

Conditional Probability & Independence

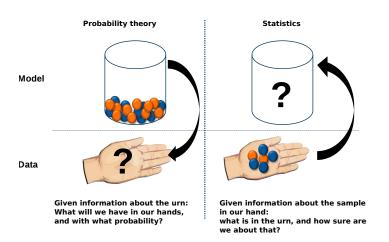
Release FS24

# Recap



(xkcd.com)

#### Recap



# Birthday problem - Problem 2.13

Probability that at least 2 out of 23 people have the same birthday:

$$1 - \frac{365 \cdot 364 \dots 343}{365^{23}} \approx 0.51$$

What about you? We have exactly 23 students in this module.



# Learning objectives

- Know the mathematical concept of independence and conditional probability
- Draw and read probability trees
- Calculate conditional probabilities using Bayes' theorem
- ▶ Apply Bayes' theorem to implement a simple spam filter

# Independence

#### Definition (Independence)

Two events A and B are called **independent** if  $P(A \cap B) = P(A) \cdot P(B)$ .

Careful: Independent events and disjoint events are not the same!

- ▶ A, B disjoint:  $P(A \cap B) = 0$ , i.e. if either event occurs, then the other cannot occur
- ▶ A, B independent:  $P(A \cap B) = P(A) \cdot P(B)$ , i.e. if either event occurs, this gives no information about the other

#### Examples

- Sex / Handedness: 10% of all people are left-handed; 12% of all men are left-handed. What do you conclude?
  Sex and handedness are not independent.
- Bonnie & Clyde: Bonnie attends class with probability 0.6; Clyde with probability 0.5. Both of them attend class with probability 0.4. What do you conclude? Attendance of Bonnie and Clyde are **not** independent. If they were, they would both attend class with probability 0.6 ⋅ 0.5 = 0.3.

# Conditional probabilities

#### Definition (Conditional probability)

Let A and B be events (with P(B) > 0). The **conditional probability of** A **given** B is defined as

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

**Consequence**: If *A* and *B* are independent, then

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

# Example

Bonnie & Clyde: Bonnie attends class with probability 0.6; Clyde with probability 0.5. Both of them attend class with probability 0.4.

$$\begin{split} \text{P(Bonnie attends}|\text{Clyde attends}) &= \frac{\text{P(Bonnie attends} \cap \text{Clyde attends})}{\text{Clyde attends}} \\ &= \frac{0.4}{0.5} \\ &= 0.8 \end{split}$$

# Example

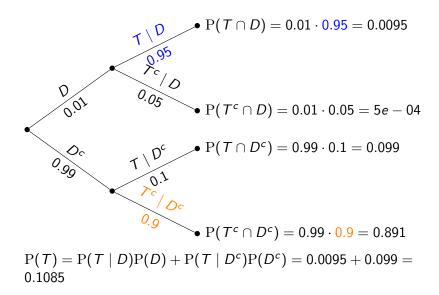
Bonnie & Clyde: Bonnie attends class with probability 0.6; Clyde with probability 0.5. Both of them attend class with probability 0.3.

$$\begin{split} P(\mathsf{Bonnie\ attends}|\mathsf{Clyde\ attends}) &= \frac{P(\mathsf{Bonnie\ attends} \cap \mathsf{Clyde\ attends})}{\mathsf{Clyde\ attends}} \\ &= \frac{0.3}{0.5} \\ &= 0.6 \\ &= P(\mathsf{Bonnie\ attends}) \end{split}$$

#### Example: PCR test

- Consider PCR-based test to assess whether someone is suffering from COVID-19. This medical test is quite accurate: it detects COVID-19 with 95% probability (sensitivity of the test), and indicates the absence of COVID-19 with 90% probability (specificity of the test).
- Notation: event D: person has COVID-19; event T: test is positive (i.e., indicates infection)
- Assume COVID-19 has incidence of 1%: P(D) = 0.01.
- What is the probability that a random person gets a positive test result?

# Easy visualization with a probability tree

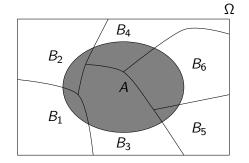


#### Law of total probability

#### Proposition (Law of total probability)

Assume  $B_1, B_2, ..., B_k$  are disjoint events with  $B_1 \cup B_2 \cup ... \cup B_k = \Omega$ . Then the probability of any event A is

$$P(A) = \sum_{i=1}^{k} P(A \cap B_i) = \sum_{i=1}^{k} P(A|B_i)P(B_i).$$



# Bayes' theorem

#### Theorem (Bayes' theorem)

Let A and B be events with P(A) > 0 and P(B) > 0. Then we have:

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

In the setting of the law of total probability, we have

$$P(B_i|A) = \frac{P(A \cap B_i)}{P(A)} = \frac{P(A|B_i)P(B_i)}{\sum_{j=1}^k P(A|B_j)P(B_j)}.$$

#### Example: PCR test

The PCR test from the previous example gives a positive result (i.e. indicates an infection).

What is the probability that you actually suffer from COVID-19?

$$P(D|T) = \frac{P(T|D) \cdot P(D)}{P(T)} = \frac{0.95 \cdot 0.01}{0.1085} = 0.0876$$