# CIS 678 - Machine Learning

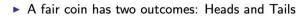
**Basics of Probability** 

**Tossing a fair coin:** Mutually exclusive events





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Tail

**Tossing a fair coin:** Mutually exclusive events





A fair coin has two outcomes: Heads and Tails





► Each outcome has equal probability:

$$P(H) = \frac{1}{2}, \quad P(T) = \frac{1}{2}$$

**Tossing a fair coin:** Mutually exclusive events





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► Sums to one (Mutually exclusive):

$$P(H) + P(T) = 1$$

**Tossing a fair coin:** Mutually exclusive events





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Range:

$$0 \le P(H) \le 1$$

**Tossing a fair coin:** Mutually exclusive events

#### Addition Rule (Independent Events)

$$P(A \cup B) = P(A) + P(B)$$

Example:

► 
$$P(H \cup T) = P(H) + P(T) = \frac{1}{2} + \frac{1}{2} = 1$$

**Tossing a fair coin:** Mutually exclusive events

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Example:

$$P(H \cup T) = P(H) + P(T) = \frac{1}{2} + \frac{1}{2} = 1$$

#### Multiplication Rule (Independent Events)

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

Example: Tossing the coin twice

►  $P(H \text{ on 1st and } H \text{ on 2nd}) = \frac{1}{2} \times \frac{1}{2} = \frac{1}{4}$ 

Tossing a fair dice: Mutually not exclusive event



Tossing a fair dice: Mutually not exclusive event



What is the probability of getting a 2 or getting an even number when rolling a fair six-sided die?

► Sample space:  $S = \{1, 2, 3, 4, 5, 6\}$ 

Tossing a fair dice: Mutually not exclusive event



- ► Sample space:  $S = \{1, 2, 3, 4, 5, 6\}$
- ▶ Let A be the event "Getting a 2":  $A = \{2\}$
- ▶ Let B be the event "Getting an even number":  $B = \{2, 4, 6\}$

Tossing a fair dice: Mutually not exclusive event



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- ▶ Let A be the event "Getting a 2":  $A = \{2\}$
- ▶ Let B be the event "Getting an even number":  $B = \{2,4,6\}$
- ▶ Find:  $P(A \cup B)$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Tossing a fair dice: Mutually not exclusive event



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$$P(A) = \frac{1}{6}, \quad P(B) = \frac{3}{6}, \quad P(A \cap B) = \frac{1}{6}$$

Tossing a fair dice: Mutually not exclusive event



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$$P(A) = \frac{1}{6}, \quad P(B) = \frac{3}{6}, \quad P(A \cap B) = \frac{1}{6}$$

$$P(A \cup B) = \frac{1}{6} + \frac{3}{6} - \frac{1}{6} = \frac{1}{2}$$

Tossing a fair coin: Mutually not exclusive event

Tossing a fair coin: Mutually not exclusive event

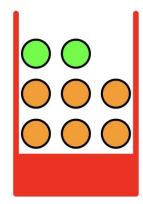
- Sample space:  $S = \{1, 2, 3, 4, 5, 6\}$
- ▶ Event A: Getting a 2  $A = \{2\}$
- Event B: Getting an even number  $B = \{2, 4, 6\}$
- ▶ We want  $P(A \cap B)$
- ▶  $A \cap B = \{2\}$
- ► This means that getting a 2 automatically satisfies both conditions

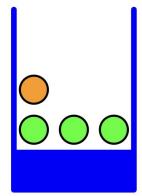
$$P(A \cap B) = \frac{\text{Number of favorable outcomes}}{\text{Total outcomes}} = \frac{1}{6}$$

- There are some orange and green balls in a red and blue box
- Someone (blinded) picked up a ball and it found to be with color orange
- What is is probability that the ball came from the red box?

Let's try to solve this question...







You have to apply Bayes Rule

#### 6.2. Bayes Rule (Theorem)

Combining equations

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \text{ and } P(B|A) = \frac{P(B \cap A)}{P(A)}$$
(B.17)

leads to the Bayes rule (Bayes theorem)

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$
 (B.18)

The Bayes rule is useful to swap events in conditional probability evaluation. The conditional probability P(A|B) can be expressed by the conditional probability P(B|A), P(A) and P(B).

The Bayes rule can be extended to a collection of events  $A_1, \dots, A_n$  conditioned on the event B

$$P(A_i|B) = \frac{P(B|A_i)P(A_i)}{P(B)} = \frac{P(B|A_i)P(A_i)}{\sum_{i=1}^{n} [P(A_i)P(B|A_i)]}$$
(B.19)

where

$$P(B) = \sum_{i=1}^{n} [P(B|A_i)P(A_i)]$$
(B.20)

**Solution:** Bishop: Pattern Recognition and Machine Learning.

(Section 1.2: Probability Theory); e-copy.