

# PROBABILITY AND STATISTICS: A RAPID RECAP

CS1951A INTRO TO DATA SCIENCE

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# OUTLINE

## ① **Probability Spaces & Probability Functions**

- ① Example: Rolling a Die
- ② Conditional Probability
- ③ Independent Events

## ② **Bayesian Statistics**

# STATISTICS $\neq$ PROBABILITY THEORY

**Probability theory:** mathematical theory that describes uncertainty.

**Statistics:** set of techniques for extracting useful information from data.

# PROBABILITY SPACE

A probability space has three components:

- (1) A sample space  $\Omega$  which is the set of all possible outcomes of the random process modeled by the probability space;
- (2) A family of sets  $F$  representing the allowable events, where each set in  $F$  is a subset of the sample space  $\Omega$ ;
- (3) A probability function  $\Pr : F \rightarrow R$  satisfying the definition below:

An element of  $\Omega$  is a simple event. In a discrete probability space, we use  $F = 2^\Omega$

# PROBABILITY FUNCTION

A **probability function** is any  $\Pr : F \rightarrow R$  that satisfies the following conditions:

- ① For any event  $E$ ,  $0 \leq \Pr(E) \leq 1$ ;
- ②  $\Pr(\Omega) = 1$ ;
- ③ For any finite or countably infinite sequence of pairwise mutually disjoint events  $E_1, E_2, E_3 \dots$

$$\Pr \left[ \bigcup_{i \in I} E_i \right] = \sum_{i \in I} \Pr(E_i)$$

The probability of an event is the sum of the probabilities of its simple events.

# EXAMPLE: TOSSING A (FAIR) COIN

$$W = \{H, T\}$$

$$F = 2^W = 2^2 = 4 \text{ Events}$$

$$F = \{\{\ }, \{H\}, \{T\}, \{H, T\}\}$$

$$\Pr(\{\ }) = 0$$

$$\Pr(\{H\}) = 0.5$$

$$\Pr(\{T\}) = 0.5$$

$$\Pr(\{H, T\}) = 1$$



# EXAMPLE: ROLLING A DIE

$$W = \{1, 2, 3, 4, 5, 6\}$$

$$F = 2^W = 2^6 \text{ Events}$$



$$\Pr(\{\ \}) = 0$$

$$\Pr(\{1\}) = \Pr(\{2\}) = \Pr(\{3\}) = \Pr(\{4\}) = \Pr(\{5\}) = \Pr(\{6\}) = \frac{1}{6}$$

$$\Pr(\{1, 2\}) = \Pr(\{1, 3\}) = \Pr(\{1, 4\}) = \Pr(\{1, 5\}) = \Pr(\{1, 6\}) = \frac{2}{6}$$

...

# COMPUTING CONDITIONAL PROBABILITY

The **conditional probability** that event  $E_1$  occurs given that event  $E_2$  occurs is:

$$\Pr(E_1 | E_2) = \frac{\Pr(E_1 \cap E_2)}{\Pr(E_2)}$$

The conditional probability is only well-defined if  $\Pr(E_2) > 0$

By conditioning on  $E_2$  we restrict the sample space to set  $E_2$ .

Thus we are interested in  $\Pr(E_1 \cap E_2)$  normalized by  $\Pr(E_2)$ .



# EXAMPLE: CONDITIONAL PROBABILITY

We have two coins: **A** is a fair coin, **B** has probability  $2/3$  to come up as HEAD. We chose a coin at random and got HEAD.

What is the probability that we chose coin **A**?

- ①  $E_1$  = the event “chose coin **A**”
- ②  $E_2$  = the event “outcome is HEAD”

Conditional probability that we chose coin **A** **given** that the outcome is HEAD is denoted:  $Pr(E_1 | E_2)$

# EXAMPLE: CONDITIONAL PROBABILITY

Define a **sample space** of **ordered** pairs: *(coin, outcome)*

The same space has four points:

- ①  $\{(A,h), (A,t), (B,h), (B,t)\}$
- ②  $Pr((A,h)) = Pr((A,t)) = 1/4$
- ③  $Pr((B,h)) = (1/2)(2/3) = 1/3$
- ④  $Pr((B,t)) = 1/2 * 1/3 = 1/6$

Define 2 **events**:

- ①  $E_1 = \text{"chose coin A"}$
- ②  $E_2 = \text{"outcome is HEAD"}$

$$Pr(E_1 | E_2) = \frac{Pr(E_1 \cap E_2)}{Pr(E_2)} = \frac{\frac{1}{4}}{\frac{1}{4} + \frac{1}{3}} = \frac{3}{7}$$

# INDEPENDENT EVENTS

Two events  $E$  and  $F$  are **independent** if and only if:

$$\Pr(E_1 \cap E_2) = \Pr(E) \Pr(F)$$

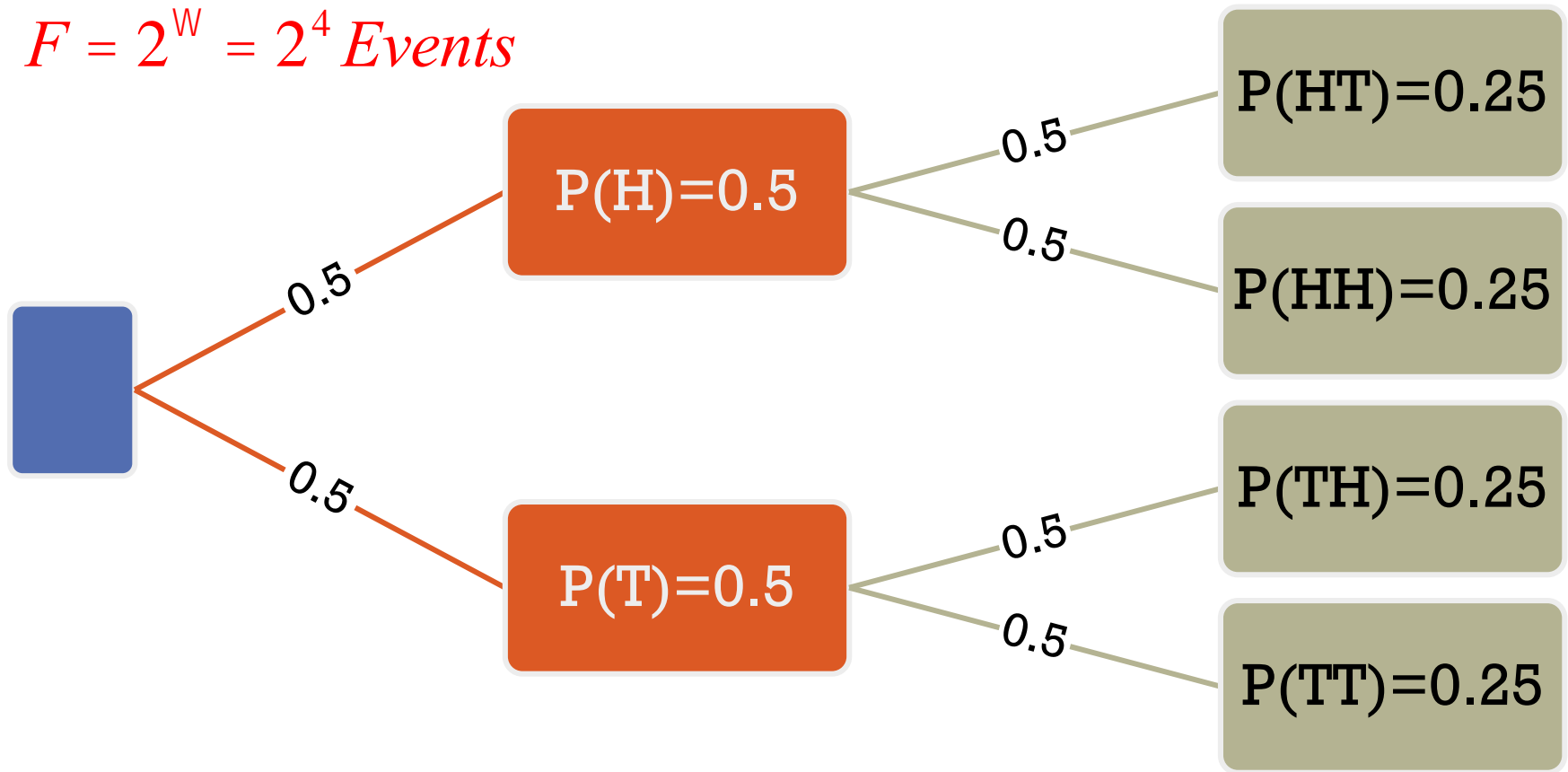
More generally, events  $E_1, E_2, \dots, E_k$  are **mutually independent** if and only if for any subset  $I \subseteq [1, k]$

$$\Pr\left(\bigcap_{i \in I} E_i\right) = \prod_{i \in I} \Pr(E_i)$$

# EXAMPLE: TOSSING A (FAIR) COIN TWICE

$$W = \{HH, HT, TH, TT\}$$

$$F = 2^W = 2^4 \text{ Events}$$



$$P(\{HH, HT\}) = P(\{HT, TT\}) = 0.5$$

# IN CLASS EXERCISES

A fair coin was tossed 10 times and always ended up on **HEAD**. What is the likelihood that it will end up **TAIL** next?

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The prior observations don't affect the likelihood.  
→  $1/2$

# IN CLASS EXERCISES

Stan has two kids. One of his kids is a boy. What is the likelihood that the other one is also a boy?

$E_1$  = Two Boys (BB)

$E_2$  = At least one kid is a boy (B)

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Stan has two kids. One of his kids is a boy. What is the likelihood that the other one is also a boy?

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$$P(BB) = \frac{1}{2} * \frac{1}{2} = \frac{1}{4}$$

$$P(B) = 1 - P(GG) = 1 - \frac{1}{2} * \frac{1}{2} = \frac{3}{4}$$

$$P(BB|B) = \frac{P(BB \cap B)}{P(B)} = \frac{\cancel{1}/4}{3/\cancel{4}} = \frac{1}{3}$$



# BAYESIAN STATISTICS

# LAW OF TOTAL PROBABILITY

## Theorem: Law of Total Probability

Let  $E_1, E_2, \dots, E_n$  be **mutually disjoint** events in a sample space  $\Omega$ , and  $\bigcup_{i=1}^n E_i = \Omega$

Then:

$$\Pr(B) = \sum_{i=1}^n \Pr(B \cap E_i) = \sum_{i=1}^n \Pr(B | E_i) \Pr(E_i)$$

# BAYES' LAW

## Theorem: Bayes' Law

Let  $E_1, E_2, \dots, E_n$  be **mutually disjoint** events in a sample space  $\Omega$ , and  $\bigcup_{i=1}^n E_i = \Omega$

Then:

$$\Pr(E_j | B) = \frac{\Pr(B \cap E_j) \Pr(E_j)}{\Pr(B) \sum_{i=1}^n \Pr(B \cap E_i) \Pr(E_i)}$$

$$\text{Conditional Probability: } \Pr(A|B) = \frac{\Pr(A \cap B)}{\Pr(B)}$$

$$\text{Law of Total Probability: } \Pr(B) = \sum_{j=1}^n \Pr(B|E_j) \Pr(E_j)$$

# BAYES' LAW

## Likelihood

Probability of  
collecting this data  
when our hypothesis  
is true

## Prior

The probability of the  
hypothesis being true  
before collecting data

$$P(H|D) = \frac{P(D|H) P(H)}{P(D)}$$

## Posterior

The probability of our  
hypothesis being true  
given the data collected

## Marginal

What is the probability of  
collecting this data under  
all possible hypotheses?

# APPLICATION: FINDING A BIASED COIN

- We are given three coins. 2 coins are fair, and the 3<sup>rd</sup> is biased (landing heads with probability  $\frac{2}{3}$ ). We need to identify the the biased coin.
- We flip each of the coins. The first and second come up heads, and the third comes up tails.
- What is the probability that the first coin was the biased one?

# APPLICATION: FINDING A BIASED COIN

Let  $E_i$  be the event that the  $i$ th coin flip is the biased one and let B be the event that the three coin flips came up HEADS, HEADS, and TAILS.

Before we flip the coins we have  $\Pr(E_i) = 1/3$  for  $i=1,\dots,3$ , thus

$$\Pr(B | E_1) = \Pr(B | E_2) = \frac{2}{3} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{6}$$

and

$$\Pr(B | E_3) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{3} = \frac{1}{12}$$

Applying Bayes' Law we have...

$$\Pr(E_1 | B) = \frac{\Pr(B | E_1)\Pr(E_1)}{\sum_{i=1}^3 \Pr(B | E_i)\Pr(E_i)} = \frac{2}{5}$$

The outcome of the 3 coin flips increases the probability that the first coin is the biased one from  $1/3$  to  $2/5$ .

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$$P(BB|B) = \frac{P(BB \cap B)}{P(B)} = \frac{\cancel{1}/4}{\cancel{3}/4} = \frac{1}{3}$$

$$P(BB|B) = \frac{P(B|BB) * P(BB)}{P(B)} = \frac{1 * \cancel{1}/4}{\cancel{3}/4} = \frac{1}{3}$$

# IN CLASS EXERCISES: DRUG TEST

- 0.4% of the Rhode Island population use Marijuana\*
- Drug Test: The test will produce 99% true positive results for drug users and 99% true negative results for non-drug users.
- If a randomly selected individual is tested positive, what is the probability he or she is a user?

$$\begin{aligned} P(User|+) &= \frac{P(+|User)P(User)}{P(+)} \\ &= \frac{P(+|User)P(User)}{P(+|User)P(User) + P(+|!User)P(!User)} \\ &= \frac{0.99 \cdot 0.004}{0.99 \cdot 0.004 + 0.01 \cdot 0.996} \\ &= 28.4\% \end{aligned}$$



# SPAM FILTERING WITH NAÏVE BAYES

$$P(\text{spam}|\text{words}) = \frac{P(\text{spam}) P(\text{words}|\text{spam})}{P(\text{words})}$$

$$P(\text{spam}|\text{viagra}, \text{rich}, \dots, \text{friend}) = \frac{P(\text{spam}) P(\text{viagra}, \text{rich}, \dots, \text{friend}|\text{spam})}{P(\text{viagra}, \text{rich}, \dots, \text{friend})}$$

$$P(\text{spam}|\text{words}) \gg \frac{P(\text{spam}) P(\text{viagra}|\text{spam}) P(\text{rich}|\text{spam}) \square P(\text{friend}|\text{spam})}{P(\text{viagra}, \text{rich}, \dots, \text{friend})}$$

# WARM UP QUESTION

- Assume a statistical test has a chance of 1% ( $P=0.01$ ) to be wrong
- How many test can you run before the likelihood of being wrong at least once is 50% or more?

$$P(T_1) = 1 - P(F_1) = 0.99$$

$$P(T_1 T_2) = 0.99 \cdot 0.99 \gg 0.98$$

$$P(F_1 F_2, F_1 T_2, T_1 F_2) = 1 - P(T_1 T_2) \gg 0.02$$

$$P(\text{Being Wrong At Least Once}) =$$

$$1 - P(T_1 T_2 \square T_{n-1} T_n) = 1 - P(T)^n$$

$$n = \log(0.5) / \log(0.99) \gg 68.96 \gg 69$$