

# **STA 130A: Mathematical Statistics: Brief Course**

## **Lecture 4: Independence and Counting Principle**

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# Announcements

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## **Homework 1 is due tomorrow (Thu, Oct 1, 11:59 PM)**

- Please submit on time and follow the submission instructions
- Instructor office hours: Wed, 2–3 PM (MSB 1143)
- TA office hours: Thu, 9–11 AM (MSB 1117)
- Feel free to post questions on Piazza (the earlier, the better)

# Agenda

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## Last time:

- Conditional probability
- Bayes' rule

## Today:

- Independence
- Modeling with independent components and repeated trials
- Counting principle and simple applications

# Independence

Sometimes knowing that  $B$  occurred does not change the probability of  $A$

- When  $P(B) > 0$ , this means  $P(A | B) = P(A)$

## Definition (Independence)

The event  $A$  is **independent** of  $B$  if

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

- Independence is symmetric:  $A$  is independent of  $B \implies B$  is independent of  $A$ 
  - We can unambiguously say “ $A$  and  $B$  are independent events”
- Events from distinct, non-interacting experiments are often modeled as independent
- Independence is **not the same as disjointness**
  - If  $A \cap B = \emptyset$  and  $P(A), P(B) > 0$ , then  $A$  and  $B$  cannot be independent (e.g.,  $A$  and  $A^c$ )
- $A$  and  $B$  are independent  $\implies A^c$  and  $B$  are independent
  - Because  $P(A^c \cap B) = P(B) - P(A \cap B) = P(B) - P(A)P(B) = P(A^c)P(B)$

## Checking independence: Two rolls of a fair die (1/3)

### Example (2 successive die rolls)

Model two independent rolls of a fair die; all  $6 \times 6 = 36$  outcomes are equally likely

**Question:** Are the following two events independent?

$$A_1 = \{\text{1st roll results in 6}\}, \quad B_1 = \{\text{2nd roll results in 6}\}$$

**Answer:** Yes, because

$$P(A_1 \cap B_1) = P(\text{the outcome of the two rolls is } (6, 6)) = \frac{1}{36}$$

$$P(A_1) = \frac{P(\{(6, 1), (6, 2), \dots, (6, 6)\})}{P(\{(1, 1), (1, 2), \dots, (6, 6)\})} = \frac{6}{36} = \frac{1}{6}$$

$$P(B_1) = \frac{P(\{(1, 6), (2, 6), \dots, (6, 6)\})}{P(\{(1, 1), (1, 2), \dots, (6, 6)\})} = \frac{6}{36} = \frac{1}{6}$$

Observe  $P(A_1 \cap B_1) = P(A_1) P(B_1)$ , which verifies the independence of  $A_1$  and  $B_1$ )

## Checking independence: Two rolls of a fair die (2/3)

### Example (2 successive die rolls)

Model two independent rolls of a fair die; all  $6 \times 6 = 36$  outcomes are equally likely

**Question:** Are the following two events independent?

$$A_2 = \{\text{1st roll is 6}\}, \quad B_2 = \{\text{the sum of the two rolls is 7}\}$$

**Answer:** Yes, because

$$\begin{aligned} P(A_2 \cap B_2) &= P(\{(6, 1)\}) = \frac{1}{36} \\ P(A_2) &= \frac{P(\{(6, 1), (6, 2), \dots, (6, 6)\})}{P(\{(1, 1), (1, 2), \dots, (6, 6)\})} = \frac{6}{36} = \frac{1}{6} \\ P(B_2) &= \frac{P(\{(6, 1), (5, 2), \dots, (1, 6)\})}{P(\{(1, 1), (1, 2), \dots, (6, 6)\})} = \frac{6}{36} = \frac{1}{6} \end{aligned}$$

Observe  $P(A_2 \cap B_2) = P(A_2) P(B_2)$ , which verifies the independence of  $A_2$  and  $B_2$

**Follow-up:** What about  $B'_2 = \{\text{the sum is 5}\}$ ? No, because  $P(A_2 \cap B'_2) = 0$  but  $P(B'_2) > 0$

## Checking independence: Two rolls of a fair die (3/3)

### Example (2 successive die rolls)

Model two independent rolls of a fair die; all  $6 \times 6 = 36$  outcomes are equally likely

**Question:** Are the following two events independent?

$$A_3 = \{\text{maximum of the two rolls is } 3\}, \quad B_3 = \{\text{minimum of the two rolls is } 3\}$$

**Answer:** No, because

$$P(A_3 \cap B_3) = P(\{(3, 3)\}) = \frac{1}{36}$$

$$P(A_3) = \frac{P(\{(1, 3), (2, 3), (3, 3), (3, 2), (3, 1)\})}{P(\{(1, 1), (1, 2), \dots, (6, 6)\})} = \frac{5}{36}$$

$$P(B_3) = \frac{P(\{(3, 6), (3, 5), (3, 4), (3, 3), (4, 3), (5, 3), (6, 3)\})}{P(\{(1, 1), (1, 2), \dots, (6, 6)\})} = \frac{7}{36}$$

Observe  $P(A_3 \cap B_3) = \frac{1}{36} \neq \frac{35}{36^2} = P(A_3)P(B_3)$ , thus  $A_3$  and  $B_3$  are not independent

# Conditional independence

Recall: for any event  $C$  with  $P(C) > 0$ , the map  $A \mapsto P(A \mid C)$  is itself a probability law

- So we can ask whether  $A$  and  $B$  are independent under this conditional law

## Definition (Conditional independence)

The events  $A$  and  $B$  are **conditionally independent** given an event  $C$  if

$$P(A \cap B \mid C) = P(A \mid C) P(B \mid C)$$

If  $P(B \cap C) > 0$ , this is equivalent to  $P(A \mid B \cap C) = P(A \mid C)$  because

$$\begin{aligned} P(A \cap B \mid C) &= \frac{P(A \cap B \cap C)}{P(C)} && \text{by definition} \\ &= \frac{P(C) P(B \mid C) P(A \mid B \cap C)}{P(C)} && \text{by multiplicative rule} \\ &= P(B \mid C) P(A \mid B \cap C) \end{aligned}$$

- Once  $C$  is known to have occurred, knowing  $B$  gives no additional information about  $A$
- *Independence can change when we condition on additional information*

# Conditioning can change independence

## Example (2 fair coin tosses)

Consider two independent fair coin tosses, and let

$$A = \{\text{1st toss is a head}\}$$

$$B = \{\text{2nd toss is a head}\}$$

- These events are independent. Now condition on

$$C = \{\text{the two tosses have different results}\}$$

- The events  $A$  and  $B$  are not conditionally independent given  $C$  because

$$P(A \mid C) = \frac{1}{2}, \quad P(B \mid C) = \frac{1}{2}, \quad P(A \cap B \mid C) = 0$$

**Why?** Given  $C$ , the only possible outcomes are  $HT$  and  $TH$ , so if one toss is Head, the other must be Tail

The independence of  $A$  and  $B$  does not imply conditional independence

## (Optional) Conditional independence $\not\Rightarrow$ Independence

### Example (2 different biased coin tosses)

Suppose there are two coins, fair and biased (say, landing on head with prob 0.9), and we choose one of the two at random with probability  $1/2$  each, and toss the coin twice:

$A_i = \{\text{the } i\text{-th toss is a head}\}$

$B = \{\text{the coin is biased}\}$

- The events  $A_1$  and  $A_2$  are independent, given the choice of a coin (=conditioned on  $B$ )
- However, without conditioning on  $B$ ,  $A_1$  and  $A_2$  are not independent:

$$P(A_i) = P(B) P(A_i | B) + P(B^c) P(A_i | B^c) = \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{9}{10} = \frac{7}{10}, \quad i \in \{1, 2\},$$
$$P(A_1 \cap A_2) = P(B) P(A_1 \cap A_2 | B) + P(B^c) P(A_1 \cap A_2 | B^c) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{9}{10} \cdot \frac{9}{10} = \frac{53}{100}$$

**Why?** The hidden variable “which coin was chosen” creates dependence between the two tosses

The conditional independence of  $A$  and  $B$  does not imply unconditional independence

## Pop-up quiz

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Suppose two independent fair coin tosses. Let

$$A = \{\text{1st toss is H}\}, \quad B = \{\text{2nd toss is H}\}, \quad C = \{\text{the two tosses are different}\}.$$

Which statement is correct?

- a)  $A$  and  $B$  are not independent, but are conditionally independent given  $C$ .
- b)  $A$  and  $B$  are independent, and are conditionally independent given  $C$ .
- c)  $A$  and  $B$  are independent, but are not conditionally independent given  $C$ .
- d)  $A$  and  $B$  are neither independent nor conditionally independent given  $C$ .

**Answer: C.**

$$P(A) = P(B) = \frac{1}{2}, \quad P(A \cap B) = \frac{1}{4},$$

so  $A$  and  $B$  are independent. But given  $C$ , the only outcomes are  $HT$  and  $TH$ , so

$$P(A \mid C) = P(B \mid C) = \frac{1}{2}, \quad P(A \cap B \mid C) = 0.$$

**Follow-up:** Would the answer change if we condition on  $C^c$  instead of  $C$ ?

# Independence of several events

We can extend the definition of the events for more than two events

## Definition (Independence (mutual independence))

The events  $A_1, A_2, \dots, A_n$  are **independent** if

$$P\left(\bigcap_{i \in S} A_i\right) = \prod_{i \in S} P(A_i), \quad \text{for every subset } S \subseteq \{1, 2, \dots, n\}$$

For the case of three events,  $A_1, A_2, A_3$  are independent if and only if

$$\begin{aligned} P(A_1 \cap A_2) &= P(A_1) P(A_2), \\ P(A_1 \cap A_3) &= P(A_1) P(A_3), \\ P(A_2 \cap A_3) &= P(A_2) P(A_3), \\ P(A_1 \cap A_2 \cap A_3) &= P(A_1) P(A_2) P(A_3) \end{aligned}$$

The first three (pairwise independence) do not imply the fourth, and vice versa

# Pairwise independence is not enough for mutual independence

## Example (2 fair coin tosses)

Consider two independent fair coin tosses, and let

$$A_1 = \{\text{1st toss is a head}\}$$

$$A_2 = \{\text{2nd toss is a head}\}$$

$$B = \{\text{the two tosses have different results}\}$$

- These three events are pairwise independent:
  - $A_1$  and  $A_2$  are independent by definition
  - $A_1$  and  $B$  are independent because

$$P(B \mid A_1) = \frac{P(A_1 \cap B)}{P(A_1)} = \frac{P(\{(H, T)\})}{P(\{(H, T), (H, H)\})} = \frac{1}{2} = P(B)$$

- However,

$$P(A_1 \cap A_2 \cap B) = 0 \neq \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = P(A_1) P(A_2) P(B)$$

and therefore, these three events are not independent

## (Optional) Three-way factorization $\nRightarrow$ mutual independence

### Example (2 fair die rolls)

Consider two independent rolls of a fair six-sided die, and let

$$A = \{1\text{st roll is } 1, 2, \text{ or } 3\}$$

$$B = \{1\text{st roll is } 3, 4, \text{ or } 5\}$$

$$C = \{\text{the sum of the two rolls is } 9\} = \{(3, 6), (4, 5), (5, 4), (6, 3)\}$$

- Observe that

$$P(A \cap B \cap C) = P(\{(3, 6)\}) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{4}{36} = P(A) P(B) P(C)$$

- Nevertheless, any pair of these three events are NOT independent:

$$P(A \cap B) = P(\{1\text{st roll is } 3\}) = \frac{1}{6} \neq P(A) P(B)$$

$$P(A \cap C) = P(\{(3, 6)\}) = \frac{1}{36} \neq P(A) P(C)$$

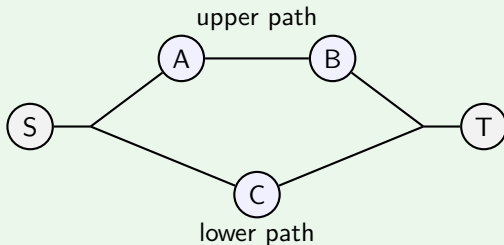
$$P(B \cap C) = P(\{(3, 6), (4, 5), (5, 4)\}) = \frac{1}{12} \neq P(B) P(C)$$

and thus, these three events are not independent

## Example: Using independence for modeling (1/2)

### Example (Network connectivity)

Consider a network connecting two nodes  $S$  and  $T$  through intermediate nodes  $A$ ,  $B$ ,  $C$  as follows:



Suppose the intermediate nodes  $A$ ,  $B$ ,  $C$  fail independently, with probabilities

$$p_a = 0.1, \quad p_b = 0.05, \quad p_c = 0.2.$$

**Question:** What is the probability that there is a path connecting  $S$  and  $T$  without failure?

## Example: Using independence for modeling (2/2)

### Example (Network connectivity)

**Answer:** Define two events

$$U = \{\text{the upper path works}\}, \quad L = \{\text{the lower path works}\}.$$

Then, since  $A, B$  are independent,

$$P(U) = (1 - p_a)(1 - p_b), \quad P(L) = 1 - p_c.$$

Since  $U$  depends only on  $A, B$  and  $L$  depends only on  $C$ , the events  $U$  and  $L$  are independent. Thus,

$$\begin{aligned} P(\text{there is a working path}) &= P(U \cup L) \\ &= 1 - P(U^c \cap L^c) \\ &= 1 - (1 - P(U))(1 - P(L)) \end{aligned}$$

With  $p_a = 0.1$ ,  $p_b = 0.05$ ,  $p_c = 0.2$ ,

$$P(U) = 0.9 \cdot 0.95 = 0.855, \quad P(L) = 0.8,$$

and consequently,

$$P(\text{there is a working path}) = 1 - (1 - 0.855)(1 - 0.8) = 0.971$$

# The counting principle

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For a finite sample space  $\Omega$  with equally likely outcomes,

$$P(A) = \frac{|A|}{|\Omega|}$$

**Question:** How do we compute the number of elements in  $A$  and  $\Omega$ ?

## The counting principle (product rule)

Suppose a procedure consists of  $r$  successive stages, and the following hold:

- (a) There are  $n_1$  possible results at the first stage
- (b) For any sequence of possible results until stage  $(i - 1)$ , there are  $n_i$  possible results at the  $i$ -th stage

Then the total number of possible results of this  $r$ -stage process is

$$n_1 n_2 \cdots n_r.$$

- **Divide-and-conquer:** Break a complicated counting problem into simple stages
- Multiply the number of choices across stages
- Example: 3 Bernoulli trials have  $2 \cdot 2 \cdot 2 = 2^3$  possible outcomes

# The counting principle: Examples

## Example (The number of subsets)

**Question:** How many subsets does a set  $A = \{a_1, a_2, \dots, a_n\}$  have?

**Answer:** We can consider a sequential process that examines one element at a time to decide whether to include it in the subset or not. Therefore, the number of subsets is

$$\underbrace{2 \times 2 \times \dots \times 2}_{n \text{ times}} = 2^n$$

## Example (The number of telephone numbers)

A local telephone number consists of a 7-digit sequence whose first digit cannot be 0 or 1

**Question:** How many distinct telephone numbers are there within the local area?

**Answer:** We have a total of 7 stages, and a choice of one out of 10 elements at each stage, except for the first where we only have 8 choices. Therefore, the answer is

$$8 \times \underbrace{10 \times 10 \times \dots \times 10}_{6 \text{ times}} = 8 \times 10^6$$

## Example: Independent trials

### Example (3 independent Bernoulli trials)

Consider a sequence of 3 independent Bernoulli trials (e.g., coin tosses):

$$P(\{\text{Head}\}) = p, \quad P(\{\text{Tail}\}) = 1 - p$$

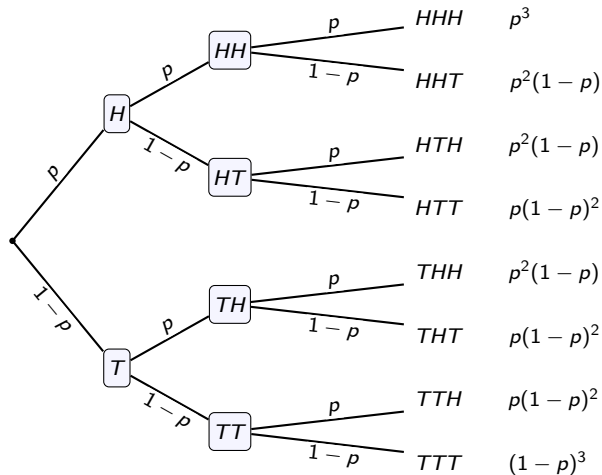
**Question:** Compute  $p(k) = P(k \text{ heads})$

**Answer:** We can visualize independent Bernoulli trials using a sequential description (next slide)

$$p(k) = \begin{cases} p^3, & k = 3, \\ 3p^2(1 - p), & k = 2, \\ 3p(1 - p)^2, & k = 1, \\ (1 - p)^3, & k = 0. \end{cases}$$

To compute  $p(k)$ , we need to count how many outcomes produce exactly  $k$  Heads  
→ This leads naturally to the basic principles of counting

# Illustration of 3 independent Bernoulli trials



**Question:** How can we count outcomes systematically, *without listing every outcome*?

# Wrap-up

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## Independence

- $A$  and  $B$  are independent if and only if  $P(A \cap B) = P(A)P(B)$
- Independence means that learning one event does not change the probability of the other
- Independence is not the same as disjointness
- Independence of several events require factorization of probabilities over every subcollection
- Conditioning can change independence

## Modeling with independence

- Combine independent components and trials using products, unions, and complements

## Counting

- Multiply numbers of choices across stages; probabilistic independence is not required
- For a finite equally likely model,  $P(A) = |A|/|\Omega|$

*Suggested reading:* [BT08, Ch. 1.5]

*Next lecture:* Permutations, combinations, and discrete random variables.

# References

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Dimitri Bertsekas and John N Tsitsiklis.

*Introduction to probability*, volume 1.

Athena Scientific, 2nd edition, 2008.