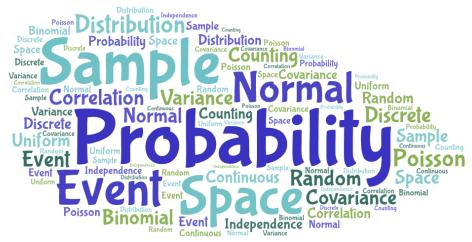
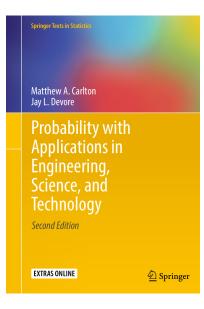
# Discrete Mathematics and Probability Week 7



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### **Topics**

- ► Counting: thinking algorithmically
- Events: what could happen in principle
- Experiments: how can events interact
- Probability: quantifying what could happen

# Counting

### Counting

Basic principles of combinatorics:

- ▶ if an experiment has n outcomes; and another experiment has m outcomes,
- ▶ then the two experiments jointly have  $n \cdot m$  outcomes.

#### **Permutations**

### **Definition**

Let  $H = \{h_1, h_2, \dots, h_n\}$  be a set of n different objects. The *permutations* of H are the different orders in which you can write all of its elements.

### Permutations with repetitions

#### **Definition**

Let  $H = \{h_1 \dots h_1, h_2 \dots h_2, \dots, h_r \dots h_r\}$  be a set of r different types of repeated objects:  $n_1$  many of  $h_1$ ,  $n_2$  of  $h_2$ , ...  $n_r$  of  $h_r$ . The permutations with repetitions of H are the different orders in which you can write all of its elements.

#### **k**-Permutations

#### **Definition**

Let  $H = \{h_1, h_2, \dots, h_n\}$  be a set of n different objects. The k-permutations of H are the different ways in which one can pick and write k of its elements of H in order.

### **k**-Permutations with repetitions

#### **Definition**

Let  $H = \{h_1, \dots, h_2, \dots, h_r, \dots\}$  be a set of r different types of repeated objects, each of infinite supply. The k-permutations with repetitions of H are the different orders in which one can write an ordered sequence of length k using the elements of H.

### **k**-Combinations

#### Definition

Let  $H = \{h_1, h_2, ..., h_n\}$  be a set of n different objects. The k-combinations of H are the different ways in which one can pick k of its elements without order.

### **Events**

#### **Events**

A mathematical model for experiments:

- ▶ Sample space: the set  $\Omega$  of all possible outcomes
- ▶ An *event* is a collection<sup>1</sup> of possible outcomes:  $E \subseteq \Omega$
- ▶ Union  $E \cup F$  and intersection  $E \cap F$  of events make sense

 $<sup>^{1}</sup>$ Sometimes  $\Omega$  is too large, and not all subsets are events. Ignore this now.

# Examples

#### Union and intersection

#### Union

Union  $E \cup F$  of events E and F means E and F. Infinite union  $\bigcup_i E_i$  of events  $E_i$  means at least one of the  $E_i$ 's.

#### Intersection

Intersection  $E \cap F$  of events E and F means E and F. Infinite intersection  $\bigcap_i E_i$  of events  $E_i$  means each of the  $E_i$ 's.

#### Definition

If  $E \cap F = \emptyset$ , we call events E and F mutually exclusive. If events  $E_1, E_2, \ldots$  satisfy  $E_i \cap E_j = \emptyset$  whenever  $i \neq j$ , we call them mutually exclusive. They cannot happen at the same time.

### Inclusion and implication

#### Remark

If the event E is a *subset* of the event F, written  $E \subseteq F$ , then the occurrence of E implies that of F.

# Complementarity

### Definition

The *complement* of an event E is  $E^c = \Omega - E$ .

This is the event that *E* does *not* occur.

# Experiments

# **Experiments**

#### How events can interact:

- Commutativity
- Distributivity
- Associativity
- ▶ De Morgan's Law

# Properties of events

Commutativity:  $E \cup F = F \cup E$  $E \cap F = F \cap E$ 

Associativity: 
$$E \cup (F \cup G) = (E \cup F) \cup G$$
  
 $E \cap (F \cap G) = (E \cap F) \cap G$ 

# Properties of events

**Distributivity**: 
$$(E \cup F) \cap G = (E \cap G) \cup (F \cap G)$$
  
 $(E \cap F) \cup G = (E \cup G) \cap (F \cup G)$ 

# De Morgan's law

▶ De Morgan's law:  $(E \cup F)^c = E^c \cap F^c$ 

# Probability

- ► Definition by axioms
- ► How to compute probabilities
- ► Inclusion-exclusion principle
- ► Equally likely outcomes

# Axioms of probability

#### Definition

The probability P on a sample space  $\Omega$  assigns numbers to events of  $\Omega$  in such a way that:

- 1. the probability of any event is non-negative:  $P(E) \ge 0$ ;
- 2. the probability of the sample space is one:  $P(\Omega) = 1$ ;
- 3. for countably many *mutually exclusive* events  $E_1, E_2, ...$ :

$$\mathbf{P}\big(\bigcup_i E_i\big) = \sum_i \mathbf{P}(E_i)$$

# How to compute probabilities

### Proposition

For any event,  $P(E^c) = 1 - P(E)$ .

### Corollary

We have  $P(\emptyset) = P(\Omega^c) = 1 - P(\Omega) = 1 - 1 = 0$ . For any event,  $P(E) = 1 - P(E^c) \le 1$ .

# How to compute probabilities

### Proposition

For any two events,  $P(E \cup F) = P(E) + P(F) - P(E \cap F)$ .

### Proposition (Boole's inequality)

For any events  $E_1, E_2, \ldots, E_n$ :

$$\mathbf{P}\left(\bigcup_{i=1}^n E_i\right) \leq \sum_{i=1}^n \mathbf{P}(E_i).$$

### Inclusion-exclusion

### Proposition

For any events:

$$P(E \cup F \cup G) = P(E) + P(F) + P(G)$$
$$- P(E \cap F) - P(E \cap G) - P(F \cap G)$$
$$+ P(E \cap F \cap G).$$

### Inclusion-exclusion

### Proposition

#### For any events:

$$P(E \cup F \cup G) = P(E) + P(F) + P(G)$$

$$- P(E \cap F) - P(E \cap G) - P(F \cap G)$$

$$+ P(E \cap F \cap G).$$

$$P(E_1 \cup E_2 \cup \dots \cup E_n) = \sum_{1 \le i \le n} P(E_i)$$

$$- \sum_{1 \le i_1 < i_2 \le n} P(E_{i_1} \cap E_{i_2})$$

$$+ \sum_{1 \le i_1 < i_2 < i_3 \le n} P(E_{i_1} \cap E_{i_2} \cap E_{i_3})$$

$$- \dots$$

$$+ (-1)^{n+1} P(E_1 \cap E_2 \cap \dots \cap E_n).$$

### Example

### Example

In a sports club,

36 members play tennis, 22 play tennis and squash,

28 play squash, 12 play tennis and badminton,

18 play badminton, 9 play squash and badminton,

4 play tennis, squash and badminton.

How many play at least one of these games?

# How to compute probabilities

### Proposition

If 
$$E \subseteq F$$
, then  $P(F - E) = P(F) - P(E)$ .

### Corollary

If  $E \subseteq F$ , then  $P(E) \leq P(F)$ .

# Equally likely outcomes

# The return of counting

Finite sample space,  $|\Omega| = N < \infty$ , has special important case where each experiment outcome has *equal probability*:

$$\mathbf{P}(\omega) = \frac{1}{N}$$
 for all  $\omega \in \Omega$ 

#### Definition

Outcomes  $\omega \in \Omega$  are also called *elementary events*.

### Example

### Example

Rolling two dice, what is the probability that the sum of the numbers shown is 7?

What's wrong with this solution? "The number 7 is one out of the possible values 2, 3, . . . , 12 for the sum, and the answer is  $\frac{1}{11}$ ."

# Summary

- ► Counting: permutations, combinations, repetitions
- ▶ Events: sample space, union, intersection, complement
- Experiments: distributivity, De Morgan's law
- ▶ Probability: axioms, how to compute, equally likely outcomes