# Chapter 2

**Combinatorial Methods** 

## Outline of Chapter 2

- 2.1 Introduction
- 2.2 Counting principle
- 2.3 Permutations
- 2.4 Combinations
- 2.5 Stirling formula

# Section 2.2

## **Counting Principles**

## Theorem 2.1 Counting Principles

• **Theorem 2.1** If the set E contains n elements and the set F contains m elements, there are nm ways in which we can choose, first, an element of E and then an element of F.

- Proof.
- Let  $E = \{a_1, a_2, \dots, a_n\}$  and  $F = \{b_1, b_2, \dots, b_m\}$ .
- The following rectangular array, which consists of *nm* elements, contains all possible ways that we can choose, first, an element of *E* and then an element of *F*.

$$(a_1,b_1), (a_1,b_2), \dots, (a_1,b_m)$$
  
 $(a_2,b_1), (a_2,b_2), \dots, (a_2,b_m)$   
 $\vdots$   
 $(a_n,b_1), (a_n,b_2), \dots, (a_n,b_m)$ 

## Theorem 2.2 Generalized Counting Principle

- Let  $E_1, E_2, \ldots, E_k$  be sets with  $n_1, n_2, \ldots, n_k$  elements, respectively.
- Then there are

$$n_1 \times n_2 \times n_3 \times \cdots \times n_k$$

ways in which we can, first, choose an element of  $E_1$ , then an element of  $E_2$ , then an element of  $E_3$ , ..., and finally an element of  $E_k$ .

## Example 2.1 and Remark 2.1

- How many outcomes are there if we throw 5 dice?
- Solution. Let  $E_i$ ,  $1 \le i \le 5$ , be the set of all possible outcomes of the i-th die.
- Then,  $E_i = \{1, 2, 3, 4, 5, 6\}.$
- The number of outcomes equals the number of ways we can choose one element each from sets  $E_1$ ,  $E_2$ ,  $E_3$ ,  $E_4$ , and finally  $E_5$ .
- From Theorem 2.2 there are 6<sup>5</sup>.
- We assume that all outcomes (sample points) in the sample space are *equally likely*.
- In section 3.5, we introduce the concept of *independence*.
- Assuming that all tosses are indepdent to one another, we can show that all outcomes are equally likely.

## Example 2.8 - Standard birthday problem

- What is the probability that at least two students in a class of size *n* have the same birthdays?
- Compute the probability for n = 23, 30, 50, 60.
- What is your intuition of this probability?

#### Solution

• Let P(n) be the probability that no students have the same birthdays.

$$P(n) = \frac{365 \times 364 \times 363 \times \dots \times [365 - (n-1)]}{365^n}$$

- The desired probability is 1 P(n).
- For n = 23, 30, 50, 60, the answers are 0.507, 0.706, 0.970 and 0.995.

#### Number of Subsets of a Set

• The set of all subsets of A is called the **power set** of A.

• Theorem 2.3 A set with n elements has  $2^n$  subsets

#### Proof of Theorem 2.3

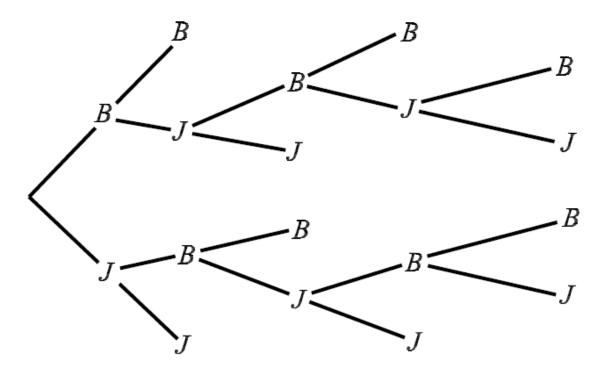
- Let  $A = \{a_1, a_2, a_3, ..., a_n\}$  be a set with n elements.
- Let B be a subset of A.
- There is a one-to-one correspondence between *B* and *A*.
- We associate a sequence  $b_1b_2b_3\cdots b_n$ , where

$$b_i = \begin{cases} 1 & \text{if } a_i \in B \\ 0 & \text{otherwise.} \end{cases}$$

- By the generalized counting principle, the number of sequences of 0's and 1's of length n is  $2 \times 2 \times \cdots \times 2 = 2^n$ .
- Thus the number of subsets of A is  $2^n$ .

## Tree Diagrams

• Tree diagrams are useful pictorial representations that break down a complex counting problem into smaller, more tractable ones.



## Example 2.11

- Bill and John keep playing chess until one of them wins two games in a row or three games altogether.
- In what percent of all possible cases does the game end because Bill wins three games without winning two in a row?
- Solution.
- The tree diagram of Figure 2.1 illustrates all possible cases.
   The total number of possible cases is equal to the number of the endpoints of the branches, which is 10.
- The number of cases in which Bill wins three games without winning two in a row, as seen from the figure, is one.
- So the answer is 10%.
- Note that the probability of this event is not 0.10 because not all
  of the branches of the tree are equiprobable.

# Section 2.3

#### **Permutations**

#### Definition of Permutations

#### Distinguishable objects

• **Definition.** An *ordered* arrangement of r objects from a set A containing n objects ( $0 < r \le n$ ) is called an r-element permutation of A, or a permutation of the elements of A taken r at a time. The number of r -element permutations of a set containing n objects is denoted by  ${}_{n}P_{r}$ .

$$_{n}P_{r} = n(n-1)(n-2)\cdots(n-r+1)$$
 $_{n}P_{n} = n(n-1)(n-2)\cdots(n-n+1) = n!$ 
 $_{n}P_{r} = \frac{n!}{(n-r)!}$ 

## Example 2.15

- If five boys and five girls sit in a row in a random order, what is the probability that no two children of the same sex sit together?
- Solution.
- There are 10! ways for 10 persons to sit in a row.
- In order that no two of the same sex sit together, boys must occupy positions 1, 3, 5, 7, 9, and girls positions 2, 4, 6, 8, 10, or vice versa.
- In each case there are  $5! \times 5!$  possibilities.
- Therefore, the desired probability is equal to

$$\frac{2 \times 5! \times 5!}{10!} \approx 0.008$$



## Distinguishable and Indistinguishable

- The formula for the number of the permutations is valid only if all objects are distinguishable.
- For example, the number of permutations of the eight letters in **STANFORD** is 8!.
- However, the number of permutations of the letters in BERKELEY is less than 8!.
- **Theorem 2.4** The number of distinguishable permutations of n objects of k different types, where  $n_1$  are alike,  $n_2$  are alike, ...,  $n_k$  are alike and  $n = n_1 + n_2 + \cdots + n_k$ , is

$$\frac{n!}{n_1! \times n_2! \times \dots \times n_k!}$$

# Making Indistinguishable Objects Distinguishable

• Distinguish the 3 Es in BERKELEY by marking them  $E_1$ ,  $E_2$ , and  $E_3$ 

```
BYE_1RE_2LE_3K BYE_2RE_3LE_1K

BYE_1RE_3LE_2K BYE_3RE_1LE_2K

BYE_2RE_1LE_3K BYE_3RE_2LE_1K
```

= BERKELEY

8!/3!

## Example 2.17

- In how many ways can we paint 11 offices so that four of them will be painted green, three yellow, two white, and the remaining two pink?
- Solution.
- Let ``ggypgwpygwy'' represent the sitution in which the first two offices are painted green, the third one yellow, and so on, with similar representation for other cases.
- Then the answer is equal to the number of distinguishable permutations of ``ggypgwpygwy'' which by Theorem 2.4 is  $11!/(4! \times 3! \times 2! \times 2!)$

# Section 2.4

## **Combinations**

#### **Definition of Combinations**

• In many combinatorial problems, unlike permutations, the order in which objects are arranged is immaterial.

#### Distinguishable objects

• **Definition.** An **unordered** arrangement of r objects from a set A containing n objects  $(r \le n)$  is called an r-element combination of A, or a combination of the elements of A taken r at a time,

$$x \times r! = {}_{n}P_{r}$$

ABC BCA ACB CAB BAC CBA • The number of r-element combinations of n objects is given by

$$\begin{pmatrix} h \\ \gamma \end{pmatrix} \approx {}_{n}C_{r} = \frac{n!}{(n-r)! \, r!}$$

$$_{n}C_{r} \times r! = {_{n}P_{r}}$$
 the same which the combinates

#### Notations and Some Useful Identities

**Notation:** By the symbol  $\binom{n}{r}$  (read: n choose r) we mean the number of all r-element N1 = n. (n-1) ... 2.1 combinations of n objects. Therefore, for  $r \leq n$ , Ol det 1

$$\binom{n}{r} = \frac{n!}{r! (n-r)!}.$$

Observe that  $\binom{n}{0} = \binom{n}{n} = 1$  and  $\binom{n}{1} = \binom{n}{n-1} = n$ . Also, for any  $0 \le r \le n$ ,

$$\binom{n}{r} = \binom{n}{n-r}$$

#### A Useful Identity

$$\begin{pmatrix} n+1 \\ r \end{pmatrix} = \begin{pmatrix} n \\ r \end{pmatrix} + \begin{pmatrix} n \\ r-1 \end{pmatrix} \tag{2.4}$$

- Eq. (2.4) can be proved algebraically or verified combinatorially. Let us prove (2.4) by a combinatorial argument.
- Consider a set of n + 1 objects,  $\{a_1, a_2, \dots, a_n, a_{n+1}\}$ . There are  $\binom{n+1}{r}$  r-element combinations of this set.
- Now we separate these *r*-element combinations into two disjoint classes:
  - $\bigcirc$  one class consisting of all r-element combinations of  $\{a_1, a_2, \dots, a_n\}$  and
  - 2 another consisting of all (r-1)-element combinations of  $\{a_1, a_2, \ldots, a_n\}$  attached to  $a_{n+1}$ .
- The latter class contains  $\binom{n}{r-1}$  elements and the former contains  $\binom{n}{r}$  elements, showing that (2.4) is valid.

## An Alternative Explanation of (2.4)

$$\binom{n+1}{r} = \binom{n}{r} + \binom{n}{r-1} \tag{2.4}$$

- The left hand side of the above equation is the number of ways to pick r out of n + 1 objects.
- Tag one of the n + 1 objects. Call it A.
- There are two possibilities.
- In the first case, A is selected as one of the r objects.
  - In this case, we further select r-1 objects from the rest n objects.
  - The number of selections is the second term in (2.4).
- In the second case, A is **not** selected as one of the r objects.
  - In this case, we select r objects from the rest n objects.
  - The number of selections is the first term in (2.4).

## Example 2.23

- From an ordinary deck of 52 cards, seven cards are drawn at random and without replacement. What is the probability that at least one of the cards is a king?
- **Solution.** In  $\binom{52}{7}$  ways seven cards can be selected from an ordinary deck of 52 cards.
- In  $\binom{48}{7}$  of these, none of the cards selected is a king.
- Therefore, the desired probability is

$$P(\text{at least one king}) = 1 - P(\text{no kings}) = 1 - \frac{\binom{48}{7}}{\binom{52}{7}} = 0.4496$$

## A Common Mistake in solving Example 2.23

- A common mistake is to calculate this and similar probabilities as follows:
- To make sure that there is at least one king among the seven cards drawn, we will first choose a king;
  - there are  $\binom{4}{1}$  possibilities
- Then we choose the remaining six cards from the remaining 51 cards;
  - there are  $\binom{51}{6}$  possibilities
- Thus the answer is

$$\frac{\binom{4}{1}\binom{51}{6}}{\binom{52}{7}} = 0.5385$$

## Why is the Analysis on the Last Page Wrong?

- This solution is wrong because it counts some of the possible outcomes several times.
- For example, the hand  $K_H$ ,  $5_C$ ,  $6_D$ ,  $7_H$ ,  $K_D$ ,  $J_C$ , and  $9_S$  is counted twice:
  - once when  $K_H$  is selected as the first card from the kings and  $5_C$ ,  $6_D$ ,  $7_H$ ,  $K_D$ ,  $J_C$ , and  $9_S$  from the remaining 51, and
  - once when  $K_D$  is selected as the first card from the kings and  $K_H$ ,  $5_C$ ,  $6_D$ ,  $7_H$ ,  $J_C$ , and  $9_S$  from the remaining 51 cards.

## Example 2.25

• Show that the number of different ways n indistinguishable objects can be placed into k distinguishable cells is

$$\binom{n+k-1}{n} = \binom{n+k-1}{k-1}$$

Example: n = 4, k = 3







• This problem is equivalent to the number of non-negative solutions of the following equation

$$x_1 + x_2 + \dots + x_k = n$$

## Number of **Non-negative** Solutions

- Consider n indistinguishable objects and k-1 indistinguishable plates.
- The number of objects arranged between the (i-1)-th and the i-th plates is the value of  $x_i$ .



n = 6, k = 3

• The number of possible arrangement is

$$\frac{(n+k-1)!}{n!\,(k-1)!}$$

#### Number of **Positive** Solutions

Consider again the equation

$$x_1 + x_2 + \dots + x_k = n$$

- How many distinct positive integer solutions does it have?
- This problem can be changed into the number of non-negative integer solutions by introducing new variables.

$$y_i = x_i - 1$$
, for  $1 \le i \le k$   
 $y_1 + y_2 + \dots + y_k = n - k$   
 $\binom{(n-k)+k-1}{n-k} = \binom{n-1}{n-k} = \binom{n-1}{k-1}$ 

## Example 2.28

• An absentminded professor wrote *n* letters and sealed them in envelopes before writing the addresses on the envelopes.

 $\bullet$  Then he wrote the n addresses on the envelopes at random.

• What is the probability that at least one letter was addressed

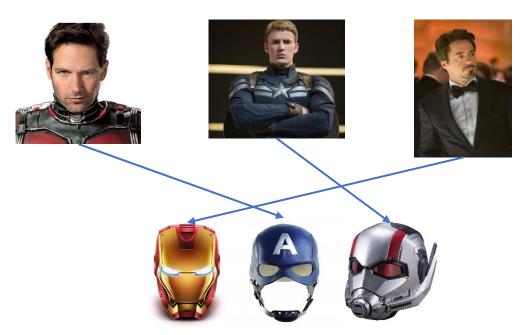
correctly?





## Equivalent version of Example 2.28

- n men throw their (distinct) helmets on the floor
- Every one picks up a helmet randomly
- What is the probability that at least one man picks up his helmet correctly?



- As n approaches infinity, does this probability approaches
  - 0, or
  - 1, or
  - something in between?

## Solution of Example 2.28

- Let  $E_i$  be the event that the *i*th man picks his hat correctly.
- Then,  $E_1 \cup E_2 \cup \cdots \cup E_n$  is the event that at least one man picks his hat correctly
- We use the inclusion-exclusion formula to calculate  $P(E_1 \cup E_2 \cup \cdots \cup E_n)$  (Chapter 1, page 20 of the textbook).

$$P\left(\bigcup_{i=1}^{n} E_{i}\right) = \sum_{i=1}^{n} P(E_{i}) - \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} P(E_{i}E_{j})$$

$$+ \sum_{i=1}^{n-2} \sum_{j=i+1}^{n-1} \sum_{k=j+1}^{n} P(E_{i}E_{j}E_{k})$$

$$- \cdots + (-1)^{n-1} P(E_{1}E_{2} \cdots E_{n})$$

## Solution Steps of Example 2.28

1 3 0 m H1 H2 H,

- First, we show that  $P(E_i) = (n-1)!/n!$
- Then, we show that  $P(E_i \cap E_j) = (n-2)!/n!$
- In general,

$$P(E_i \cap E_j \cap E_k) = (n-3)!/n!$$

- We need to count the number of terms in the inclusion-exclusion formula
  - There are *n* terms of the form  $P(E_i)$ .
  - There are  ${}_{n}C_{2}$  terms of the form  $P(E_{i} \cap E_{j})$ .
  - There are  ${}_{n}C_{3}$  terms of the form  $P(E_{i} \cap E_{j} \cap E_{k})$ .

• Substitute these results into the inclusion-exclusion formula.

$$P\left(\bigcup_{i=1}^{n} E_{i}\right) = n\frac{(n-1)!}{n!} - \binom{n}{2} \frac{(n-2)!}{n!} + \dots + \left(-1\right)^{n-2} \binom{n}{n-1} \frac{(n-(n-1))!}{n!} + (-1)^{n-1} \binom{n}{n} \frac{1}{n!}$$

• The expression above is simplified as

$$P\left(\bigcup_{i=1}^{n} E_{i}\right) = 1 - \frac{1}{2!} + \frac{1}{3!} - \frac{1}{4!} + \dots + \frac{(-1)^{n-1}}{n!}$$

$$= 1 - \sum_{i=0}^{n} \frac{(-1)^{i}}{i!}$$
P(no match) =  $1 - P\left(\bigcup_{i=1}^{n} E_{i}\right) = \sum_{i=0}^{n} (-1)^{i}/i!$ 

# Taylor Expansion of the Exponential Function

Taylor expansion of exponential functions

$$e^x = \sum_{i=0}^{\infty} \frac{x^i}{i!}$$

• Probability of at least one match is

$$\frac{1}{1} \approx 0.632$$

$$P\left(\bigcup_{i=1}^{n} E_i\right) \approx 1 - \sum_{i=0}^{\infty} \frac{x^i}{i!} = 1 - \frac{1}{e} \approx 0.632.$$

Probability of no match is

$$P(\text{no match}) = 1 - P\left(\bigcup_{i=1}^{n} E_i\right) = \sum_{i=0}^{\infty} \frac{x^i}{i!} \approx \frac{1}{e}.$$

## Theorem 2.5 (Binomial Expansion)

• **Theorem 2.5** For any integer  $n \ge 0$ ,

$$(x+y)^n = \sum_{i=0}^n \binom{n}{i} x^{n-i} y^i.$$

• **Proof.** From

$$(x+y)^n = (x+y)(x+y)\cdots(x+y) \tag{*}$$

we obtain only terms of the form  $x^{n-i}y^i$ ,  $0 \le i \le n$ .

- Therefore, all we have to do is to find out how many times the term  $x^{n-i}y^i$  appears,  $0 \le i \le n$ .
- $x^{n-i}y^i$  emerges because n-i pairs of parentheses in (\*) contribute x and i pairs contribute y.

## Example 2.30

• Evaluate the sum  $\binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{n}$ 

$$(x+y)^n = \sum_{i=0}^n \binom{n}{i} x^{n-i} y^i.$$

$$\sum_{i=0}^{n} \binom{n}{i} 1^{n-i} 1^{i} = (1+1)^{n} = 2^{n}.$$

## Example 2.31

- Evaluate the sum  $\binom{n}{1} + 2 \binom{n}{2} + \dots + n \binom{n}{n}$ .
- Solution.

$$i\binom{n}{i} = i \cdot \frac{n!}{i! (n-i)!} = \frac{n \cdot (n-1)!}{(i-1)! (n-i)!} = n\binom{n-1}{i-1}$$

So

$$\binom{n}{1} + 2\binom{n}{2} + \dots + n\binom{n}{n} = n\left[\binom{n-1}{0} + \binom{n-1}{1} + \binom{n-1}{2} + \dots + \binom{n-1}{n-1}\right] = n \cdot 2^{n-1}$$

by example 2.30

## Theorem 2.6 Multinomial Expansion

• **Theorem 2.6** *In the expansion of* 

$$(x_1+x_2+\cdots+x_k)^n,$$

the coefficient of the term

$$x_1^{n_1}x_2^{n_2}\cdots x_k^{n_k}, \quad n_1+n_2+\cdots+n_k=n$$

is

$$\frac{n!}{n_1!n_2!\cdots n_k!}.$$

Therefore,

$$(x_1 + x_2 + \dots + x_k)^n = \sum_{n_1 + n_2 + \dots + n_k = n} \frac{n!}{n_1! n_2! \cdots n_k!} x_1^{n_1} x_2^{n_2} \cdots x_k^{n_k}.$$

#### Proof of Theorem 2.6

- Distribute n distinguishable balls into k distinguishable cells so that
  - $n_i$  balls are placed in cell i, i = 1, 2, ..., k
  - $\bullet \ n_1 + n_2 + \dots + n_k = n$
- How many ways are there?

$$\binom{n}{n_1} \binom{n-n_1}{n_2} \binom{n-n_1-n_2}{n_3} \cdots \binom{n-n_1-n_2-\cdots-n_{k-1}}{n_k}$$

$$= \frac{n!}{(n-n_1)! n_1!} \times \frac{(n-n_1)!}{(n-n_1-n_2)! n_2!} \times \frac{(n-n_1-n_2)!}{(n-n_1-n_2-n_3)! n_3!} \times \cdots$$

$$\times \frac{(n-n_1-n_2-n_3-\cdots-n_{k-1})!}{(n-n_1-n_2-n_3-\cdots-n_{k-1}-n_k)! n_k!} = \frac{n!}{n_1! n_2! n_3! \cdots n_k!}$$

# Section 2.5

## Stirling's Formula

## Theorem 2.7 Stirling's Formula

• **Theorem 2.7** (Stirling's Formula)

$$n! \sim \sqrt{2\pi n} n^n e^{-n}$$

where the sign  $\sim$  means

$$\lim_{n \to \infty} \frac{n!}{\sqrt{2\pi n} n^{ne^{-n}}} = 1$$

## Stirling's Formula is very accurate

n	n!	$\sqrt{2\pi n} n^n e^{-n}$	R(n)
1	1	0.922	1.084
2	2	1.919	1.042
5	120	118.019	1.017
8	40,320	39,902.396	1.010
10	3,628,800	3,598,695.618	1.008
12	479,001,600	475,687,486.474	1.007

#### Homework 2

• Section 2.2: B.30, B.36

• Section 2.3: A.29, B.34

• Section 2.4: A.47, B.51, B.66

• Due date: 5 pm, Wednesday, March 15, 2022