



# CS215 DISCRETE MATH

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# Three Different Behaviors

- Compare the iteration for the recurrences

$$T(n) = 2T(n/2) + n$$

$$T(n) = T(n/2) + n$$

$$T(n) = 4T(n/2) + n$$



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- ◇ all three recurrences iterate  $\log_2 n$  times
- ◇ in each case, size of subproblem in next iteration is **half** the size in the preceding iteration level



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- **Theorem** Suppose that we have a recurrence of the form

$$T(n) = aT(n/2) + n,$$

where  $a$  is a positive integer and  $T(1)$  is nonnegative. Then we have the following big  $\Theta$  bounds on the solution:

1. If  $a < 2$ , then  $T(n) = \Theta(n)$ .
2. If  $a = 2$ , then  $T(n) = \Theta(n \log n)$ .
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## Proof

We already proved Case 1 when  $a = 1$  in Example 3.  
(will not prove it for  $1 < a < 2$ )

We already proved Case 2 in Example 1.

We will now prove Case 3.



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$$T(n) = a^{\log_2 n} T(1) + n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i$$

Work at  
“bottom”

Iterated  
Work





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Since  $a > 2$ , the geometric series is  $\Theta$  of the largest term.

$$n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i = n \frac{1 - (a/2)^{\log_2 n}}{1 - a/2} = n\Theta((a/2)^{\log_2 n - 1})$$

# Total work

- $n$  times the largest term in the geometric series is

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This matches with the exact answer of  $2n^2 - n$ .



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# The Master Theorem

- **Theorem** Suppose that we have a recurrence of the form

$$T(n) = aT(n/b) + cn^d,$$

where  $a$  is a positive integer,  $b \geq 1$ ,  $c, d$  are real numbers with  $c$  positive and  $d$  nonnegative, and  $T(1)$  is nonnegative. Then we have the following **big  $\Theta$**  bounds on the solution:

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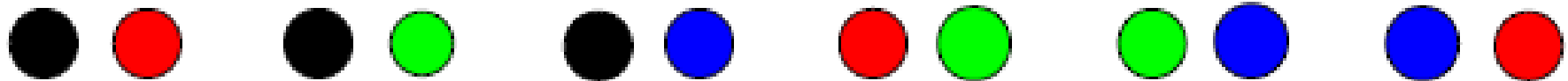
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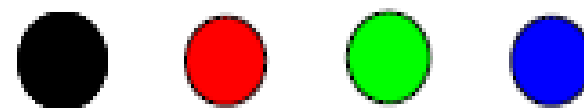
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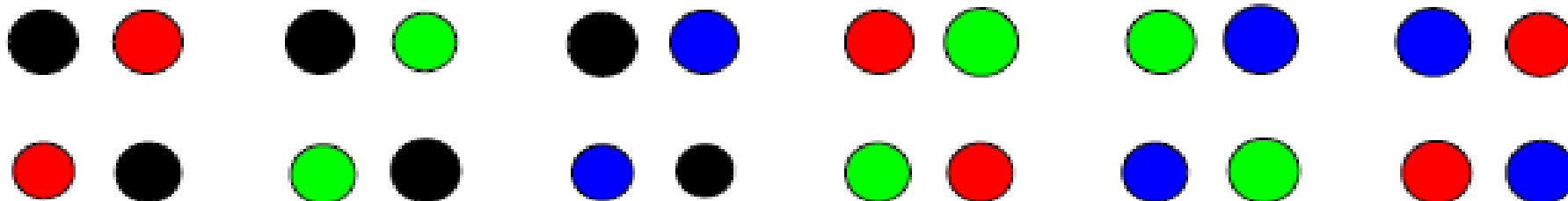
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Counting may be very hard, not trivial.

- simplify the solution by decomposing the problem



# Basic Counting Rules

- *the Product Rule*

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In an auditorium, the seats are labeled by a letter and numbers in between 1 to 50 (e.g., A23). What is the total number of seats?

We may either list all or use the product rule.

$$26 \times 50 = 1300$$



# The Product Rule

- **Product Rule:** If a count of elements can be broken down into a **sequence of dependent counts** where the first count yields  $n_1$  elements, the second  $n_2$  elements, and  $k$ th count  $n_k$  elements, then the total number of elements is

$$n = n_1 \cdot n_2 \cdot \dots \cdot n_k$$



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How many **onto** functions?



# The Product Rule

- The following loop is a part of program computing the product of two matrices.

```
(1) for i = 1 to r
(2)   for j = 1 to m
(3)     S = 0
(4)     for k = 1 to n
(5)       S = S + A[i,k] * B[k,j]
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How many **multiplications** (in terms of  $r, m, n$ ) does this program carry out in total among all iterations of line 5?



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You need to travel from city A to B. You may either fly, take a train, or a bus. There are 12 different flights, 5 different trains and 10 buses. **How many options do you have to get from A to B?**



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We may **use the sum rule.**

$$12 + 5 + 10$$



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$$n = n_1 + n_2 + \cdots + n_k$$



# The Sum Rule

- The following loop is from [selection sort](#).

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(1) for i = 1 to n-1
(2)     for j = i+1 to n
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How many **comparisons** (in terms of  $n$ ) does this program carry out in total among all iterations of line 3?

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## Example

Each password is **6 to 8 characters** long, where each character is a lowercase letter or a digit. Each password must contain **at least one digit**. How many possible passwords are there?



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Each password is 6 to 8 characters long, where each character is a lowercase letter or a digit. Each password must contain at least one digit. How many possible passwords are there?

$$P = P_6 + P_7 + P_8$$



# Tree Diagrams

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## Example

What is the number of bit strings of length 4 that **do not have two consecutive 1's**?



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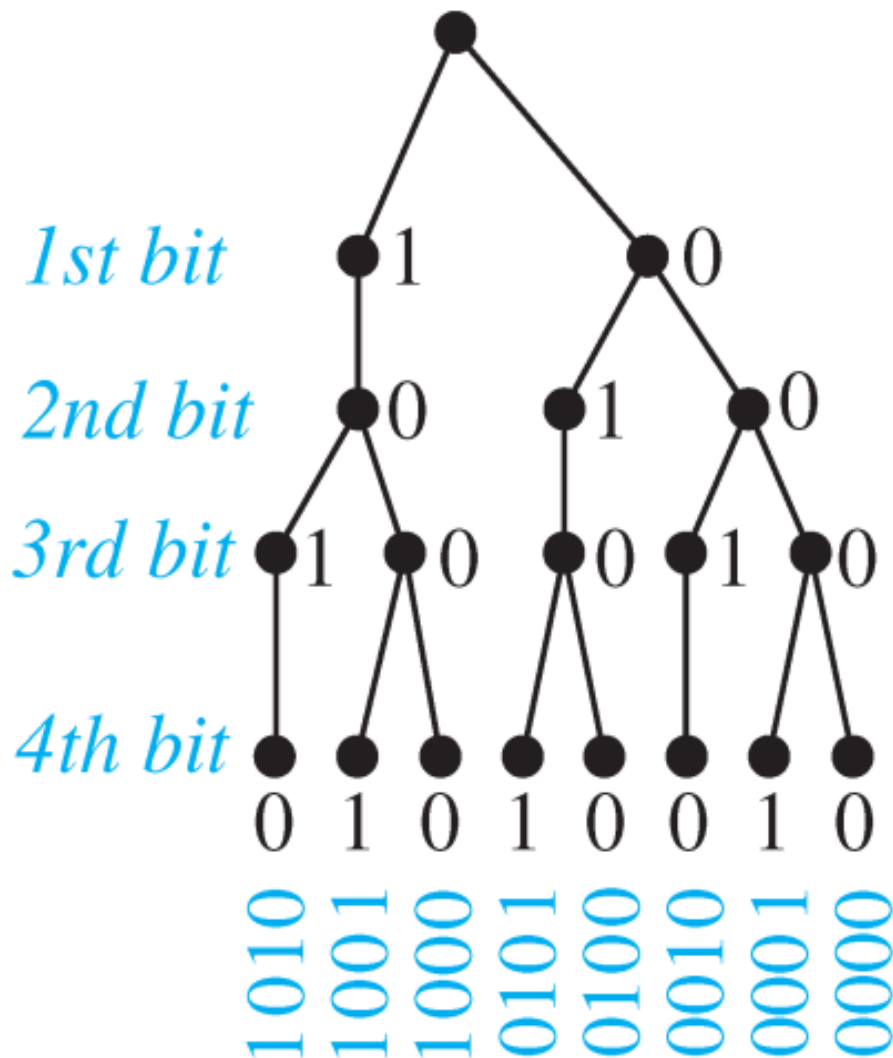
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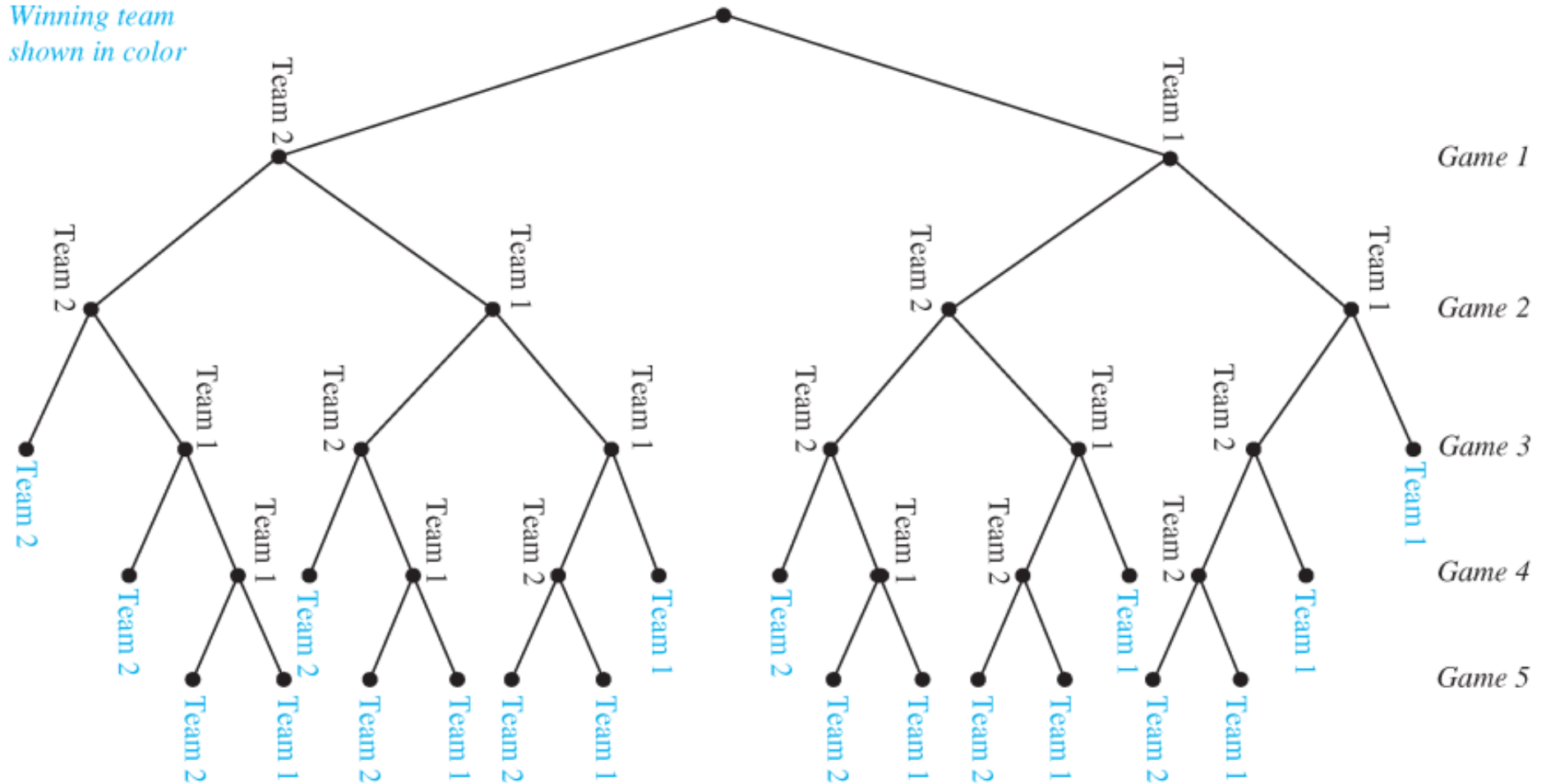
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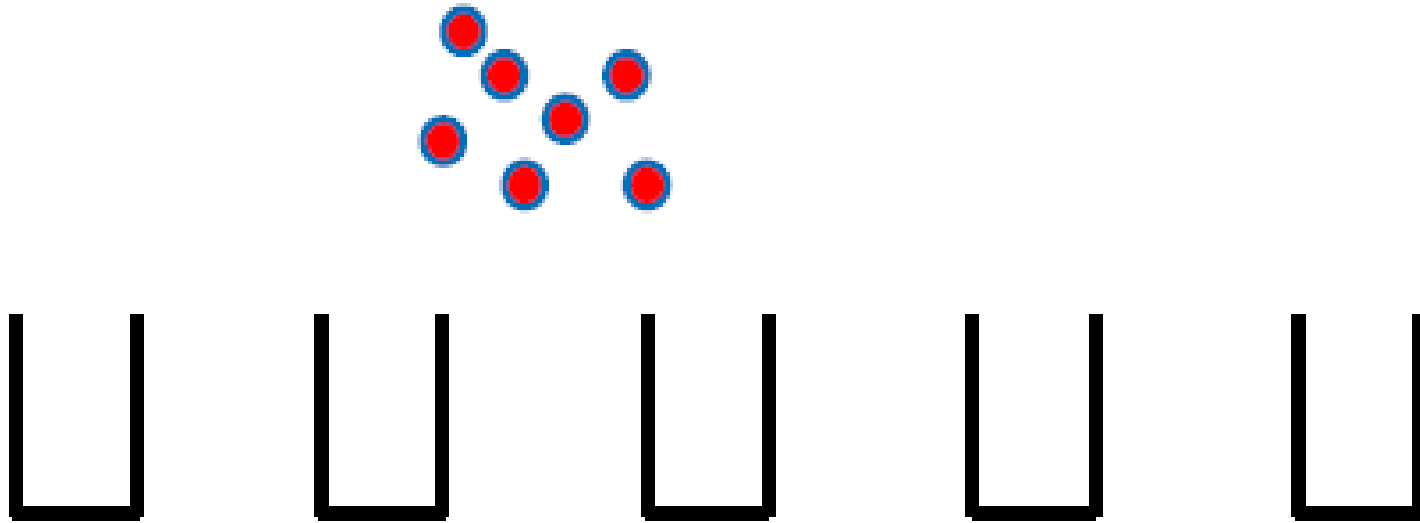


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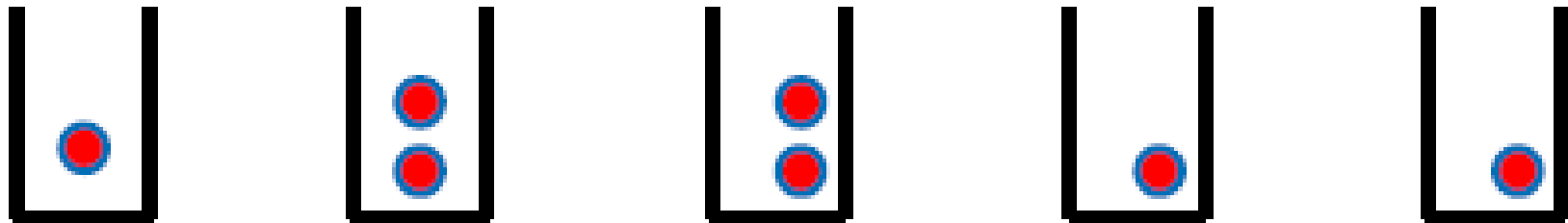


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## Proof by contradiction

## Example

Assume that there are 367 students. Are there any two people who have the same birthday?

There are 5 bins and 12 objects. Then there must be a bin with at least 3 objects. Why?



# Generalized Pigeonhole Principle

- If  $N$  objects are placed into  $k$  bins, then there is at least one bin containing at least  $\lceil N/k \rceil$  objects.





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## Example

Assume there are 100 students. How many of them were born in the same month?



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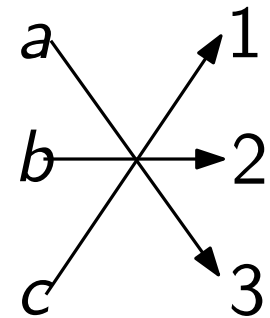
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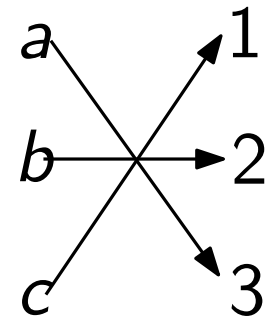


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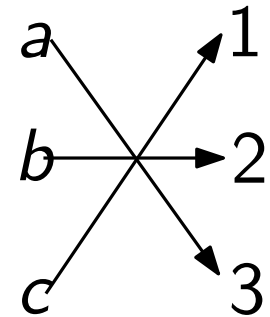
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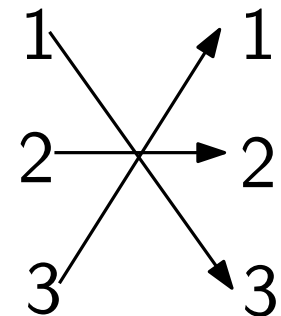
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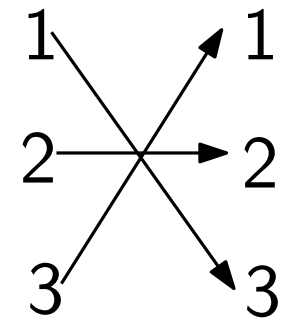
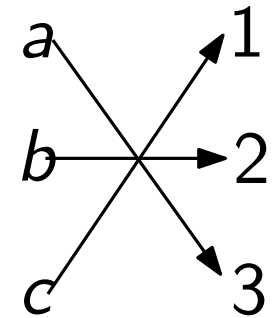
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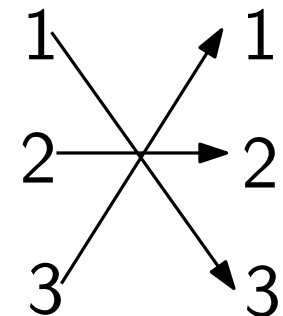
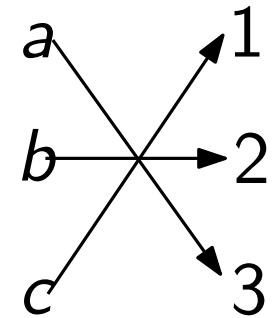
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Thus,

**the left and right sides must have the same size.**





# The Bijection Principle

- The following loop is a part of program to determine the number of triangles formed by  $n$  points in the plane.

```
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```

Among all iterations of line 5, what is the total number of times this line checks three points to see if they are collinear?

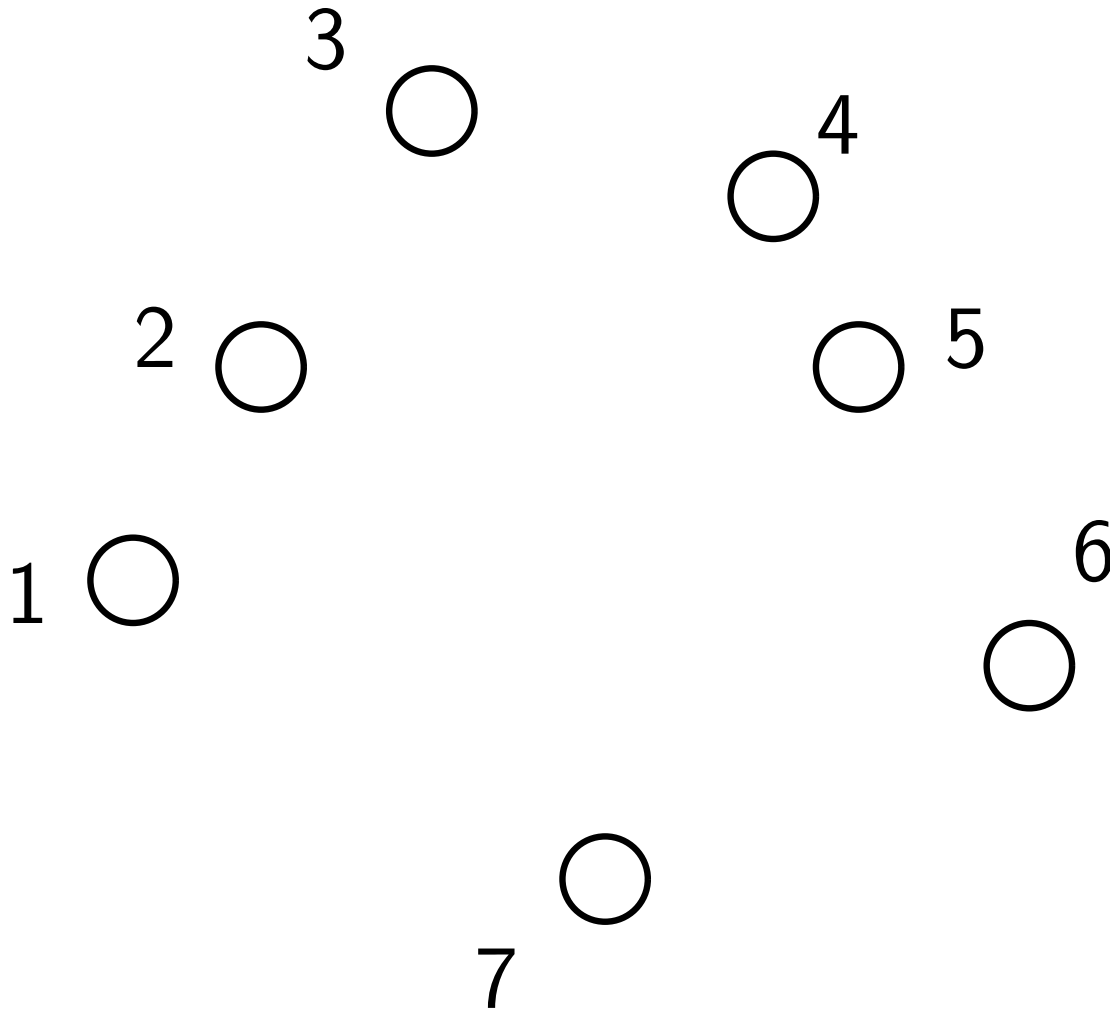
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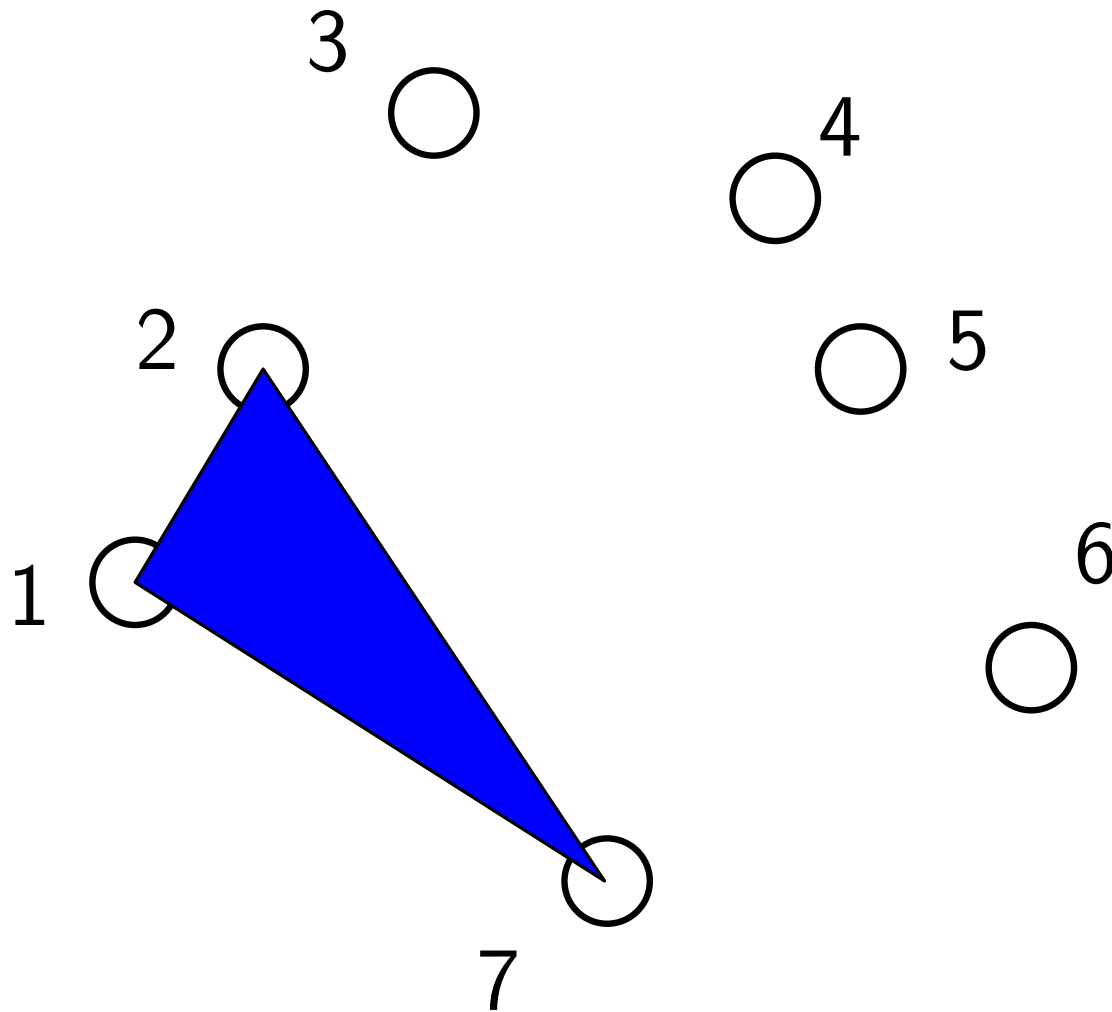


$$28 - 2$$



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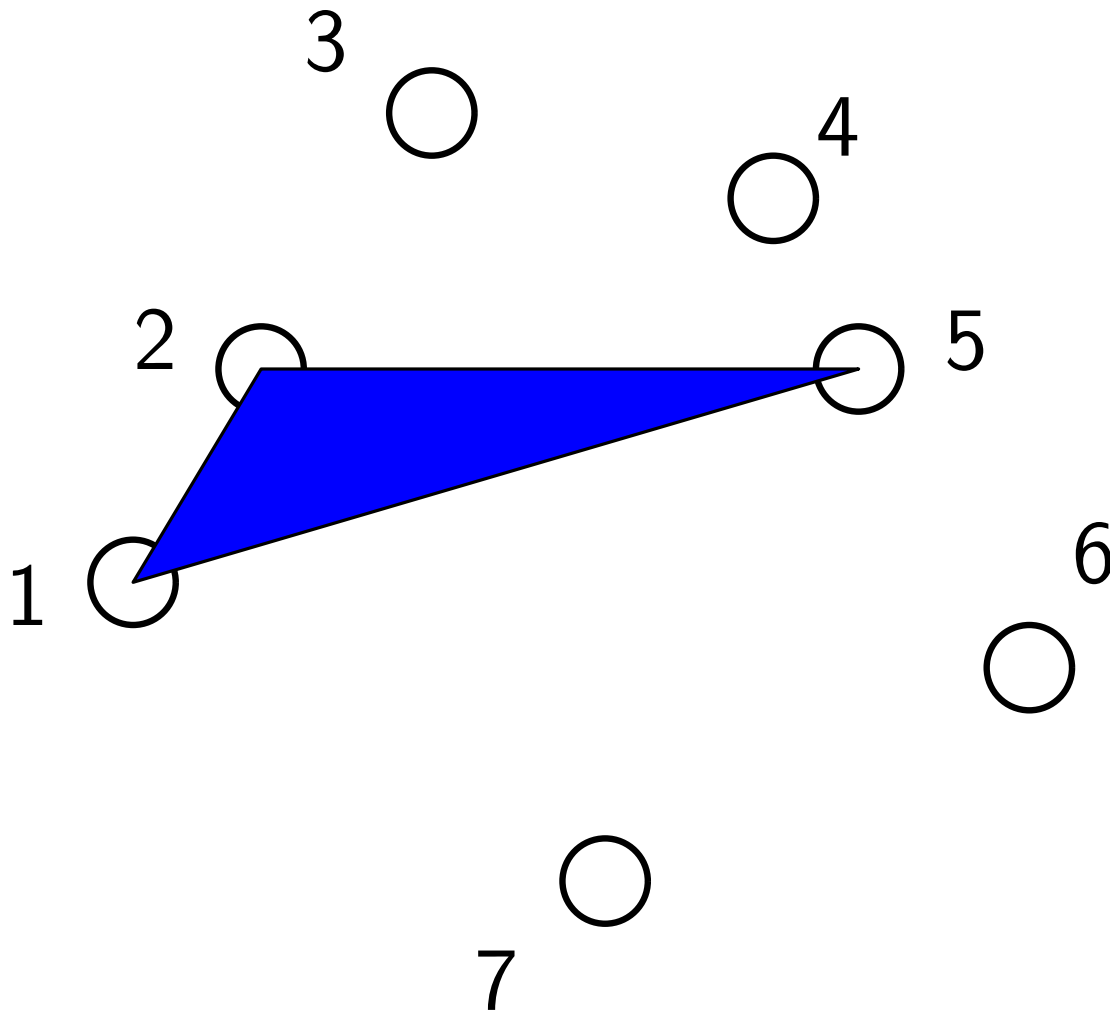
1 – 2 – 7: yes

28 - 3



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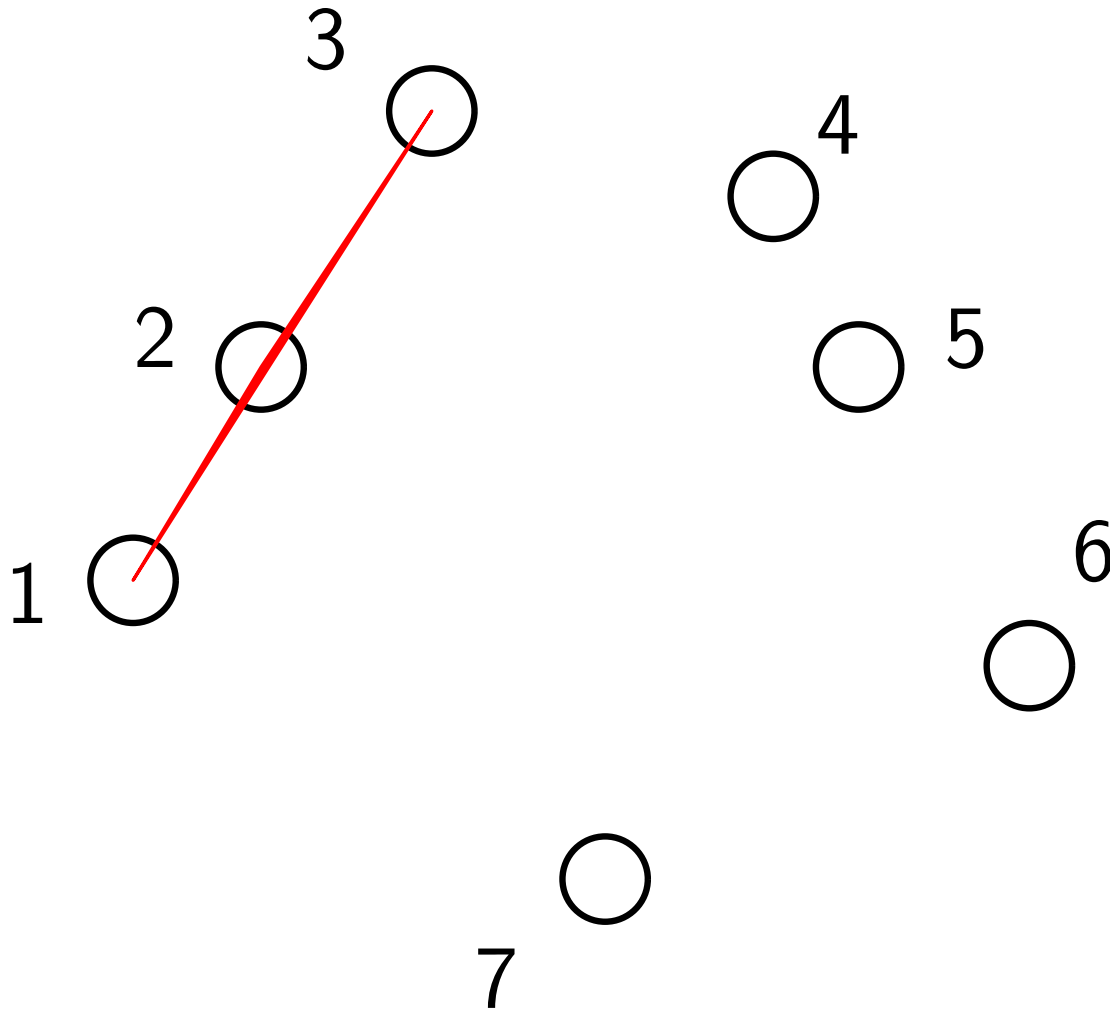


1 – 2 – 7: yes

1 – 2 – 5: yes

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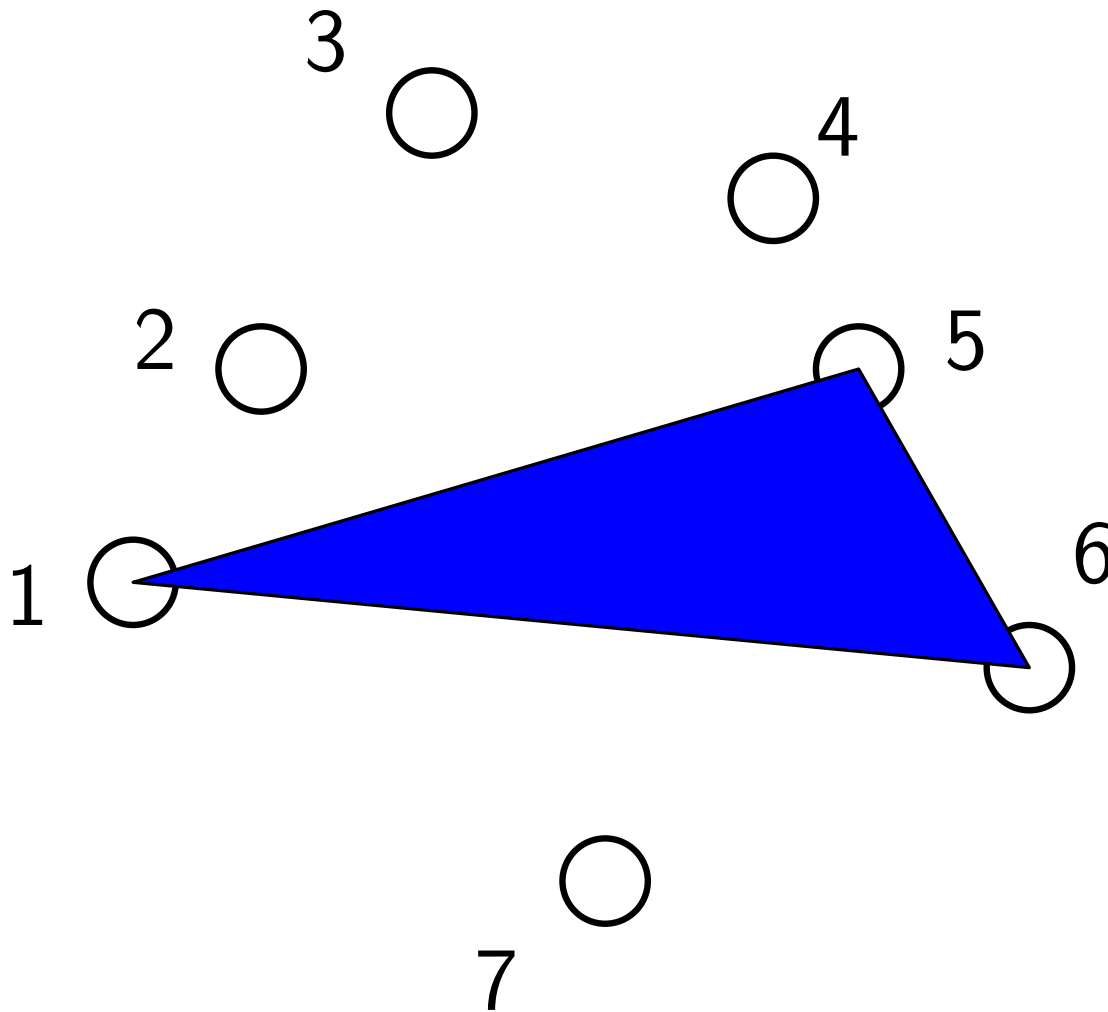
1 – 2 – 7: yes

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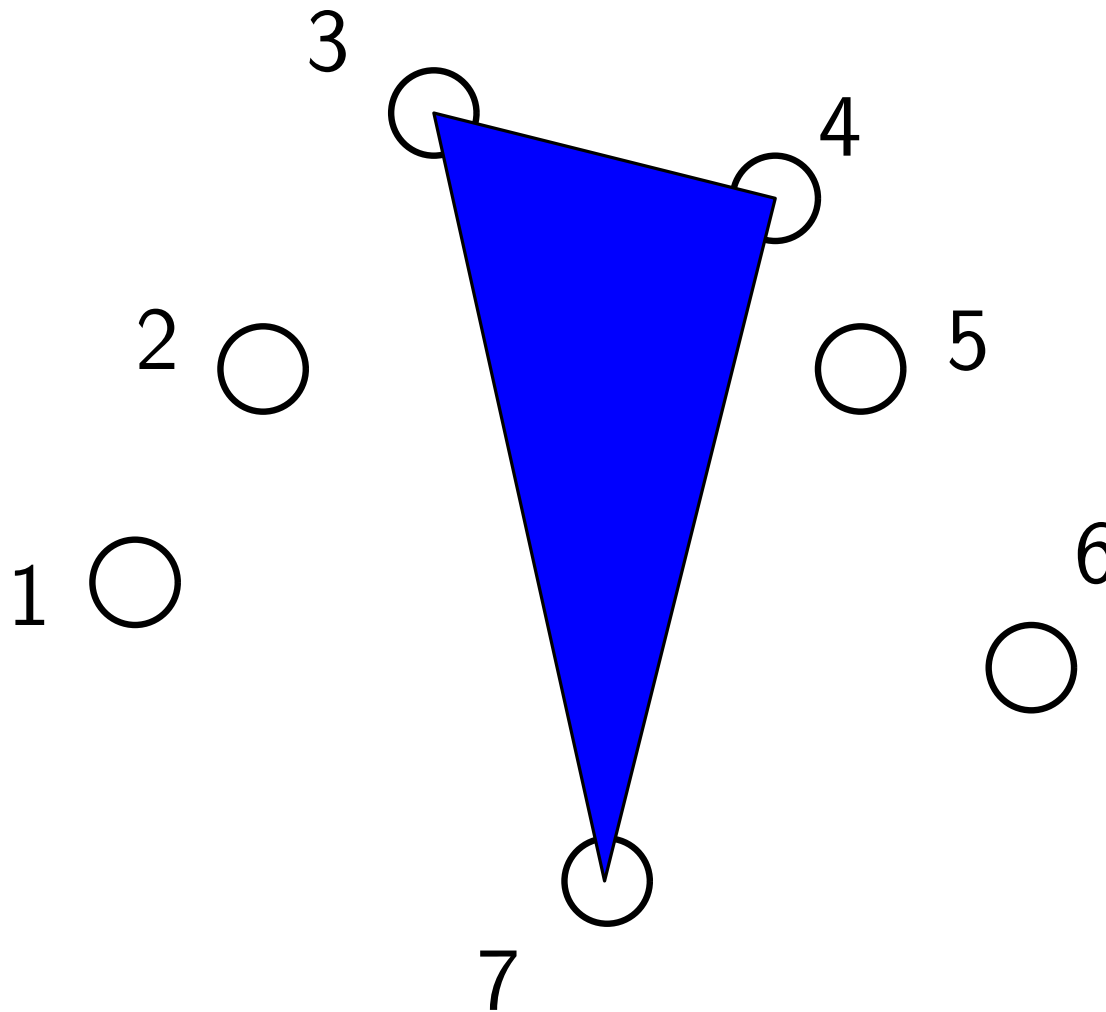
1 – 2 – 3: no

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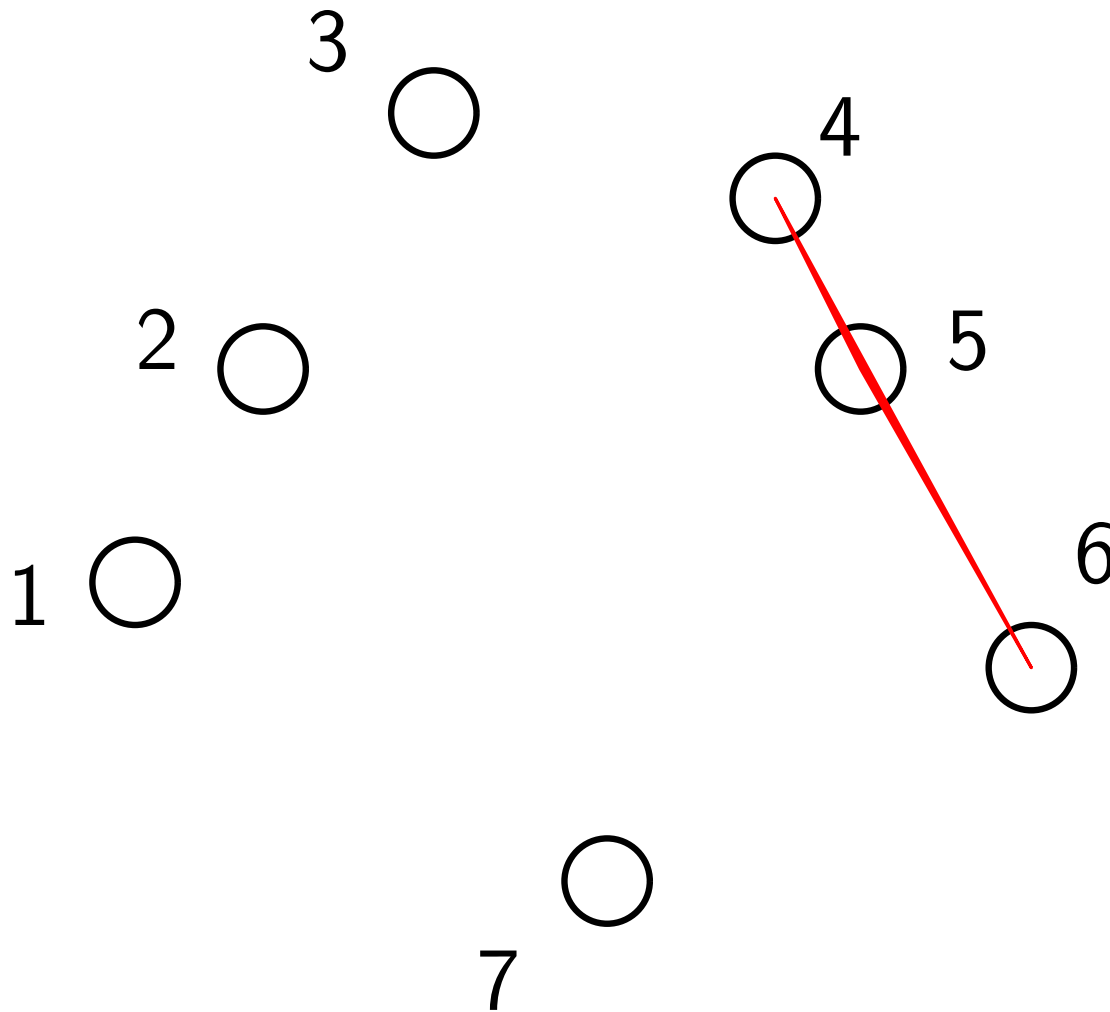
1 – 2 – 3: no

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3 – 4 – 7: yes

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1 – 2 – 7: yes

1 – 2 – 5: yes

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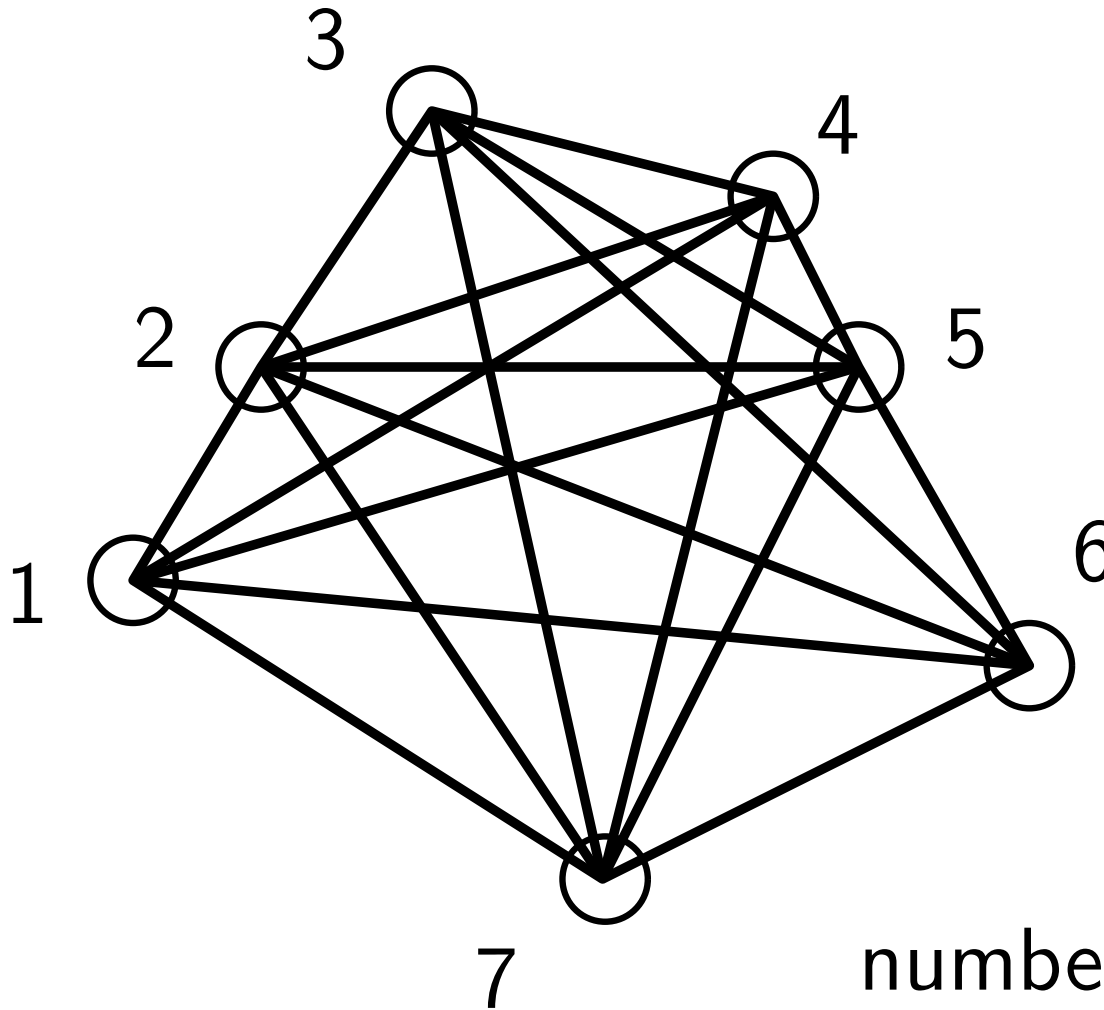
1 – 5 – 6: yes

3 – 4 – 7: yes

4 – 5 – 6: **no**

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4 – 5 – 6: no

number of triangles: 33

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A loop embedded in a loop

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For example, if  $n = 4$ , then triples  $(i, j, k)$  used by algorithm are  $(1, 2, 3)$ ,  $(1, 2, 4)$ ,  $(1, 3, 4)$ , and  $(2, 3, 4)$ .

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$f$  is a bijection because

$f$  is one-to-one

if  $(i, j, k) \neq (i', j', k') \Rightarrow f((i, j, k)) \neq f((i', j', k'))$

$f$  is onto

if  $\gamma$  is a 3-element subset then it can be written as  $\gamma = \{i, j, k\}$

where  $i < j < k$  so  $f((i, j, k)) = \gamma$ .

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We actually already saw that  $|X| = |Y| = \binom{n}{2}$



# The Bijection Principle

- Two sets **have the same size** if and only if there is a **one-to-one function from one set onto the other**.



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Currently, we started with the problem of counting the **# of increasing triples** and changed it to the problem of counting the **# of 3-element sets from  $\{1, 2, \dots, n\}$**



# Inclusion-Exclusion Principle

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**Inclusion-Exclusion Principle:** uses a sum rule and then corrects for the overlapping elements.

$$|A \cup B| = |A| + |B| - |A \cap B|$$



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Overcounting!!!





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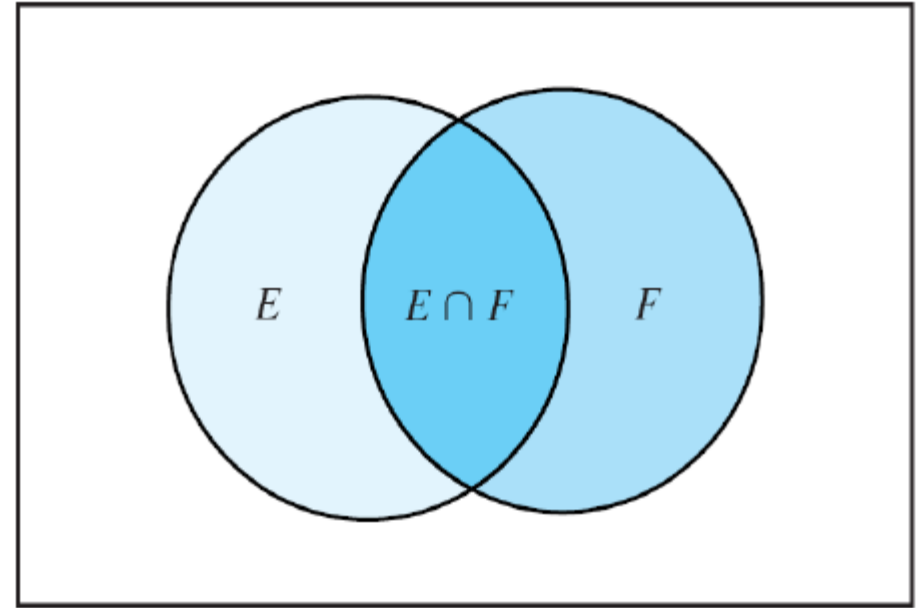
◇ deduct the number of strings starting with '1' and ending with "00":  $2^5$



# Inclusion-Exclusion Principle

- Two sets

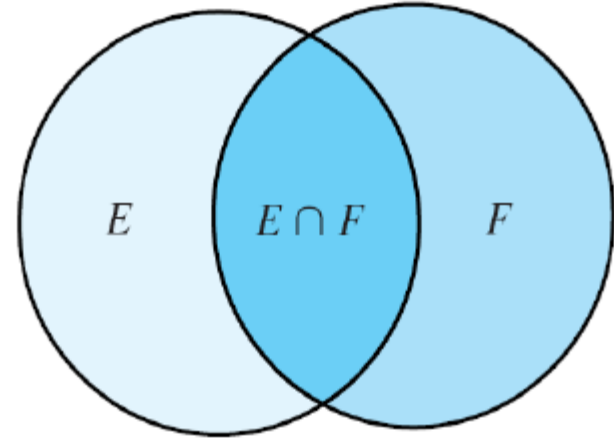
$$|E \cup F| = |E| + |F| - |E \cap F|$$



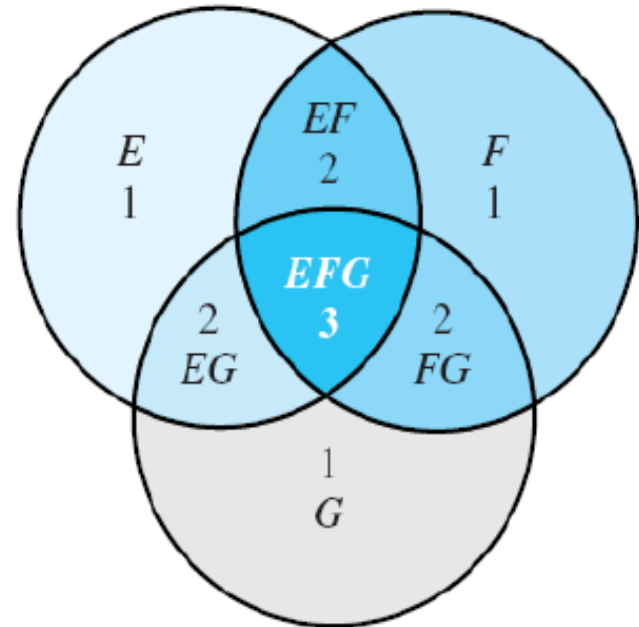
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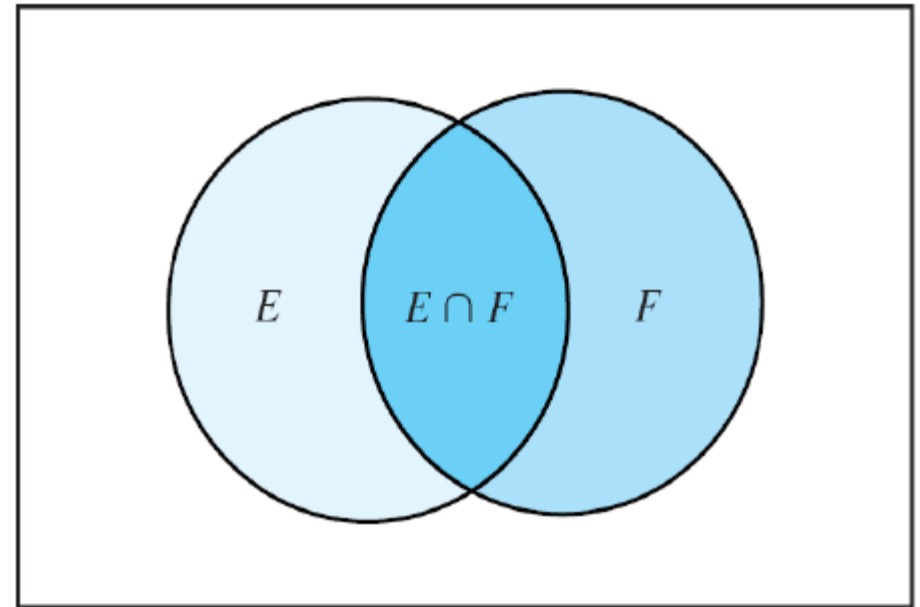
## Three sets



# Inclusion-Exclusion Principle

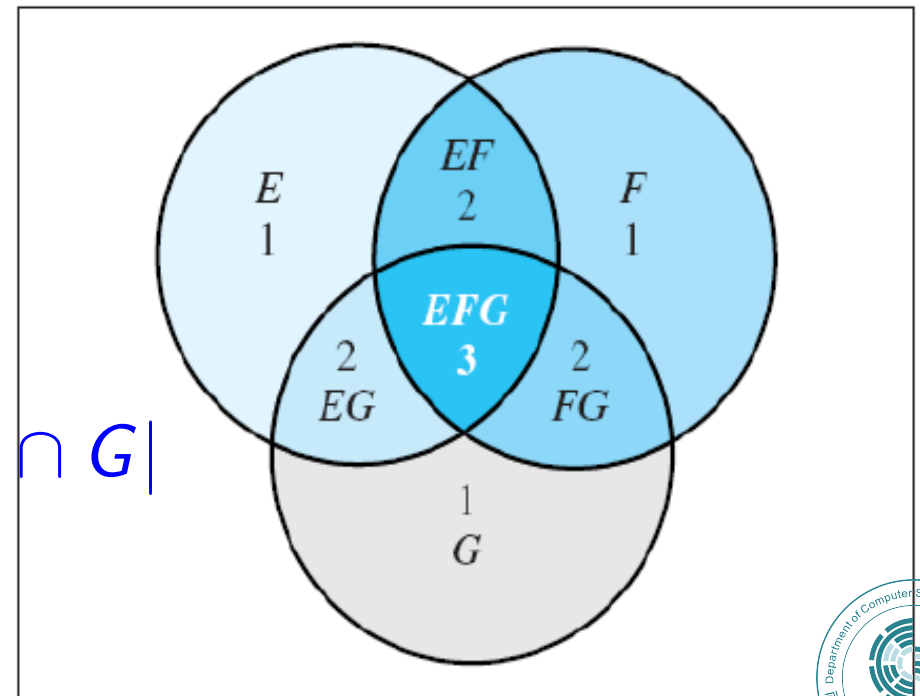
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## Three sets

$$\begin{aligned} &|E \cup F \cup G| \\ &= |E| + |F| + |G| \\ &\quad - |E \cap F| - |E \cap G| - |F \cap G| \\ &\quad + |E \cap F \cap G| \end{aligned}$$



# Inclusion-Exclusion Principle



$$|\cup_{i=1}^n E_i| = \sum_{k=1}^n (-1)^{k+1} \sum_{1 \leq i_1 < i_2 < \dots < i_k \leq n} |E_{i_1} \cap E_{i_2} \cap \dots \cap E_{i_k}|$$

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**Proof by induction**



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Base case ( $n = 2$ )

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Inductive Hypothesis

$$|\cup_{i=1}^{n-1} E_i| = \sum_{k=1}^{n-1} (-1)^{k+1} \sum_{1 \leq i_1 < i_2 < \dots < i_k \leq n-1} |E_{i_1} \cap E_{i_2} \cap \dots \cap E_{i_k}|$$



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Set  $E = E_1 \cup \cdots \cup E_{n-1}$ , and  $F = E_n$ .



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$$|\cup_{i=1}^n E_i| = |\cup_{i=1}^{n-1} E_i| + |E_n| - |(\cup_{i=1}^{n-1} E_i) \cap E_n|$$

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For the third term, by distributive law,

$$|(\cup_{i=1}^{n-1} E_i) \cap E_n| = |\cup_{i=1}^{n-1} (E_i \cap E_n)| = |\cup_{i=1}^{n-1} G_i|$$

where  $G_i = E_i \cap E_n$ .

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Some discussion:

**first summation** sums  $(-1)^{k+1} |E_{i_1} \cap E_{i_2} \cap \dots \cap E_{i_k}|$  over **all lists**  $i_1, i_2, \dots, i_k$  that **do not contain**  $n$   
**do not contain**  $n$  and **second summation** together sum  $(-1)^{k+1} |E_{i_1} \cap E_{i_2} \cap \dots \cap E_{i_k}|$  over **all lists**  $i_1, i_2, \dots, i_k$  that **do contain**  $n$

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(a) How many onto functions are there from  $A$  to  $B$ ?

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$$\#(a) + \#(b) = n^m$$

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Note that the case of  $k = n$  is special;

An  **$n$ -element permutation** of a **set  $N$**  of size  $|N| = n$  is what we earlier simply called a **permutation**.



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Ex: When  $n = 4$ , there are  $4 \times 3 \times 2 = 24$   
3 -element permutations of  $\{1, 2, 3, 4\}$

$L = \{123, 124, 132, 134, 142, 143, 213, 214, 231, 234, 241, 243, 312, 314, 321, 324, 341, 342, 412, 413, 421, 423, 431, 432\}.$



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Note: This type of "dictionary" ordering of tuples (assuming that we treat numbers the same as letters) is called a **lexicographic ordering** and is used quite often.



# $k$ -Element Permutations of a Set

- **Theorem** If  $N$  is a positive integer and  $k$  is an integer with  $1 \leq k \leq n$ , then there are

$$P(n, k) = n(n-1)(n-2) \cdots (n-k+1)$$

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$$P(n, 3) = 3! \cdot C(n, 3)$$



# Binomial Coefficient

- **Theorem** For integers  $n$  and  $k$  with  $0 \leq k \leq n$ , the number of  $k$ -element subsets of an  $n$ -element set is

$$\binom{n}{k} = C(n, k) = \frac{P(n, k)}{k!} = \frac{n!}{k!(n - k)!}.$$

This is the number of  $k$ -combinations of a set with  $n$  elements.



# Some Properties of Binomial Coefficients

- $\binom{n}{k} = \frac{n!}{k!(n-k)!}$  is the number of  $k$ -element subsets of an  $n$ -element set.

$$\binom{n}{0} = 1 \text{ only one set of size } 0.$$

$$\binom{n}{n} = 1 \text{ only one set of size } n.$$

$\binom{n}{k} = \binom{n}{n-k}$  Obvious from equation. Can you think of a simple bijection that explains this?

# Some Properties of Binomial Coefficients (cont.)



$$\sum_{i=0}^n \binom{n}{i} = 2^n$$



# Some Properties of Binomial Coefficients (cont.)

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Use Sum Rule

Let  $P$  = set of all subsets of  $\{1, 2, \dots, n\}$

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$$\Rightarrow |P| = \sum_{i=0}^n |S_i| = \sum_{i=0}^n \binom{n}{i}$$



# Some Properties of Binomial Coefficients (cont.)

■ Let  $L = L_1 L_2 \dots L_n$  be a list of size  $n$  from  $\{0, 1\}$

If  $\mathcal{L}$  = set of all such lists  $\Rightarrow |\mathcal{L}| = 2^n$

There is a *bijection* between  $\mathcal{L}$  and  $P$  so

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If  $L \in \mathcal{L}$  then  $f(L)$  is the set  $S \subseteq \{1, 2, \dots, n\}$  defined by

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Ex:  $n = 5$

$$f(10101) = \{1, 3, 5\}, \quad f(11101) = \{1, 2, 3, 5\}, \quad f(00000) = \emptyset$$

# Binomial Coefficients

$n \backslash k$	0	1	2	3	4	5	6
0	1						
1	1	1					
2	1	2	1				
3	1	3	3	1			
4	1	4	6	4	1		
5	1	5	10	10	5	1	
6	1	6	15	20	15	6	1

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Sum of items on  $n$ -th row is  $2^n$

# Pascal's Triangle

Take the table

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and shift each row slightly  
so that middle element is  
in middle

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# Pascal's Triangle

				1				
			1		1			
		1		2		1		
	1		3		3		1	
	1	4		6		4	1	
1		5	10		10	5		1
1	6	15	20	15	6		1	



# Pascal's Triangle

				1				
			1		1			
		1		2		1		
	1		3		3		1	
	1	4		6		4	1	
1		5	10		10	5		1
1	6	15	20	15	6		1	

What is the next row in the table?

# Pascal's Triangle

				1				
			1		1			
		1		2		1		
	1		3		3		1	
	1	4		6		4		1
	1	5	10		10	5		1
	1	6	15	20		15	6	1
1	7	21	35	35	21	7		1





# Pascal's Triangle

				1				
			1		1			
		1		2		1		
	1		3		3		1	
	1	4		6		4		1
	1	5	10		10	5		1
	1	6	15	20		15	6	1
1	7	21	35	35	21	7		1

## Pascal identity

Each (non-1) entry in Pascal's Triangle is the sum of the two entries directly above it (to left and to right).

# Pascal's Triangle

				1				
			1		1			
		1		2		1		
	1		3		3		1	
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A purely *algebraic* proof (manipulating formulas) is possible.

We will use a *combinatorial proof*.



# A Combinatorial Proof

- $\binom{n}{k}$  is the number of  $k$ -element subsets of an  $n$ -element set.



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

Therefore, each term (left and right) represents the number of subsets of a particular size chosen from an appropriately sized set.

# A Combinatorial Proof



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





# A Combinatorial Proof

■

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

Number of  $k$ -subsets of an  $n$ -element set.



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

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Number of  $(k-1)$ -subsets of an  $(n-1)$ -element set.



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Try to use sum principle to explain relationship among these three terms.

Example:  $n = 5, k = 2$

$$\binom{5}{2} = \binom{4}{1} + \binom{4}{2}.$$

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Set  $S_1$  of 2-subsets of  $S$  can be partitioned into 2 disjoint parts.

$S_2$  the 2-subsets that contain  $E$  and

$S_3$ , the set of 2-subsets that do not contain  $E$ .

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# A Combinatorial Proof

- If  $n$  and  $k$  are integers satisfying  $0 < k < n$ , then

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**Proof:** Apply **sum rule**.

Let  $S_1$  be set of all  $k$ -element subsets.

To apply **sum rule**, partition  $S_1$  into  $S_2$  and  $S_3$ .

Let  $S_2$  be set of  $k$ -element subsets that **contain**  $x_n$ .

Let  $S_3$  be set of  $k$ -element subsets that **don't contain**  $x_n$ .



# Blaise Pascal

Born 1623; Died 1662

French Mathematician

A Founder of Probability Theory

Inventor of one of the first mechanical  
calculating machines

**Pascal** Programming Language named for him



# Next Lecture

- counting II ...

