# Mathematical Review

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# Inductive Proof

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# Inductive Proof









# Inductive Proof

# 3 Logarithms







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# Inductive Proof

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# Permutations

# **5** Combinations

# 6 The Binomial Theorem





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# 6 The Binomial Theorem

# O Summations

# Order of magnitude of functions

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# Motivation

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*Order theory* enables us to compare functions, just as the theory of arithmetic enables us to compare numbers.

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(i) Which function grows faster:  $100 \cdot x^2$  or  $\frac{1}{10^6} \cdot x^3$ ?

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- (i) Which function grows faster:  $100 \cdot x^2$  or  $\frac{1}{10^6} \cdot x^3$ ?
- (ii) Which function grows faster:  $x^2 10$  or x + 10?

### Order of Magnitude Inductive Proof

Inductive Proof Logarithms Permutations Combinations The Binomial Theorem Summations

# Order of Magnitude (contd.)

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# Definition

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Let f and g be functions mapping non-negative reals to non-negative reals.

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Let f and g be functions mapping non-negative reals to non-negative reals.

Then f = O(g), if there exist constants c and  $n_0$  such that for all  $n \ge n_0$ ,  $f(x) \le c \cdot g(x)$ .

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Let f and g be functions mapping non-negative reals to non-negative reals.

Then  $f = \Omega(g)$ , if there exist constants c and  $n_0$  such that for all  $n \ge n_0$ ,  $f(x) \ge c \cdot g(x)$ .

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# Definition

Let f and g be functions mapping non-negative reals to non-negative reals.

Then 
$$f = \Omega(g)$$
, if there exist constants  $c$  and  $n_0$  such that for all  $n \ge n_0$ ,  $f(x) \ge c \cdot g(x)$ .

### Definition

Let f and g be functions mapping non-negative reals to non-negative reals.

Then f = o(g), if there exist constants c and  $n_0$  such that for all  $n \ge n_0$ ,  $f(x) < c \cdot g(x)$ .

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Then  $f = \Theta(g)$ , if f = O(g) and g = O(f).

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Order of Magnitude Inductive Proof Logarithms Permutations The Binomial Theorem

# Examples

Summations

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(i) Let 
$$f(x) = 2 \cdot x^2 - 2$$
 and  $g(x) = \frac{1}{100} \cdot x^2 - 100$ .

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 and  $g(x) = \frac{1}{100} \cdot x - 100$ .

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$$f(x) = 2 \cdot x^2 - 2$$
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## Examples

- (i) Let  $f(x) = 2 \cdot x^2 2$  and  $g(x) = \frac{1}{100} \cdot x^2 100$ .  $f = \Theta(g)$ .
- (ii) Let  $f(x) = 2 \cdot x^2 2$  and  $g(x) = \frac{1}{100} \cdot x 100$ .  $f = \Omega(g)$ . Furthermore, g = o(f).

# Test to determine order

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## The limit test

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Let f and g denote two functions mapping non-negative reals to non-negative reals.

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Let f and g denote two functions mapping non-negative reals to non-negative reals.

Let  $I = \lim_{x \to \infty} \frac{f(x)}{g(x)}$ .

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Let f and g denote two functions mapping non-negative reals to non-negative reals.

Let 
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(i) If *I* is a positive constant,

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Let f and g denote two functions mapping non-negative reals to non-negative reals.

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Let  $f \mbox{ and } g$  denote two functions mapping non-negative reals to non-negative reals.

Let 
$$I = \lim_{x \to \infty} \frac{f(x)}{g(x)}$$
. Then,  
(i) If  $I$  is a positive constant, then  $f = \Theta(g)$ .  
(ii) If  $I = 0$ ,

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Let f and g denote two functions mapping non-negative reals to non-negative reals.

Let 
$$l = \lim_{x \to \infty} \frac{f(x)}{g(x)}$$
. Then,  
(i) If l is a positive constant, then  $f = \Theta(g)$ .  
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If 
$$\lim_{x\to\infty} f(x) = \infty$$
 and if  $\lim_{x\to\infty} g(x) = \infty$ , then,

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If 
$$\lim_{x\to\infty} f(x) = \infty$$
 and if  $\lim_{x\to\infty} g(x) = \infty$ , then,

$$\lim_{x \to \infty} \frac{f(x)}{g(x)} = \lim_{x \to \infty} \frac{f'(x)}{g'(x)}$$

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## Note

If 
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$$\lim_{x \to \infty} \frac{f(x)}{g(x)} = \lim_{x \to \infty} \frac{f'(x)}{g'(x)}$$

The above rule is called L'Hospital's rule.

# Examples

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# Examples

(i) Show that 
$$x = o(x^2)$$
.

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- (i) Show that  $x = o(x^2)$ .
- (ii) Show that  $x = o(x \cdot \log x)$ .

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- (iii) Show that  $\log x = o(x)$ .

# Induction

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## Motivation

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Reaching arbitrary rungs of a ladder.

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## Well-Ordering Principle

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How about all integers?

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How about non-negative reals?

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How about all integers?

How about non-negative reals?

How about non-negative rationals?
# The first principle of Mathematical Induction

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The first principle of Mathematical Induction

## Principle

Assume that the domain is the set of positive integers.

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- **2**  $(\forall k)[P(k) \text{ is true} \rightarrow P(k+1) \text{ is true}]$

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P(n) is **true**, for all positive integers *n*.

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### Observations

(i) Showing that P(1) is true is called the basis step.

## The first principle of Mathematical Induction

## Principle

Assume that the domain is the set of positive integers.

Let P(n) denote a conjecture (argument) that we need to show holds, for every  $n \ge 1$ . If

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2 (\forall k)[P(k) \text{ is true} \rightarrow P(k+1) \text{ is true}]
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### then,

P(n) is true, for all positive integers n.

## Observations

- (i) Showing that P(1) is true is called the basis step.
- (ii) Assuming that P(k) is **true**, in order to show that P(k+1) is **true** is called the inductive hypothesis.

# First Example

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Show that the sum of the first *n* integers is  $\frac{n \cdot (n+1)}{2}$ .

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Show that the sum of the first *n* integers is  $\frac{n \cdot (n+1)}{2}$ .

Formally,  $(\forall n) \left[\sum_{i=1}^{n} i = \frac{n \cdot (n+1)}{2}\right].$ 

# Formal Proof

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# Formal Proof

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Let P(n) denote the predicate  $\sum_{i=1}^{n} i = \frac{n \cdot (n+1)}{2}$ . We are required to prove the conjecture:  $(\forall n)P(n)$ . BASIS (P(1)):

LHS =

# Formal Proof

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LHS = 
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### Proof.

Let P(n) denote the predicate  $\sum_{i=1}^{n} i = \frac{n \cdot (n+1)}{2}$ . We are required to prove the conjecture:  $(\forall n)P(n)$ . BASIS (P(1)):

LHS = 
$$\sum_{i=1}^{1} i$$
  
= 1  
RHS =  $\frac{1 \cdot (1+1)}{2} = \frac{1 \cdot (2)}{2} = \frac{2}{2} = 1$ 

Thus, LHS = RHS and P(1) is true.

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Let us assume that P(k) is true, i.e.,

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Let us assume that P(k) is true, i.e., assume that

$$\sum_{i=1}^{k} i =$$

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## Proof.

Let us assume that P(k) is true, i.e., assume that

$$\sum_{i=1}^k i = \frac{k \cdot (k+1)}{2}$$
## Example

### Proof.

Let us assume that P(k) is true, i.e., assume that

$$\sum_{i=1}^k i = \frac{k \cdot (k+1)}{2}$$

We need to show that P(k+1) is true,

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### Proof.

Let us assume that P(k) is true, i.e., assume that

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We need to show that P(k+1) is true, i.e., we need to show that

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Let us assume that P(k) is true, i.e., assume that

$$\sum_{i=1}^k i = \frac{k \cdot (k+1)}{2}$$

We need to show that P(k+1) is true, i.e., we need to show that

$$\sum_{i=1}^{k+1} i =$$

## Example

### Proof.

Let us assume that P(k) is true, i.e., assume that

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# Proof (contd.)

### Proof.

# Proof (contd.)

Proof.			
Observe that,			
LHS	=		

# Proof (contd.)

## Proof.

$$LHS = \sum_{i=1}^{k+1} i$$

# Proof (contd.)

## Proof.

LHS = 
$$\sum_{i=1}^{k+1} i$$
  
= 1+2+3+...+k+(k+1)

# Proof (contd.)

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LHS = 
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LHS = 
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= (k+1) \cdot (\frac{k}{2}+1)  
= (k+1) \cdot (\frac{k+2}{2})  
=  $\frac{(k+1) \cdot (k+2)}{2}$  = RHS

# Completing the proof

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### Final Steps

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# Completing the proof

#### **Final Steps**

Since, LHS=RHS, we have shown that  $P(k) \rightarrow P(k+1)$ .

## Completing the proof

#### **Final Steps**

Since, LHS = RHS, we have shown that  $P(k) \rightarrow P(k+1)$ .

Applying the first principle of mathematical induction, we conclude that the conjecture is **true**, i.e.,  $(\forall n)P(n)$  holds.





### Main Ideas

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#### Main Ideas

(i) Mathematicize the conjecture.



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- (ii) Prove the basis (usually P(1) and usually easy.)



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```
(iii) Assume P(k).
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- (i) Mathematicize the conjecture.
- (ii) Prove the basis (usually P(1) and usually easy.)
- (iii) Assume P(k).
- (iv) Show P(k+1).



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- (i) Mathematicize the conjecture.
- (ii) Prove the basis (usually P(1) and usually easy.)
- (iii) Assume P(k).
- (iv) Show P(k + 1). (The hard part. Use mathematical manipulation.)
- (v) To show  $P(k) \rightarrow P(k+1)$ , you may use any of the proof techniques discussed, including exhaustive proof, direct proof, contraposition, contradiction, serendipity and induction!

## Another Induction Example

## Another Induction Example

#### Example

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## Another Induction Example

#### Example

Show that the sum of the squares of the first *n* integers is  $\frac{n \cdot (n+1) \cdot (2 \cdot n+1)}{6}$ ,

## Another Induction Example

#### Example

Show that the sum of the squares of the first *n* integers is  $\frac{n \cdot (n+1) \cdot (2 \cdot n+1)}{6}$ , i.e., show that  $\sum_{i=1}^{n} i^2 = \frac{n \cdot (n+1) \cdot (2 \cdot n+1)}{6}$ .

## Proving the Basis

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### Proof.

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### Proof.

BASIS (*P*(1)):
# Proving the Basis

Proof.	
BASIS ( <i>P</i> (1)):	
	LHS =

# Proving the Basis

#### Proof.

LHS = 
$$\sum_{i=1}^{1} i^{i}$$

# Proving the Basis

#### Proof.

$$LHS = \sum_{i=1}^{1} i^2$$
$$= 1$$

# Proving the Basis

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$$LHS = \sum_{i=1}^{1} i^{2}$$
  
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=  $\frac{6}{6}$ 

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=  $\frac{6}{6}$   
= 1

# Proving the Basis

#### Proof.

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LHS = 
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=  $\frac{6}{6}$   
= 1

Thus, LHS = RHS and P(1) is true.

# Induction example (contd.)

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#### Proof.

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# Induction example (contd.)

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We need to show that P(k+1) is true,

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$$\sum_{i=1}^{k} i^{2} = \frac{k \cdot (k+1) \cdot (2 \cdot k + 1)}{6}$$

LHS = 
$$\sum_{i=1}^{k+1} i^2$$
  
=  $1^2 + 2^2 + 3^2 + \ldots + k^2 + (k+1)^2$ 

#### Induction example (contd.)

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$$HS = \sum_{i=1}^{k+1} i^2$$
  
=  $1^2 + 2^2 + 3^2 + \dots + k^2 + (k+1)^2$   
=  $(1^2 + 2^2 + 3^2 + \dots + k^2) + (k+1)^2$ 

#### Induction example (contd.)

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Let us assume that P(k) is true, i.e., assume that

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# Induction proof (contd.)

# Induction proof (contd.)



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# Induction proof (contd.)

Proof.

# $= \frac{k \cdot (k+1) \cdot (2 \cdot k+1)}{6} + (k+1)^2, \text{ using the inductive hypothesis}$

# Induction proof (contd.)

$$= \frac{k \cdot (k+1) \cdot (2 \cdot k+1)}{6} + (k+1)^2, \text{ using the inductive hypothesis}$$
$$= \frac{k+1}{6} (k \cdot (2 \cdot k+1) + 6 \cdot (k+1))$$

# Induction proof (contd.)

$$= \frac{k \cdot (k+1) \cdot (2 \cdot k+1)}{6} + (k+1)^2, \text{ using the inductive hypothesis} = \frac{k+1}{6} (k \cdot (2 \cdot k+1) + 6 \cdot (k+1)) = \frac{k+1}{6} (2 \cdot k^2 + k + 6 \cdot k + 6)$$

# Induction proof (contd.)

$$= \frac{k \cdot (k+1) \cdot (2 \cdot k+1)}{6} + (k+1)^2, \text{ using the inductive hypothesis}$$

$$= \frac{k+1}{6} (k \cdot (2 \cdot k+1) + 6 \cdot (k+1))$$

$$= \frac{k+1}{6} (2 \cdot k^2 + k + 6 \cdot k + 6)$$

$$= \frac{k+1}{6} (2 \cdot k^2 + 7 \cdot k + 6)$$

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# Induction Proof (contd.)

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#### Proof.

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# Induction Proof (contd.)



# Induction Proof (contd.)

# Proof. $= \frac{(k+1) \cdot (k+2) \cdot (2 \cdot (k+1)+1)}{6}$ = RHS.
## Induction Proof (contd.)

#### Proof.

$$= \frac{(k+1) \cdot (k+2) \cdot (2 \cdot (k+1) + 1)}{6}$$
  
= RHS.

Since, LHS = RHS, we have shown that  $P(k) \rightarrow P(k+1)$ .

## Induction Proof (contd.)

# Proof. $= \frac{(k+1) \cdot (k+2) \cdot (2 \cdot (k+1) + 1)}{6}$ = RHS.Since, LHS=RHS, we have shown that $P(k) \rightarrow P(k+1)$ .

Applying the first principle of mathematical induction, we conclude that the conjecture is true.  $\hfill \Box$ 

# Induction Example

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## Example

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## Induction Example

#### Example

Show that the sum of the first n odd integers is  $n^2$ ,

## Induction Example

#### Example

Show that the sum of the first *n* odd integers is  $n^2$ , i.e., show that  $\sum_{i=1}^{n} (2 \cdot i - 1) = n^2$ .

## Proving the conjecture

## Proving the conjecture

#### Proof.

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Proof.
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Proof.		
BASIS ( <i>P</i> (1)):		
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=  $2 \cdot 1 - 1$   
= 1

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=  $1$   
RHS =  $1^{2}$ 

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RHS =  $1^{2}$   
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Thus, LHS = RHS and P(1) is true.





## Proof.

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Let us assume that P(k) is true, i.e., assume that



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Proof.		

Proof.		

Proof.	
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# Completing the proof

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# Completing the proof

LHS = 
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=  $(1 + 3 + 5 + \dots (2 \cdot k - 1)) + (2 \cdot (k + 1))$ 

# Completing the proof

$$LHS = \sum_{i=1}^{k+1} (2 \cdot i - 1)$$
  
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= (1+3+5+...(2 \cdot k - 1)) + (2 \cdot k + 1)  
= k<sup>2</sup> + (2 \cdot k + 1), using the inductive hypothesis

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Since LHS = RHS, we have shown that  $P(k) \rightarrow P(k+1)$ .

# Completing the proof

### Proof.

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= (1+3+5+...(2 \cdot k - 1)) + (2 \cdot k + 1)  
= k^2 + (2 \cdot k + 1), using the inductive hypothesis  
= (k + 1)^2  
= RHS

Since LHS = RHS, we have shown that  $P(k) \rightarrow P(k+1)$ . Applying the first principle of mathematical induction, we conclude that the conjecture is true.

# One Final Example

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## Example

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## Example

Show that  $7^n - 5^n$  is always an even number for  $n \ge 0$ ,

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Show that  $7^n - 5^n$  is always an even number for  $n \ge 0$ , i.e., show that  $2 \mid (7^n - 5^n)$ ,  $\forall n \ge 0$ .

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BASIS (P(0)):

LHS =

# One Final Example

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$$LHS = 7^0 - 5^0$$

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Show that  $7^n - 5^n$  is always an even number for  $n \ge 0$ , i.e., show that  $2 \mid (7^n - 5^n)$ ,  $\forall n \ge 0$ .

### Proof.

$$LHS = 7^0 - 5^0$$
  
= 1 - 1

# One Final Example

## Example

Show that  $7^n - 5^n$  is always an even number for  $n \ge 0$ , i.e., show that  $2 \mid (7^n - 5^n)$ ,  $\forall n \ge 0$ .

### Proof.

$$LHS = 7^0 - 5^0$$
$$= 1 - 1$$
$$= 0$$

# One Final Example

## Example

Show that  $7^n - 5^n$  is always an even number for  $n \ge 0$ , i.e., show that  $2 \mid (7^n - 5^n)$ ,  $\forall n \ge 0$ .

### Proof.

$$LHS = 7^0 - 5^0$$
$$= 1 - 1$$
$$= 0$$

# One Final Example

## Example

Show that  $7^n - 5^n$  is always an even number for  $n \ge 0$ , i.e., show that  $2 \mid (7^n - 5^n)$ ,  $\forall n \ge 0$ .

### Proof.

BASIS (P(0)):

$$LHS = 7^0 - 5^0$$
  
= 1 - 1  
= 0

Since the LHS is even, we have proven the basis P(0).



# Proof (contd.)



# Proof (contd.)

## Proof.

Let us assume that P(k) is true, i.e.,

# Proof (contd.)

### Proof.

Let us assume that P(k) is true, i.e., assume that  $(7^k - 5^k)$  is divisible by 2 for some k.

# Proof (contd.)

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Let us assume that P(k) is true, i.e., assume that  $(7^k - 5^k)$  is divisible by 2 for some k. It follows that  $(7^k - 5^k) = 2 \cdot m$ , for some integer m.

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# Proof (contd.)

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$$7^{k+1} - 5^{k+1} =$$

# Proof (contd.)

### Proof.

$$7^{k+1} - 5^{k+1} = 7 \cdot 7^k - 5 \cdot 5^k$$

# Proof (contd.)

### Proof.

$$\begin{array}{lll} 7^{k+1}-5^{k+1} & = & 7\cdot7^k-5\cdot5^k \\ & = & 7\cdot(2\cdot m+5^k)-5\cdot5^k, \text{ using the inductive hypothesis} \end{array}$$

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$$\begin{array}{lll} 7^{k+1} - 5^{k+1} & = & 7 \cdot 7^k - 5 \cdot 5^k \\ & = & 7 \cdot (2 \cdot m + 5^k) - 5 \cdot 5^k, \text{ using the inductive hypothesis} \\ & = & 14 \cdot m + 7 \cdot 5^k - 5 \cdot 5^k \\ & = & 14 \cdot m + 5^k \cdot (7 - 5) \\ & = & 14 \cdot m + 2 \cdot 5^k \end{array}$$

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=  $2 \cdot (7 \cdot m + 5^k)$  = some even number!

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Let us assume that P(k) is true, i.e., assume that  $(7^k - 5^k)$  is divisible by 2 for some k. It follows that  $(7^k - 5^k) = 2 \cdot m$ , for some integer m. We need to show that P(k + 1) is true, i.e.,  $(7^{k+1} - 5^{k+1})$  is divisible by 2. Observe that,

We have thus shown that  $P(k) \rightarrow P(k+1)$ .

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We have thus shown that  $P(k) \rightarrow P(k+1)$ . Applying the first principle of mathematical induction, we conclude that the conjecture is true.

# Second Principle of Induction

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## Principle

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# Second Principle of Induction

## Principle

Assume that the domain is the set of positive integers.
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- (ii)  $(\forall r)(1 \le r \le k)[P(r) \text{ is true}] \rightarrow P(k+1) \text{ is true}]$

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#### Note

Also called Strong Induction. Is necessary, when the first principle does not help us.

# Example of Strong Induction

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### Example

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# Example of Strong Induction

#### Example

Show that every number greater than or equal to 8 can be expressed in the form  $5 \cdot a + 3 \cdot b$ , for suitably chosen *a* and *b*.



## Proving the conjecture

### Proof.

(i) The conjecture is clearly true for 8, 9 and 10.

## Proving the conjecture

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- (v) Observe that (k + 1) 3 = (k 2) is at least 8 and less than k.
- (vi) As per the inductive hypothesis, (k-2) can be expressed in the form  $3 \cdot a + 5 \cdot b$ , for suitably chosen a and b.

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- (vii) It follows that  $(k+1) = (k-2) + 3 = 3 \cdot a + 5 \cdot b + 3 = 3 \cdot (a+1) + 5 \cdot b$  can also be so expressed.
- (viii) Applying the second principle of mathematical induction, we conclude that the conjecture is true, for all  $n \ge 8$ .

# Another Example

# Another Example

### Example

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## Another Example

### Example

Show that every element in the set  $S = \{2, 3, \dots, \}$  is either a prime number or a product of primes.





# Proving the conjecture

#### Proof.

• For the basis, observe that 2 is a prime.

## Proving the conjecture

- For the basis, observe that 2 is a prime.
- 2 Assume that the conjecture holds for all  $r, 2 \le r \le k$ .

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In other words, assume that every number in the set  $S_k = \{2, 3, ..., k\}$  is either a prime or can be expressed as a product of primes.

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In other words, assume that every number in the set  $S_k = \{2, 3, ..., k\}$  is either a prime or can be expressed as a product of primes.

**(a)** Now consider the number (k + 1). If (k + 1) is a prime, then
# Proving the conjecture

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- If (k+1) is composite, then  $(k+1) = a \cdot b$ , where a, b < (k+1).

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- As per the inductive hypothesis, both a and b are either primes themselves or can be expressed as products of primes.

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- If (k+1) is composite, then  $(k+1) = a \cdot b$ , where a, b < (k+1).
- As per the inductive hypothesis, both a and b are either primes themselves or can be expressed as products of primes.
- **(**) In either case, it follows that (k + 1) can be expressed as a product of primes.
- Applying the second principle of mathematical induction, we conclude that the conjecture is true for all elements in the domain.

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# Motivating Examples

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### Example

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# Motivating Examples

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How many 4 digit numbers can you create using the digits 1, 2, 3, and 4, assuming no digit repeats?

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# Permutations

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## Definition

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## Permutations

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#### Definition

$$n! = \begin{cases} 1, & \text{if } n = 0\\ n \cdot (n-1)!, & \text{otherwise} \end{cases}$$

Computing the number of permutations

Computing the number of permutations

Computing P(n, r)

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Computing the number of permutations

Computing P(n, r)

Computing the number of permutations

### Computing P(n, r)

Using the multiplication principle,

P(n,r) =

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Computing the number of permutations

### Computing P(n, r)

$$P(n,r) = r$$

Computing the number of permutations

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Using the multiplication principle,

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$$P(n,r) = n \cdot (n-1) \cdot \dots (n-r+1) \\ = n \cdot (n-1) \cdot \dots (n-r+1) \cdot \frac{(n-r) \cdot (n-r-1) \cdot \dots 1}{(n-r) \cdot (n-r-1) \cdot \dots 1}$$

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# Permutations (contd.)

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### Example

Subramani Mathematical Review

## Permutations (contd.)

### Example

Compute P(7,3), P(n,0), P(n,1), and P(n,n).

## Permutations (contd.)

### Example

Compute P(7,3), P(n,0), P(n,1), and P(n,n).

### Solution:

## Permutations (contd.)

### Example

Compute P(7,3), P(n,0), P(n,1), and P(n,n). Solution: 210,

## Permutations (contd.)

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Compute P(7,3), P(n,0), P(n,1), and P(n,n). Solution: 210, 1,

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Compute P(7,3), P(n,0), P(n,1), and P(n,n).

**Solution:** 210, 1, *n*, and *n*!.

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How many 3 letter words can be formed using the letters in the word "compiler"?

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In how many ways can a president and vice-president be chosen from a group of 20 people?

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**Solution:** 210, 1, *n*, and *n*!.

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Solution: P(20, 2).

## One more example

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### Example

Subramani Mathematical Review

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Now consider the case in which the books of a given subject are required to be together.

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First arrange the three subjects.

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First arrange the three subjects. This can be done in P(3,3) = 3! ways.

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and the complexity books can be permuted in P(3,3) = 3! ways.

Using the multiplication principle, the total number of arrangements is  $3! \cdot 4! \cdot 7! \cdot 3!$ .

# More Examples

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### Example

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# More Examples

### Example

Solve the motivating examples.

Subramani Mathematical Review

# Motivating Examples

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# Motivating Examples

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- **②** There cannot be more than two men on the committee.

# Combinations

### Definition

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Solution: 35,

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## Combinations (examples)

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- it can contain at most one freshman. Solution:  $C(34, 8) + C(19, 1) \cdot C(34, 7)$ .
- it contains at least one freshman. Solution: C(53, 8) C(34, 8).

## More examples

## More examples

#### Example

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## More examples

#### Example

Solve the motivating examples.

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## Exercises

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#### Identities

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- In how many ways, can you seat 11 men and 8 women in a row, so that no two women sit together?
- A committee of three has to be chosen from five Democrats, three Republicans and four independents.

In how many ways can the committee be chosen, if it cannot include both Democrats and Republicans?

# Motivation
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#### Expansions

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We want a general formula that permits us to write down the terms of  $(a + b)^n$  without actual multiplication.

### Pascal's Triangle



### The coefficient table

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The coefficient table

Consider the following table:

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Row 0:

C(0, 0)

### Pascal's Triangle

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Row 0:	C(0, 0	C(0, 0)					
Row 1:	C(1, 0)	C(1, 1)					

### Pascal's Triangle

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Row 1:	С(	1, 0)	C(1, 1)	
Row 2:	C(2, 0)	C(2, 1)		C(2, 2)

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Row 0:				C(0, 0)			
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Row 1:		C(1, 0)		C(1, 1)			
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Row 3: C(3, 0)		C(3, 1)		C(3, 2)		C(3, 3)	
:							
Row n: C(n, 0)	C(n, 1)				C(n, n - 1)		C(n, n)

# Pascal's triangle (contd.)

#### The Value Table

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## Pascal's triangle (contd.)

The Value Table

Writing down the values of the terms gives the following table:

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:												
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## Pascal's formula

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#### Theorem

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 $C(n,k) = C(n-1,k-1) + C(n-1,k), 1 \le k \le n-1.$ 

### Proving Pascal's formula

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#### Proof.

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- (viii) Using the addition principle,  $C(n,k) = T_1 + T_2 = C(n-1,k-1) + C(n-1,k)$ .

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Recall the combinatorial proof for proving that C(n, r) = C(n, n - r).



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### Exercise

Find bounds on  $H_n$ , the  $n^{th}$  Harmonic number.