

Lecture #3

Recursion (1)

Algorithm

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In This Lecture

□ Concept of recursion

- What is recursion? Why do we need recursion?

□ How to design and analyze recursion

- Design by divide and conquer
- Correctness analysis by mathematical induction

□ Recursive complexity

- How is $T(n)$ of a recursive algorithm represented?

Outline

□ Recursion

□ How to Design Recursion

□ Correctness Analysis of Recursion

□ Recursive Complexity

Recursion

- We say “**Something is recursive**” when it is defined in terms of itself



Recursion In Math & CS

□ Recurrence relation in Mathematics

- Equation that is recursively defined by itself

□ Recursive function in CS

- Function that is recursively defined by itself

$$a_n = \begin{cases} n \times a_{n-1}, & n > 1 \\ 1, & n = 1 \end{cases}$$

Recurrence relation

```
def f(n):  
    if n == 1:  
        return 1  
    else:  
        return n * f(n - 1)
```

Recursive function

- They are the same intrinsically under the concept of recursion

Why Do We Need Recursion?

- Numerous operations are compactly represented in just a few words (i.e., it improves readability!)

$$n! = \begin{cases} 1 & n = 0 \text{ or } 1 \\ n \times (n - 1)! & n > 1 \end{cases}$$

⚠ Do not misunderstand!

- **Not saying** recursion is always proper for every problem
 - It's effective when your problem has a recursive property
- **Not saying** a recursive function is always computationally efficient and optimized

Definition of Recursive Function

□ A function is recursive when it is defined by

- 1) Simple base case(s)

- Terminating scenario that doesn't use recursion to produce an answer
 - If there is no base case, the function will run forever, incurring a stack overflow error

- 2) Recursive step

- Rules that reduces all other cases towards the base by calling itself

```
def function(n):  
    if n == 1: # base case (example)  
        do something  
    else:      # recursive step  
        do something with function(k)  
        where k is reduced toward the base case  
        (e.g., k = n-1, n/2, etc.)
```

Example: Factorial

❑ Factorial: $n! = 1 \times 2 \times \cdots \times (n - 1) \times n$

❑ Recurrence relation of $n!$

$$n! = \begin{cases} 1 & n = 0 \text{ or } 1 \\ n \times (n - 1)! & n > 1 \end{cases}$$

Base case
Recursion step

❑ Recursive function of $n!$

```
def factorial(n):  
    if n == 0 or n == 1:  
        return 1  
    else:  
        return n * factorial(n-1)
```

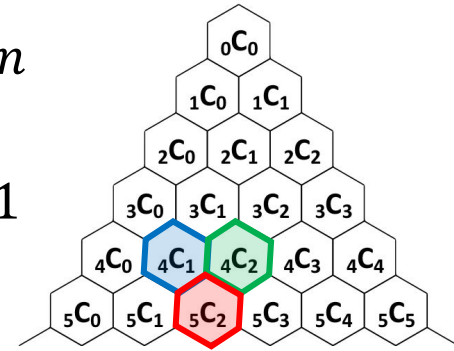

Example: Binomial Coefficient

□ Binomial coefficient of n & k (where $0 \leq k \leq n$)

$${}_nC_k = \binom{n}{k} = \frac{n!}{k! (n-k)!}$$

□ Recurrence relation

$$\binom{n}{k} = \begin{cases} 1 & k = 0 \text{ or } k = n \\ \binom{n-1}{k-1} + \binom{n-1}{k} & 1 \leq k \leq n-1 \end{cases}$$



□ Recursive function

```
def bin-coeff(n, k):  
    if k == 0 or k == n: return 1  
    else: return bin-coeff(n-1, k-1) + bin-coeff(n-1, k)
```

Outline

- ❑ Recursion

- ❑ How to Design Recursion

- ❑ Correctness Analysis of Recursion

- ❑ Recursive Complexity

How To Design Recursion? (1)

□ Problem: Exponentiation (or power)

- Input: base number a and exponent n
- Output: a^n

□ Let's design the problem in a recursive way!

- One strategy is **Divide & Conquer**
 - **Divide** the problem into several (smaller) sub-problems
 - **Conquer** them separately
 - **Aggregate** the answers of the sub-problems if necessary

$$a^n = \underbrace{a \times a \times \cdots \times a}_{n-1} \times \underbrace{a}_1 = a^{n-1} \times a^1$$

How To Design Recursion? (2)

□ Let's define the recurrence relation for the problem

$$\begin{array}{c} \text{power}(a, n) \qquad \qquad \qquad a^1 = \text{power}(a, 1) \\ \underbrace{\qquad\qquad\qquad} \qquad \qquad \qquad \underbrace{\qquad\qquad\qquad} \\ a^n = \underbrace{a \times a \times \cdots \times a}_{a^{n-1} = \text{power}(a, n-1)} \times a \end{array}$$

- **Assume** that a function called $\text{power}(a, n)$ computes a^n
 - **Base case**: return 1 if $n = 0$
 - **Recursive step**: return $\text{power}(a, n - 1) \times a$ if $n > 0$

```
def power(a, n):  
    if n == 0: return 1  
    else:      return power(a, n-1) * a
```

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How To Prove Its Correctness?

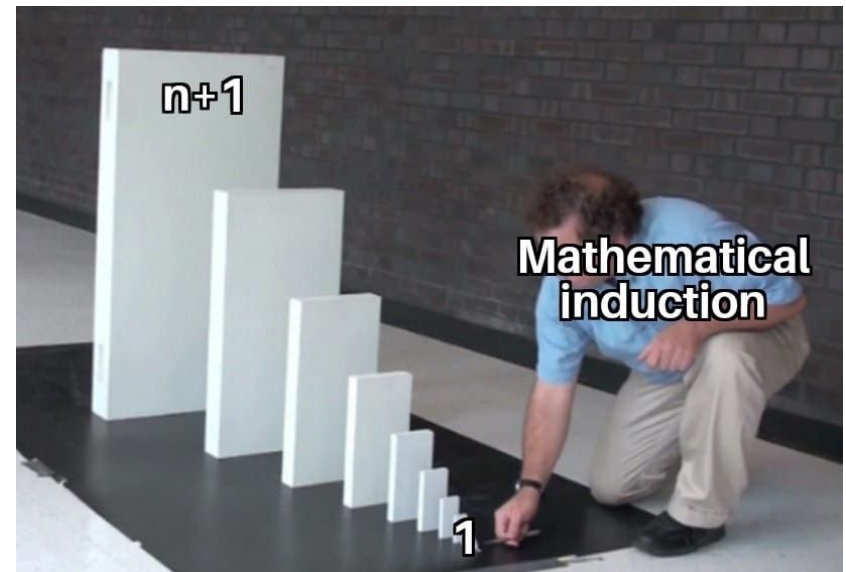
□ Prove the correctness using **Mathematical Induction**

□ Main idea of mathematical induction

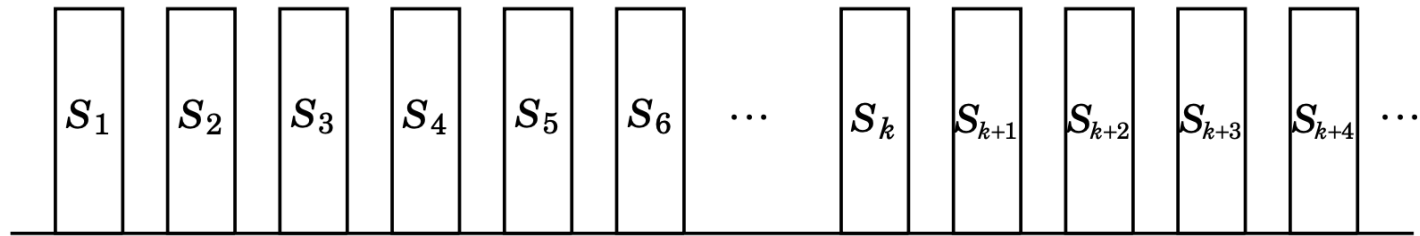
- If the previous domino falls, show the next domino also falls!
- All remaining dominoes are expected to be fallen when the above is revealed!

⚠ **Do not misunderstand!**

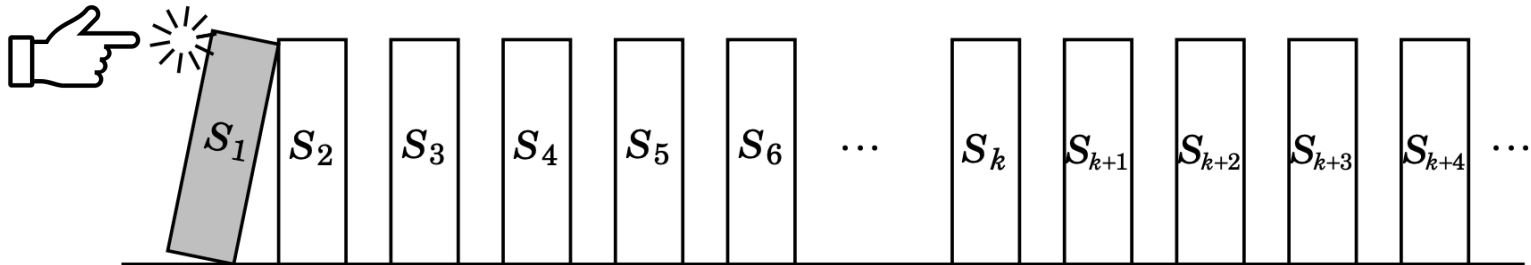
- Mathematical induction is **not inductive reasoning**.
 - It is rather deductive reasoning!



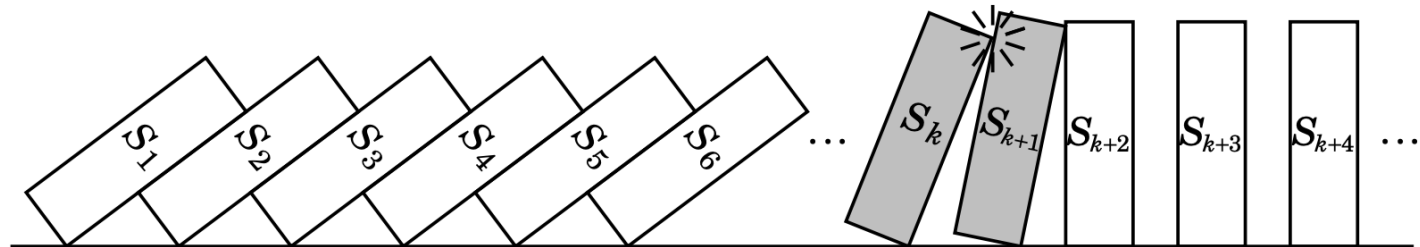
The Simple Idea Behind Mathematical Induction



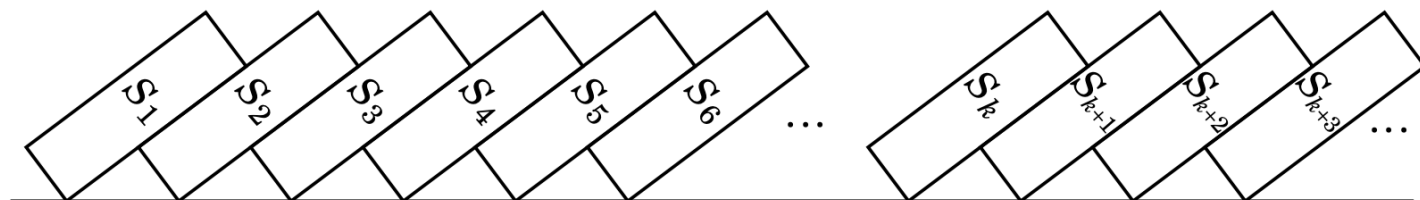
Push! Statements are lined up like dominoes.



(1) Suppose the first statement falls (is proved true);



(2) Suppose the k th falling always causes the $(k + 1)$ th to fall;



Then all must fall (all are proved true).

The last one eventually falls!!

Mathematical Induction

□ Claim. S_n holds for every natural number n .

□ 1) Base case(s)

- Prove that S_n holds when n is base case(s).

□ 2) Inductive step

- Previous case: Assume that S_k is true for $n = k$.
- Next case: Does S_{k+1} also hold for $n = k + 1$?
 - Prove it also holds for $k + 1$ based on the assumption at k .
 - The increment does not need to be 1.
 - Any increment such as +2 and $\times 2$ is possible (it depends on problems).

□ Then, say “ S_n holds for every n by induction!”

Example: Power (1)

□ S_n : $\text{power}(a, n)$ correctly computes a^n for all $n \in \mathbb{N}_0$.

```
def power(a, n):  
    if n == 0:    return 1  
    else:         return power(a, n-1) * a
```

□ Proof by induction

- 1) Base case

- The base case is $n = 0$, and in this case, $\text{power}(a, n)$ always returns 1.
- That is, $a^0 = 1$ is obviously true for any a .
- Thus, S_0 is true for the base case!

Example: Power (2)

□ S_n : `power(a, n)` correctly computes a^n for all $n \in \mathbb{N}$.

```
def power(a, n):  
    if n == 0:    return 1  
    else:        return power(a, n-1) * a
```

□ Proof by induction

- 2) Inductive step

- Previous case: assume the claim holds for k (i.e., assume S_k is true).
- Next case: does S_{k+1} also hold?

$$\text{power}(a, k+1) = \overset{\substack{\text{true by } S_k \\ \downarrow}}{\text{power}(a, k)} \times a = a^k \times a = a^{k+1}$$

- Thus, S_n holds for every n by induction!

Example: Inequality

□ $S_n : 2^n > n + 4$ for $n \geq 3$.

- **Base case:** $n = 3$ and $2^3 = 8 > 3 + 4 = 7$; thus, it holds.

- **Inductive step**

- Previous case: assume S_k holds for $n = k$ (where $k > 3$).

$$S_k : 2^k > k + 4 \text{ is true (assumed)}$$

- Next case: Does S_{k+1} also hold?

$$S_{k+1} : 2^{k+1} > k + 5 \Leftrightarrow 2 \times 2^k - k - 5 > 0 \quad \Leftarrow \text{Is it really true?}$$

$$2 \times \boxed{2^k} - k - 5 > 2 \times \boxed{(k + 4)} - k - 5 = k + 3 > 0 \quad (\because k > 3)$$

true by S_k \swarrow

- Thus, S_n holds for $n \geq 3$ by induction!

Strong Induction

□ We use strong induction when

- It is hard to specify a previous case (e.g., not k), or
- Multiple previous cases are needed for proving.

□ Claim. S_n holds for every $n \geq a$.

$b = \#$ of base cases

□ 1) **Base case:** Prove $S_a, S_{a+1}, \dots, S_{a+b}$ holds.

□ 2) **Inductive step**

- Previous case: Assume that all S_i hold for $a \leq i \leq k$
- Next case: Does S_{k+1} also hold?
 - Prove the claim using **one or more assumptions** S_i in the range (that's why it's called **strong**)

Example: Strong Induction

□ $S_n : a_n = a_{n-2} + 2a_{n-1}$ is odd for every $n \geq 1$ given $a_1 = 1$ and $a_2 = 3$.

□ 1) Base cases

- S_1 and S_2 are trivially true because a_1 and a_2 are odd.

□ 2) Inductive step

- Previous: Assume S_i holds, i.e., a_i are odd for $1 \leq i \leq k$
- Next: is S_{k+1} also true? \Rightarrow is $a_{k+1} = a_{k-1} + 2a_k$ odd?
 - Use two assumptions on S_{k-1} and S_k .
 - The sum of odd and even leads to odd, implying a_{k+1} is odd too!

□ By induction, a_n is odd for $n \geq 1$ [Q.E.D]

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Complexity of Recursive Function

□ What is the time complexity of the below algorithm?

```
def power(a, n):  
    if n == 0:  
        return 1  
    else:  
        return power(a, n-1) * a
```

- $T(n)$ of the recursive algorithm is also described **recursively**.
- We need to solve the recurrence relation of $T(n)$ to obtain its closed solution.

Example: Power (1)

□ Let $T(n)$ denote the time complexity of `power(a, n)`

- **Base case:** $T(n) = C$ because it just returns $1 \in O(1)$ time
- **Recursive step:** $T(n) = T(n - 1) + C$
 - 1) It recursively calls `power(a, n-1)` which requires $T(n - 1)$ time
 - 2) After then, the result is multiplied by a requiring $O(1)$ time

```
def power(a, n):  
    if n == 0:  
        return 1  
    else:  
        return power(a, n-1) * a
```



$$T(n) = \begin{cases} C & n = 0 \\ T(n - 1) + C & n > 0 \end{cases}$$

Q. What is its closed solution?

Example: Power (2)

□ Let's divide the problem as follows:

If n is even: $a^n = (a \times a) \times (a \times a) \times \cdots \times (a \times a) = (a^2)^{\frac{n}{2}}$

If n is odd: $a^n = a \times (a \times a) \times (a \times a) \times \cdots \times (a \times a) = a \times (a^2)^{\frac{n-1}{2}}$

```
def power(a, n):  
    if n == 0:  
        return 1  
    else if n is even:  
        return power(a * a, n/2)  
    else:  
        return a * power(a * a, (n-1)/2)
```



$$T(n) = \begin{cases} C & n = 0 \\ T(n/2) + C & n \text{ is even} \\ T\left(\frac{n-1}{2}\right) + C & n \text{ is odd} \end{cases}$$

Q. What is its closed solution?

What You Need To Know

□ Concept of recursion

- **Something is recursive** when it is defined in terms of itself

□ How to design and analyze recursion

- Divide & conquer: **Divide** the problem, **conquer** its sub-problems, and **aggregate** their answers
- **Mathematical induction**: if the previous domino falls, show the next domino also falls!

□ Recursive complexity

- The time complexiy $T(n)$ of a recursive algorithm is also recursive. \Rightarrow How to solve it?

In Next Lecture

□ How to obtain the closed solution of a recursive time complexity?

- Method 1) Subsitute method
- Method 2) Mathematical induction
- Method 3) Master theorem



Thank You