#### **CS204: Discrete Mathematics**

# Ch 5. Induction and Recursion

# Recursive Algorithms and Recurrence Relations

### **Sungwon Kang**

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### Ch 5. Induction and Recursion

- 5.1 Mathematical Induction
- 5.2 Strong Induction and Well-Ordering
- 5.3 Recursive Definitions and Structural Induction
- 5.4 Recursive Algorithms -
- 5.5 Program Correctness

### **Recursive Algorithms and Recurrence Relations**

- 1. Recursive Algorithms
- 2. Recurrence Relations
- 3. Closed Form Solutions

# 1. Recursive Algorithms

**Definition**: An algorithm is said to be *recursive* if it solves a problem by reducing it to an instance of the same problem with smaller input.

For a recursive algorithm to terminate, the instance of the problem it solves must eventually be reduced to some initial case for which the solution is known.

# **Factorial Algorithm**

**Example**: Give a recursive algorithm for computing n!, where n is a nonnegative integer.

**Example**: Give a recursive algorithm for computing *n*!, where *n* is a nonnegative integer.

**Solution**: Use the recursive definition of the factorial function.

```
procedure factorial(n: nonnegative integer)
if n = 0 then return 1
    else return n·factorial(n - 1)
{output is n!}
```

# **Exponentiation Algorithm**

**Example**: Give a recursive algorithm for computing  $a^n$ , where a is a nonzero real number and n is a nonnegative integer.

**Example**: Give a recursive algorithm for computing  $a^n$ , where a is a nonzero real number and n is a nonnegative integer.

**Solution**: Use the recursive definition of  $a^n$ .

```
procedure power(a: nonzero real number, n:
    nonnegative integer)

if n = 0 then return 1
    else return a power (a, n - 1)
{output is an}
```

# **GCD** Algorithm

**Example**: Give a recursive algorithm for computing the greatest common divisor of two nonnegative integers a and b with a < b.

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**Solution**: Use the reduction

$$gcd(a,b) = gcd(b \mod a, a)$$

and the condition gcd(0,b) = b when b > 0.

**Example**: Give a recursive algorithm for computing the greatest common divisor of two nonnegative integers *a* and *b* with *a* < *b*.

**Solution**: Use the reduction

$$gcd(a,b) = gcd(b \mod a, a)$$

and the condition gcd(0,b) = b when b > 0.

```
gcd(18,84) = gcd(84 \mod 18, 18)
= gcd(12, 18)
= gcd(18 \mod 12, 12)
= gcd(6, 12)
= gcd(12 \mod 6, 6)
= gcd(0,6)
= 6
```

**Example**: Give a recursive algorithm for computing the greatest common divisor of two nonnegative integers a and b with a < b.

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### 2. Recurrence Relations

**Definition:** A recurrence relation for the sequence  $\{a_n\}$  is an equation that expresses  $a_n$  in terms of one or more of the previous terms, that is,  $a_0$ ,  $a_1$ , ...,  $a_{n-1}$ .

A sequence is called a *solution* of a recurrence relation if its terms satisfy the recurrence relation.

$$P(n) = \begin{cases} 1 & \text{if } n = 0 \\ n + P(n-1) & \text{if } n > 0 \end{cases}$$

- A kind of recursive definitions
- Different from recursive algorithms
- Often complexity of recursive algorithms can be expressed as recurrence relations

### Bottom-up evaluation

Find P(5) using the recurrence relation (bottom-up).

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$$P(0) = 1.$$

$$P(1) = 1 + P(0) = 1 + 1 = 2.$$

$$P(2) = 2 + P(1) = 2 + 2 = 4.$$

$$P(3) = 3 + P(2) = 3 + 4 = 7.$$

$$P(4) = 4 + P(3) = 4 + 7 = 11.$$

$$P(5) = 5 + P(4) = 5 + 11 = 16.$$

### Top-down evaluation

Find P(5) using the recurrence relation (top-down).

$$P(n) = \begin{cases} 1 & \text{if } n = 0 \\ n + P(n-1) & \text{if } n > 0 \end{cases}$$

### Top-down evaluation

Find P(5) using the recurrence relation (top-down).

$$P(n) = \begin{cases} 1 & \text{if } n = 0 \\ n + P(n-1) & \text{if } n > 0 \end{cases}$$

$$P(5) = 5 + P(4)$$

$$= 5 + (4 + P(3))$$

$$= 5 + (4 + (3 + P(2)))$$

$$= 5 + (4 + (3 + (2 + P(1))))$$

$$= 5 + (4 + (3 + (2 + (1 + P(0)))))$$

$$= 5 + (4 + (3 + (2 + (1 + 1))))$$

$$= 5 + (4 + (3 + (2 + 2)))$$

$$= 5 + (4 + (3 + 4))$$

$$= 5 + (4 + 7)$$

$$= 5 + 11 = 16$$

### Recurrence relation for the size of a power set

Let X be a finite set with n elements. Find a recurrence relation C(n) for the number of elements in the power set  $\mathcal{P}(X)$ .

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Let X be a finite set with n elements. Find a recurrence relation C(n) for the number of elements in the power set  $\mathcal{P}(X)$ .

#### Solution

$$C(n) = \begin{cases} 1 & \text{if } n = 0 \\ 2 \cdot C(n-1) & \text{if } n > 0 \end{cases}$$

(Why?)

In the early thirteenth century, the Italian mathematician Leonardo Pisano Fibonacci asked:

A certain man put a pair of rabbits in a place surrounded on all sides by a wall. How many pairs of rabbits can be produced from that pair in a year if it is supposed that every month each pair begets a new pair which from the second month on becomes productive?

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Number of						
of Rabbits	$\bigcirc$					
Month	1	2	3	4	5	

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Number	$\circ$	0	00		

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A certain man put a pair of rabbits in a place surrounded on all sides by a wall. How many pairs of rabbits can be produced from that pair in a year if it is supposed that every month each pair begets a new pair which from the second month on becomes productive?

Recurrence relation answer:

$$F(n) = \begin{cases} 1 & \text{if } n = 1 \text{ or } n = 2 \\ F(n-1) + F(n-2) & \text{if } n > 2 \end{cases}$$

The sequence  $F(1), F(2), F(3), \ldots$  is called the *Fibonacci* sequence.

### 3. Closed Form Solutions

Complexity of recursive algorithms expressed with recurrence relations can be directly known by finding their closed-form solutions

The recurrence relation

$$P(n) = \begin{cases} 1 & \text{if } n = 1 \\ n + P(n-1) & \text{if } n > 1 \end{cases}$$

and the formula

$$P(n) = \frac{n(n+1)}{2}$$

give exactly the same answers for all values of *n*. (Mathematical Induction can be used to prove this ) The latter, non-recursive, formula is called a *closed-form solution* to the recurrence relation.

### Guessing a closed-form solution: finding patterns

**Evaluating** 

$$M(n) = \begin{cases} 500 & \text{if } n = 0 \\ 1.10 \cdot M(n-1) & \text{if } n > 0 \end{cases}$$

gives

$$M(0) = 500$$
 =  $500 (1.10)^{0}$   
 $M(1) = 500 \cdot 1.10$  =  $500(1.10)^{1}$   
 $M(2) = 500 \cdot 1.10 \cdot 1.10$  =  $500(1.10)^{2}$   
 $M(3) = 500 \cdot 1.10 \cdot 1.10 \cdot 1.10$  =  $500(1.10)^{3}$   
 $M(4) = 500 \cdot 1.10 \cdot 1.10 \cdot 1.10 \cdot 1.10$  =  $500(1.10)^{4}$ 

The pattern suggests that  $\underline{M(n)} = 500(1.10)^n$  is a closed-form solution to the recurrence relation.

$$H(n) = \begin{cases} 1 & \text{if } n = 1 \\ H(n-1) + 6n - 6 & \text{if } n > 1 \end{cases}$$

#### How to find a closed-form solution?

#### Approach 1)

 $= 3k^2 - 3k + 1$ 

```
\begin{aligned} &H(k) = H(k-1) + 6k - 6 \\ &= \{H(k-2) + 6(k-1) - 6\} + 6k - 6 \\ &= \{[H(k-3) + 6(k-2) - 6] + 6(k-1) - 6\} + 6k - 6 \\ &= H(k-3) + 6x\{(k-2) + (k-1) + k\} - 3x6 \\ & \dots \\ &= H(1) + 6x\{2 + \dots + (k-2) + (k-1) + k\} - (k-1)x6 \quad -- \text{ pattern has been found} \quad - (A) \\ &= 1 + 6x(k-1)(k+2)/2 - (k-1)x6 & -- \text{ pattern has been simplified - (B)} \\ &= 1 + 3x(k-1)(k+2) - (k-1)x6 \\ &= 1 + 3x(k-1)k \end{aligned}
```

$$H(n) = \begin{cases} 1 & \text{if } n = 1 \\ H(n-1) + 6n - 6 & \text{if } n > 1 \end{cases}$$

#### How to find a closed-form solution?

#### Approach 2)

$$\Sigma_{n=1}^{k} (H(n) - H(n-1)) = \Sigma_{n=1}^{k} (6n - 6) -- (C)$$

H(k) - H(1) = 6 
$$(\Sigma_{n=1}^{k} n - \Sigma_{n=1}^{k} 1)$$
  
= 6  $(k(k+1)/2 - k)$   
= 6  $(k(k-1)/2)$   
=  $3k^2 - 3k$  -- (D)

Since 
$$H(1) = 1$$
,  $H(k) = 3k^2 - 3k + 1$ .

### Sequences of differences

#### Approach 3)

$$H(n) = \begin{cases} 1 & \text{if } n = 1 \\ H(n-1) + 6n - 6 & \text{if } n > 1 \end{cases}$$

The first six values are:

The second sequence of differences is constant. This suggests that the sequence may have a formula of the form

$$H(n) = An^2 + Bn + C.$$

### Sequences of differences

$$H(n) = An^2 + Bn + C -- (E)$$

Solving the system of equations

$$n = 1$$
:  $1 = A + B + C$   
 $n = 2$ :  $7 = 4A + 2B + C$   
 $n = 3$ :  $19 = 9A + 3B + C$ 

gives 
$$H(n) = 3n^2 - 3n + 1$$

### Inductively verifying a solution: general template

Given:

$$R(n)$$
 = some recurrence relation

$$f(n)$$
 = hypothesized closed-form formula

We wish to prove that R(n) = f(n) for all  $n \ge 0$ . A proof by induction has the following parts:

Base Case: Verify that R(0) = f(0).

**Inductive Hypothesis:** Let  $k \ge 0$  be some (unspecified) integer. Suppose as *inductive hypothesis* that R(k-1) = f(k-1).

Part of Inductive Step

**Inductive Step:** Prove that R(k) = f(k).

### Why bother proving anything?

Let P(n) be defined by the following recurrence relation.

$$P(n) = \begin{cases} 1 & \text{if } n = 0 \\ 3 \cdot P(n-1) - (n-1)^2 & \text{if } n > 0 \end{cases}$$

Is  $f(n) = (n+2) \cdot 2^{n-1}$  the solution for P(n)?

$$f(0) = 2 \times 2^{-1} = 1 = P(0) = 1$$
  
 $f(1) = 3 \times 2^{0} = 3 = P(1) = 3 \times P(0) - 0^{2} = 3$   
 $f(2) = 4 \times 2^{1} = 8 = P(2) = 3 \times P(1) - 1^{2} = 8$   
 $f(3) = 5 \times 2^{2} = 20 = P(3) = 3 \times P(2) - 2^{2} = 20$ 

### Why bother proving anything?

Let P(n) be defined by the following recurrence relation.

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Is  $f(n) = (n+2) \cdot 2^{n-1}$  the solution for P(n)?

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 $f(3) = 5 \times 2^{2} = 20 = P(3) = 3 \times P(2) - 2^{2} = 20$   
 $f(4) = 6 \times 2^{3} = 48 \neq P(4) = 3 \times P(3) - 3^{2} = 51$ 

# Verifying closed form solutions – Example 1

Let C(n) be defined by the following recurrence relation (giving the size of the power set of a set with n elements).

$$C(n) = \begin{cases} 1 & \text{if } n = 0 \\ 2 \cdot C(n-1) & \text{if } n > 0 \end{cases}$$

Prove that  $C(n) = 2^n$  for all  $n \ge 0$ .

### Proof:

We use induction on n. Let  $f(n) = 2^n$ .

**Base Case:** If n = 0, the recurrence relation says that C(0) = 1, and the formula says that  $f(0) = 2^0 = 1$ , so C(0) = f(0).

**Inductive Hypothesis:** Let k > 0. Suppose as inductive hypothesis that

$$C(k-1)=2^{k-1}.$$

Part of inductive step

**Inductive Step:** Using the recurrence relation,

$$C(k) = 2 \cdot C(k-1)$$
, by the second part of the recurrence relation  $= 2 \cdot 2^{k-1}$ , by inductive hypothesis  $= 2^k$ 

so, by induction, 
$$C(n) = 2^n$$
 for all  $n \ge 0$ .

# Verifying closed form solutions – Example 2

Recall the Fibonacci numbers:

$$F(n) = \begin{cases} 1 & \text{if } n = 1 \text{ or } n = 2 \\ F(n-1) + F(n-2) & \text{if } n > 2 \end{cases}$$

#### Theorem

For  $n \geq 1$ , the nth Fibonacci number is

$$g(n) = \frac{\alpha^n - \beta^n}{\alpha - \beta} \qquad -- (1)$$

where

$$\alpha = \frac{1 + \sqrt{5}}{2}$$
 and  $\beta = \frac{1 - \sqrt{5}}{2}$ . — (2)

### Proof:

It is easy to check that  $\alpha$  and  $\beta$  are the solutions to the following equation.

$$x^2 = x + 1$$

Thus we can use this equation as an identity for both  $\alpha$  and  $\beta$ . We proceed by induction on n.

$$\alpha^2 = \alpha + 1$$

$$\beta^2 = \beta + 1$$

## Proof: (continued)

**Base Case:** If n = 1 or n = 2, the recurrence relation says that F(1) = 1 = F(2). The closed-form formula gives

$$F(1) = \frac{\alpha^1 - \beta^1}{\alpha - \beta} = \frac{\alpha - \beta}{\alpha - \beta} = 1$$

and

$$F(2) = \frac{\alpha^2 - \beta^2}{\alpha - \beta}$$

$$= \frac{(\alpha + 1) - (\beta + 1)}{\alpha - \beta}, \text{ using the identity}$$

$$= \frac{\alpha - \beta}{\alpha - \beta}$$

$$= 1$$

so the result holds for n = 1 and n = 2.

# Proof: (continued)

**Inductive Hypothesis:** Let k > 2. Suppose as inductive hypothesis that

$$F(i) = \frac{\alpha^{i} - \beta^{i}}{\alpha - \beta}$$

Part of inductive step

for all i such that  $1 \le i < k$ .

Inductive Step: Using the recurrence relation,

$$F(k) = F(k-1) + F(k-2)$$

$$= \frac{\alpha^{k-1} - \beta^{k-1}}{\alpha - \beta} + \frac{\alpha^{k-2} - \beta^{k-2}}{\alpha - \beta}, \text{ by inductive hypothesis}$$

$$= \frac{\alpha^{k-2}(\alpha + 1) - \beta^{k-2}(\beta + 1)}{\alpha - \beta}$$

$$= \frac{\alpha^{k-2}(\alpha^2) - \beta^{k-2}(\beta^2)}{\alpha - \beta} = \frac{\alpha^k - \beta^k}{\alpha - \beta}$$

as required.  $\square$ 

### **Quiz 16-1**

#### Which of the following is NOT true?

- (a) A recurrence relation is a kind of recursive definition.
- (b) A recursive algorithm is a kind of algorithm.
- (c) The complexity of an algorithm can be expressed as a recurrence relation.
- (d) The complexity of a recursive algorithm can be expressed as a recurrence relation.
- (e) A closed form solution of a recurrence relation should be verified for its correctness.

