

# COMP9020 Week 4

## Term 3, 2019

### Recursion

# Summary of topics

- Recursion
- Recursive Data Types
- Recursive programming
- Solving recurrences

# Summary of topics

- Recursion
- Recursive Data Types
- Recursive programming
- Solving recurrences

# Recursion

Fundamental concept in Computer Science

- Recursion in algorithms: Solving problems by reducing to smaller cases
  - Factorial
  - Towers of Hanoi
  - Mergesort, Quicksort

# Recursion

Fundamental concept in Computer Science

- Recursion in algorithms: Solving problems by reducing to smaller cases
  - Factorial
  - Towers of Hanoi
  - Mergesort, Quicksort
- Recursion in data structures: Finite definitions of **arbitrarily large** objects
  - Natural numbers
  - Words
  - Linked lists
  - Formulas
  - Binary trees

# Recursion

Fundamental concept in Computer Science

- Recursion in algorithms: Solving problems by reducing to smaller cases
  - Factorial
  - Towers of Hanoi
  - Mergesort, Quicksort
- Recursion in data structures: Finite definitions of **arbitrarily large** objects
  - Natural numbers
  - Words
  - Linked lists
  - Formulas
  - Binary trees
- Analysis of recursion: Proving properties
  - Recursive sequences (e.g. Fibonacci sequence)
  - Structural induction

# Recursion

Consists of a basis (B) and recursive process (R).

A sequence/object/algorithm is recursively defined when (typically)  
(B) some initial terms are specified, perhaps only the first one;  
(R) later terms stated as functional expressions of the earlier terms.

## NB

*(R) also called* **recurrence formula (especially when dealing with sequences)**

# Example: Factorial

## Example

Factorial:

$$(B) \quad 0! = 1$$

$$(R) \quad (n + 1)! = (n + 1) \cdot n!$$

fact( $n$ ):

$$(B) \quad \text{if}(n = 0): 1$$

$$(R) \quad \text{else: } n * \text{fact}(n - 1)$$



## Example: Euclid's gcd algorithm

### Example

$$\gcd(m, n) = \begin{cases} m & \text{if } m = n \\ \gcd(m - n, n) & \text{if } m > n \\ \gcd(m, n - m) & \text{if } m < n \end{cases}$$

## Example: Towers of Hanoi

- There are 3 towers (pegs)
- $n$  disks of decreasing size placed on the first tower
- You need to move all disks from the first tower to the last tower
- Larger disks cannot be placed on top of smaller disks
- The third tower can be used to temporarily hold disks

# Example: Towers of Hanoi

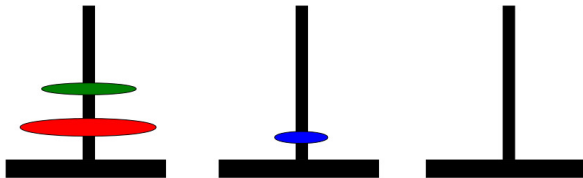
## Questions

- Describe a general solution for  $n$  disks
- How many moves does it take?

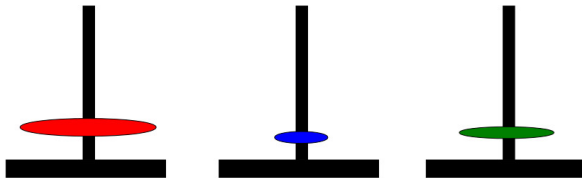
## Example: Towers of Hanoi



## Example: Towers of Hanoi



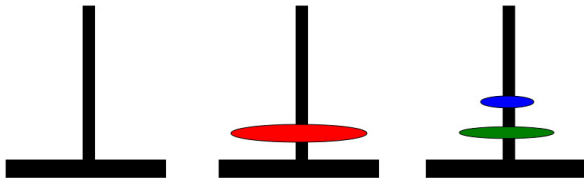
## Example: Towers of Hanoi



## Example: Towers of Hanoi

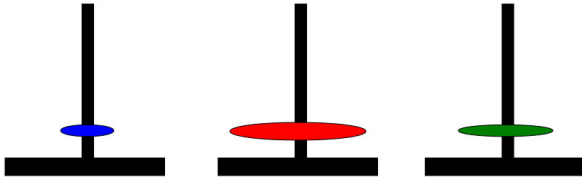


## Example: Towers of Hanoi

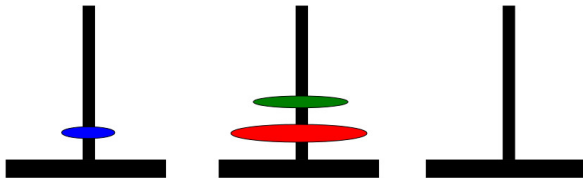




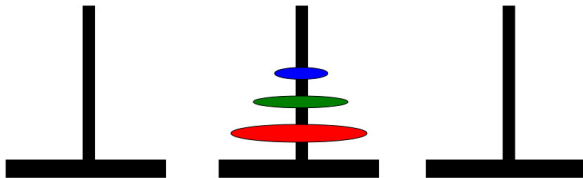
## Example: Towers of Hanoi



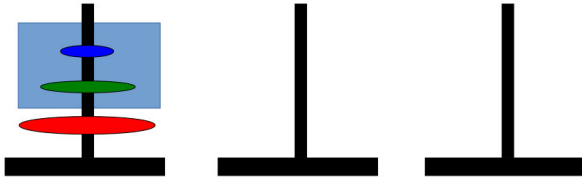
## Example: Towers of Hanoi



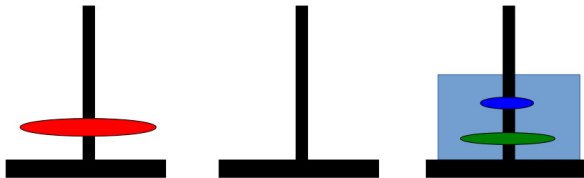
## Example: Towers of Hanoi



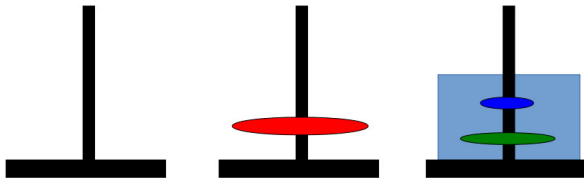
## Example: Towers of Hanoi



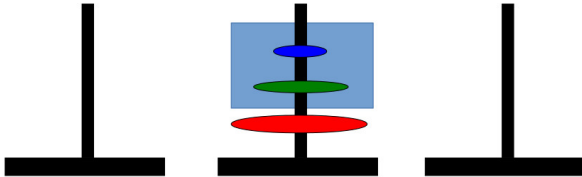
## Example: Towers of Hanoi



## Example: Towers of Hanoi



## Example: Towers of Hanoi



# Example: Towers of Hanoi

## Questions

- Describe a general solution for  $n$  disks
- How many moves does it take? ?



# Summary of topics

- Recursion
- Recursive Data Types
- Recursive programming
- Solving recurrences

## Example: Natural numbers

### Example

A natural number is either 0 (B) or one more than a natural number (R).

Formal definition of  $\mathbb{N}$ :

- (B)  $0 \in \mathbb{N}$
- (R) If  $n \in \mathbb{N}$  then  $(n + 1) \in \mathbb{N}$

## Example: Fibonacci numbers

### Example

The Fibonacci sequence starts  $0, 1, 1, 2, 3, \dots$  where, after  $0, 1$ , each term is the sum of the previous two terms.

Formally, the set of Fibonacci numbers:  $\mathbb{F} = \{F_n : n \in \mathbb{N}\}$ , where the  $n$ -th Fibonacci number  $F_n$  is defined as:

- (B)  $F_0 = 0$ ,
- (B)  $F_1 = 1$ ,
- (R)  $F_n = F_{n-1} + F_{n-2}$

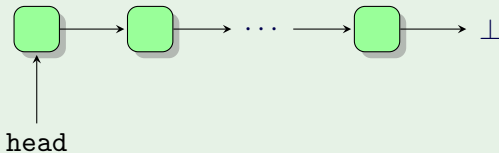
### NB

*Could also define the Fibonacci sequence as a function*  
 $\text{FIB} : \mathbb{N} \rightarrow \mathbb{F}$ . *Choice of perspective depends on what structure you view as your base object (ground type).*

## Example: Linked lists

### Example

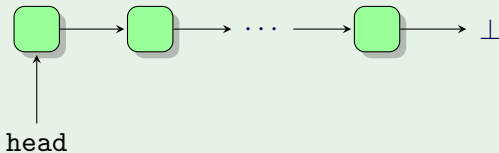
A linked list is zero or more linked list nodes:



## Example: Linked lists

### Example

A linked list is zero or more linked list nodes:



In C:

```
struct node{  
    int data;  
    struct node *next;  
}
```

## Example: Linked lists

### Example

We can view the linked list **structure** abstractly. A linked list is either:

- (B) an empty list, or
- (R) an ordered pair (Data, List).

## Example: Words over $\Sigma$

### Example

A word over an alphabet  $\Sigma$  is either  $\lambda$  (B) or a symbol from  $\Sigma$  followed by a word (R).

Formal definition of  $\Sigma^*$ :

- (B)  $\lambda \in \Sigma^*$
- (R) If  $w \in \Sigma^*$  then  $aw \in \Sigma^*$  for all  $a \in \Sigma$

### NB

*This matches the recursive definition of a **Linked List** data type.*

# Example: Propositional formulas

## Example

A well-formed formula (wff) over a set of propositional variables,  $\text{PROP}$  is defined as:

- (B)  $\top$  is a wff
- (B)  $\perp$  is a wff
- (B)  $p$  is a wff for all  $p \in \text{PROP}$
- (R) If  $\varphi$  is a wff then  $\neg\varphi$  is a wff
- (R) If  $\varphi$  and  $\psi$  are wffs then:
  - $(\varphi \wedge \psi)$ ,
  - $(\varphi \vee \psi)$ ,
  - $(\varphi \rightarrow \psi)$ , and
  - $(\varphi \leftrightarrow \psi)$  are wffs.



# Exercises

## Exercises

4.4.4 (a) Give a recursive definition for the sequence

$(2, 4, 16, 256, \dots)$

(b) Give a recursive definition for the sequence

$(2, 4, 16, 65536, \dots)$

# Exercises

## Exercises

4.4.4 (a) Give a recursive definition for the sequence

$$(2, 4, 16, 256, \dots)$$

?

(b) Give a recursive definition for the sequence

$$(2, 4, 16, 65536, \dots)$$

?

# Summary of topics

- Recursion
- Recursive Data Types
- Recursive programming
- Solving recurrences

# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

The factorial function:

```
fact( $n$ ):  
( $B$ )    if( $n = 0$ ): 1  
( $R$ )    else:  $n * \text{fact}(n - 1)$ 
```

# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

Summing the first  $n$  natural numbers:

```
sum( $n$ ):  
( $B$ )    if( $n = 0$ ): 0  
( $R$ )    else:  $n + \text{sum}(n - 1)$ 
```

# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

Summing elements of a linked list:

```
sum(L):  
(B)   if(L.isEmpty()):  
       return 0  
(R)   else:  
       return L.data + sum(L.next)
```

# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

Sorting elements of a linked list (insertion sort):

```
sort(L):  
  (B)    if(L.isEmpty()):  
          return L  
          else:  
  (R)    L2 = sort(L.next)  
          insert L.data into L2  
          return L2
```

# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

Concatenation of words (defining  $wv$ ):

$$\begin{array}{ll} & \text{For all } w, v \in \Sigma^* \text{ and } a \in \Sigma : \\ (B) & \lambda v = v \\ (R) & (aw)v = a(wv) \end{array}$$



# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

Length of words:

$$(B) \quad \text{length}(\lambda) = 0$$

$$(R) \quad \text{length}(aw) = 1 + \text{length}(w)$$

# Programming over recursive datatypes

Recursive datatypes make recursive programming/functions easy.

## Example

“Evaluation” of a propositional formula

# Exercise

## Exercise

Let  $\Sigma$  be a finite set.

Define  $\text{append} : \Sigma^* \times \Sigma \rightarrow \Sigma^*$  by

$$\text{append}(w, a) = wa$$

Give a (direct) definition of  $\text{append}$  [i.e. only concatenates symbols on the left].

# Exercise

## Exercise

Let  $\Sigma$  be a finite set.

Define  $\text{append} : \Sigma^* \times \Sigma \rightarrow \Sigma^*$  by

$$\text{append}(w, a) = wa$$

Give a (direct) definition of  $\text{append}$  [i.e. only concatenates symbols on the left].

?

## Pitfall: Correctness of Recursive Definition

A recurrence formula is correct if the computation of any later term can be reduced to the initial values given in (B).

### Example (Incorrect definition)

- Function  $g(n)$  is defined recursively by

$$g(n) = g(g(n-1) - 1) + 1, \quad g(0) = 2.$$

The definition of  $g(n)$  is incomplete — the recursion may not terminate:

Attempt to compute  $g(1)$  gives

$$g(1) = g(g(0) - 1) + 1 = g(1) + 1 = \dots = g(1) + 1 + 1 + 1 \dots$$

When implemented, it leads to an overflow; most static analyses cannot detect this kind of ill-defined recursion.

# Pitfall: Correctness of Recursive Definition

## Example (continued)

However, the definition could be repaired. For example, we can add the specification specify  $g(1) = 2$ .

Then  $g(2) = g(2 - 1) + 1 = 3$ ,  
 $g(3) = g(g(2) - 1) + 1 = g(3 - 1) + 1 = 4$ ,  
...

In fact, by induction ...  $g(n) = n + 1$

# Pitfall: Correctness of Recursive Definition

Check your base cases!

## Example

Function  $f(n)$  is defined by

$$f(n) = f(\lceil n/2 \rceil), \quad f(0) = 1$$

When evaluated for  $n = 1$  it leads to

$$f(1) = f(1) = f(1) = \dots$$

This one can also be repaired. For example, one could specify that  $f(1) = 1$ .

This would lead to a constant function  $f(n) = 1$  for all  $n \geq 0$ .

# Mutual Recursion

Sometimes recursive definitions use more than one function, with each calling each other.

## Example (Fibonacci, again)

Recall:

- (B)  $f(0) = 0$ ;  $f(1) = 1$ ,
- (R)  $f(n) = f(n-1) + f(n-2)$



# Mutual Recursion

Sometimes recursive definitions use more than one function, with each calling each other.

## Example (Fibonacci, again)

Recall:

- (B)  $f(0) = 0$ ;  $f(1) = 1$ ,
- (R)  $f(n) = f(n-1) + f(n-2)$

Alternative, mutually recursive definition:

- (B)  $f(1) = 1$ ;  $g(1) = 0$
- (R)  $f(n) = f(n-1) + g(n-1)$
- (R)  $g(n) = f(n-1)$

# Mutual Recursion

Sometimes recursive definitions use more than one function, with each calling each other.

## Example (Fibonacci, again)

Recall:

- (B)  $f(0) = 0$ ;  $f(1) = 1$ ,
- (R)  $f(n) = f(n-1) + f(n-2)$

Alternative, mutually recursive definition:

- (B)  $f(1) = 1$ ;  $g(1) = 0$
- (R)  $f(n) = f(n-1) + g(n-1)$
- (R)  $g(n) = f(n-1)$

$$\begin{pmatrix} f(n) \\ g(n) \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} f(n-1) \\ g(n-1) \end{pmatrix}$$

# Summary of topics

- Recursion
- Recursive Data Types
- Recursive programming
- Solving recurrences

# Solving recurrences

Approaches:

- Unwinding the recurrence
- Approximating with big-O
- The Master Theorem

## NB

*Each approach gives an informal “solution”: ideally one should prove a solution is correct (using e.g. induction).*

# Examples

## Example (Unwinding)

$$f(0) = 1 \quad f(n) = 2f(n-1)$$

# Examples

## Example (Unwinding)

$$f(0) = 1 \quad f(n) = 2f(n-1)$$

Unwinding:

$$\begin{aligned} f(n) &= 2f(n-1) \\ &= 2(2f(n-2)) = 4f(n-2) \\ &= 4(2f(n-3)) = 8f(n-3) \\ &\vdots \\ &= 2^i f(n-i) \\ &\vdots \\ &= 2^n f(0) = 2^n \end{aligned}$$

# Examples

## Example (Unwinding)

$$f(1) = 0 \quad f(n) = 1 + f\left(\left\lfloor \frac{n}{2} \right\rfloor\right)$$

# Examples

## Example (Unwinding)

$$f(1) = 0 \quad f(n) = 1 + f\left(\left\lfloor \frac{n}{2} \right\rfloor\right)$$

Unwinding:

$$\begin{aligned} f(n) &= 1 + f(n/2) \\ &= 1 + (1 + f(n/4)) = 2 + f(n/4) \\ &= 2 + (1 + f(n/8)) \\ &\quad \vdots \\ &= i + f(n/2^i) \\ &\quad \vdots \\ &= \log(n) + f(0) = \log(n) \end{aligned}$$



## Examples

### Example (Approximating with big-O)

$$f(0) = 1 \quad f(1) = 1 \quad f(n) = f(n-1) + f(n-2)$$

## Examples

### Example (Approximating with big-O)

$$f(0) = 1 \quad f(1) = 1 \quad f(n) = f(n-1) + f(n-2)$$

Assuming  $f(n)$  is increasing:

$$f(n-2) \leq f(n-1)$$

## Examples

### Example (Approximating with big-O)

$$f(0) = 1 \quad f(1) = 1 \quad f(n) = f(n-1) + f(n-2)$$

Assuming  $f(n)$  is increasing:

$$f(n-2) \leq f(n-1)$$

so:

$$f(n) \leq 2f(n-1)$$

## Examples

### Example (Approximating with big-O)

$$f(0) = 1 \quad f(1) = 1 \quad f(n) = f(n-1) + f(n-2)$$

Assuming  $f(n)$  is increasing:

$$f(n-2) \leq f(n-1)$$

so:

$$f(n) \leq 2f(n-1)$$

so (by unwinding):

$$f(n) \leq 2^n$$

## Examples

### Example (Approximating with big-O)

$$f(0) = 1 \quad f(1) = 1 \quad f(n) = f(n-1) + f(n-2)$$

Assuming  $f(n)$  is increasing:

$$f(n-2) \leq f(n-1)$$

so:

$$f(n) \leq 2f(n-1)$$

so (by unwinding):

$$f(n) \leq 2^n$$

so:

$$f(n) \in O(2^n)$$

# Master Theorem

The following result covers many recurrences that arise in practice (e.g. divide-and-conquer algorithms)

## Theorem

*Suppose*

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n)$$

*where  $f(n) \in \Theta(n^c(\log n)^k)$ .*

*Let  $d = \log_b(a)$ . Then:*

**Case 1:** *If  $c < d$  then  $T(n) = O(n^d)$*

**Case 2:** *If  $c = d$  then  $T(n) = O(n^c(\log n)^{k+1})$*

**Case 3:** *If  $c > d$  then  $T(n) = O(f(n))$*

# Master Theorem: Examples

## Example (Master Theorem)

$$T(n) = T\left(\frac{n}{2}\right) + n^2, \quad T(1) = 1$$

# Master Theorem: Examples

## Example (Master Theorem)

$$T(n) = T\left(\frac{n}{2}\right) + n^2, \quad T(1) = 1$$

Here  $a = 1$ ,  $b = 2$ ,  $c = 2$ ,  $k = 0$  and  $d = 0$ . So we have Case 3 and the solution is

$$T(n) = O(n^c) = n^2$$



# Master Theorem: Examples

## Example (Master Theorem)

Mergesort has

$$T(n) = 2T\left(\frac{n}{2}\right) + (n - 1)$$

for the number of comparisons.

# Master Theorem: Examples

## Example (Master Theorem)

Mergesort has

$$T(n) = 2T\left(\frac{n}{2}\right) + (n - 1)$$

for the number of comparisons.

Here  $a = b = 2$ ,  $c = 1$ ,  $k = 0$  and  $d = 1$ . So we have Case 2, and the solution is

$$T(n) = O(n^c \log(n)) = O(n \log(n))$$

# Master Theorem: Examples

## Example (Master Theorem)

Unwinding example:

$$T(1) = 0 \quad T(n) = 1 + T(\lfloor \frac{n}{2} \rfloor)$$

# Master Theorem: Examples

## Example (Master Theorem)

Unwinding example:

$$T(1) = 0 \quad T(n) = 1 + T(\lfloor \frac{n}{2} \rfloor)$$

Here  $a = 1$ ,  $b = 2$ ,  $c = 0$ ,  $k = 0$ , and  $d = 0$ . So we have Case 2, and the solution is

$$T(n) = O(\log(n))$$

# The Master Theorem: Pitfalls

## NB

- $a, b, c, k$  have to be constants (not dependent on  $n$ ).
- *Only one recursive term.*
- Recursive term is of the form  $T(n/b)$ , not  $T(n - b)$ .
- Solution is only an asymptotic bound.

## Examples

The Master theorem does not apply to any of these:

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

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 2T(n - 1)$$

# The Master Theorem: Linear differences

## NB

*The Master Theorem applies to recurrences where  $T(n)$  is defined in terms of  $T(n/b)$ ; not in terms of  $T(n-1)$ .*

However, the following is a consequence of the Master Theorem:

## Theorem

*Suppose*

$$T(n) = a \cdot T(n-1) + bn^k$$

*Then*

$$T(n) = \begin{cases} O(n^{k+1}) & \text{if } a = 1 \\ O(a^n) & \text{if } a > 1 \end{cases}$$

# Exercise

## Exercise

Solve  $T(n) = 3^n T(\frac{n}{2})$  with  $T(1) = 1$

# Exercise

## Exercise

Solve  $T(n) = 3^n T(\frac{n}{2})$  with  $T(1) = 1$

?