

Föreläsning 13

Recursion

TDDD86: DALP

Utskriftsversion av Föreläsning i *Datastrukturer, algoritmer och programmeringsparadigm*
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IDA, Linköpings universitet

13.1

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1 Introduction

Recursion

- **recursion:** Defining an operation in terms of itself
 - To solve a problem recursively requires solving smaller instances of the same problem
- **recursive programming:** Write functions that call themselves to solve problems recursively
 - As powerful as *iteration* (loops)
 - Particularly suitable for certain types of problems

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Why learn recursion??

- **“Cultural experience”:** Another way to think about problem solving.
- **powerful:** can solve certain types of problems better than iteration
- Can result in elegant, simple and short code (if used correctly)
- Many (functional languages such as Scheme, ML and Haskell) programming languages use recursion exclusively (no loops)
- A key component in many of the remaining labs in the course

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Recursion and case analysis

- Any recursive algorithm involves at least two cases:
 - **base case:** A simple instance of the problem that can be solved directly.
 - **recursive case:** A more complex instance of the problem for which the solution can be described in terms of solutions to smaller instances of the same problem..
 - Some recursive algorithms have more than one base case. All have at least one.
 - Key to recursive programming is to identify these cases.

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2 Recursion in C++

Recursion i C++

- Consider the following function to write a line of stars

```
// Prints a line containing the given number of stars.  
// Precondition: n >= 0  
void printStars(int n) {  
    for (int i = 0; i < n; i++) {  
        cout << "*";  
    }  
    cout << endl; // end the line of output  
}
```

- Write a recursive version of the function (it should call itself).
 - Solve the problem without using loops.
 - Tips: Your solution should write a single star at a time.

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Use recursion correctly

- Condense recursive cases to one case:

```
void printStars(int n) {  
    if (n == 1) {  
        // base case; just print one star  
        cout << "*" << endl;  
    } else {  
        // recursive case; print one more star  
        cout << "*";  
        printStars(n - 1);  
    }  
}
```

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“Recursion-zen”

- The actual, simpler, base case is when n is 0, not 1:

```
void printStars(int n) {  
    if (n == 0) {  
        // base case; just end the line of output  
        cout << endl;  
    } else {  
        // recursive case; print one more star  
        cout << "*";  
        printStars(n - 1);  
    }  
}
```

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Exercise - printBinary

- Write a recursive function printBinary that takes a natural number and that writes it in base 2 (binary)

- Example: printBinary(7) prints 111
- Example: printBinary(12) prints 1100

plats	10	1
värde	4	2

32	16	8	4	2	1
1	0	1	0	1	0

- Example: printBinary(42) prints 101010
- Write a recursive function without loops

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Case analysis

- Recursion is about solving parts of a larger problem
 - what is 69743 in base 2?
 - * what do we know about its representation in base 2?
 - Case analysis:
 - * Which numbers are simple to write in base 2?
 - * Can we express a larger number in terms of (some) smaller one(s)?

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Find the pattern

- Assume an arbitrary number N .
 - If the representation of N in base 2 is
 - Then the representation of $(N/2)$
 - and the representation of $(N\%2)$ is
 - * What can we deduce?

10010101011

1001010101

1

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Solution - printBinary

```
// Prints the given integer's binary representation.
// Precondition: n >= 0
void printBinary(int n) {
    if (n < 2) {
        // base case; same as base 10
        cout << n;
    } else {
        // recursive case; break number apart
        printBinary(n / 2);
        printBinary(n % 2);
    }
}
```

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Exercise - reverseLines

- Write a recursive function `reverseLines` that takes a file stream as input and that prints the lines

Exempelindatafil:

```
Roses are red,
Violets are blue.
All my base
Are belong to you.
```



Förväntat utdata:

```
Are belong to you.
All my base
Violets are blue.
Roses are red,
```

in reverse order

- Which cases should be considered?
 - * How can we solve part of the problem at a time?
 - * What would be a file that is easy to reverse?

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Pseudocode for reversing

- Reverse lines in a file:
 - Read a line L from the file
 - Print the rest of the lines in reverse order.
 - Print the line L
- If we only could reverse the or the lines in the file...

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Solution - reverseLines

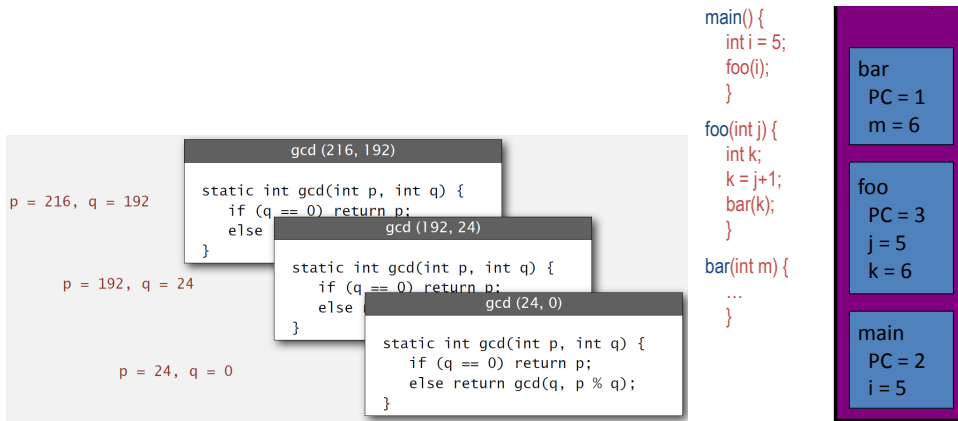
```
void reverseLines(ifstream& input) {
    string line;
    if (getline(input, line)) {
        // recursive case
        reverseLines(input);
        cout << line << endl;
    }
}
```

- What is the base case?

2.1 Implementation av recursion

Recall: stacks and function calls

- Compiler implement functions:
 - Function calls: *push*: a local context and return address
 - Return: *pop*: a return address and local context
 - This enables recursion.



2.2 Tail recursion

Tail recursion

A recursive call is *tail recursive* iff the first instruction after the control gets back after the call is a **return**.

- The stack is not needed
- Tail recursive functions can be rewritten into iterative functions

The recursive call in FACT is *not* tail recursive:

```
function FACT(n)
    if n = 0 then return 1
    else return n * FACT(n - 1)
```

First instruction after the return from the recursive call is a *multiplication* $\Rightarrow n$ needs to be kept on the stack

A tail recursive function

```
function BINSEARCH(v[a, ..., b], x)
    if a < b then
        m ← ⌊(a+b)/2⌋
        if v[m].key < x then
            return BINSEARCH(v[m+1, ..., b], x)
        else return BINSEARCH(v[a, ..., m], x)
    if v[a].key = x then return a
    else return 'not found'
```

The two recursive calls are *tail recursive*.

Eliminating tail recursion

The two tail recursive calls can be eliminated:

```
1: function BINSEARCH(v[a,...,b],x)
2:   if a < b then
3:     m ← ⌊ (a+b)/2 ⌋
4:     if v[m].key < x then
5:       a ← m + 1 {was: return BINSEARCH(v[m+1,...,b],x)}
6:     else b ← m {was: return BINSEARCH(v[a,...,m],x)}
7:     goto (2)
8:   if v[a].key = x then return a
9:   else return 'not found'
```

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Tail recursive factorial

fact can be rewritten by using a help function:

```
function FACT(n)
  return FACT2(n, 1)
```

```
function FACT2(n, f)
  if n = 0 then return f
  else return FACT2(n - 1, n · f)
```

FACT2 is *tail recursive* ⇒ *memory usage after eliminating the recursive in O(1)*

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2.3 One more exercise

Exercise - pow

- Write a recursive function `pow` that takes two natural numbers as arguments: a base and an exponent and that returns the base to the power of the exponent.
 - Example: `pow(3, 4)` returns 81
 - Solve the problem recursively without loops

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Solution - pow

```
// Returns base ^ exponent.
// Precondition: exponent >= 0
int pow(int base, int exponent) {
  if (exponent == 0) {
    // base case; any number to 0th power is 1
    return 1;
  } else {
    // recursive case: x^y = x * x^(y-1)
    return base * pow(base, exponent - 1);
  }
}
```

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An optimization?

- Observe the following mathematical properties:
$$3^{12} = 531441 = 9^6 = (3^2)^6$$
$$531441 = (9^2)^3 = ((3^2)^2)^3$$
 - When does this work?
 - How can we leverage on it?
 - Why use it when the code already works?

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Solution 2 - pow

```
// Returns base ^ exponent.
// Precondition: exponent >= 0
int pow(int base, int exponent) {
    if (exponent == 0) {
        // base case; any number to 0th power is 1
        return 1;
    } else if (exponent % 2 == 0) {
        // recursive case 1: x^y = (x^2)^(y/2)
        return pow(base * base, exponent / 2);
    } else {
        // recursive case 2: x^y = x * x^(y-1)
        return base * pow(base, exponent - 1);
    }
}
```

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3 Algorithm analysis

3.1 Analysis of algorithms

Analysis of algorithms

What is analysis?

- Correctness (not in this course)
- Termination (not in this course)
- Efficiency, resources, complexity

Time complexity — how long it takes an algorithm in the worst case?

- as a function of what?
- what is a time step?

Memory complexity — how much memory is required?

- as a function of what?
- how is it measured?
- remember that code and function calls also takes memory

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How can you compare different effectiveness

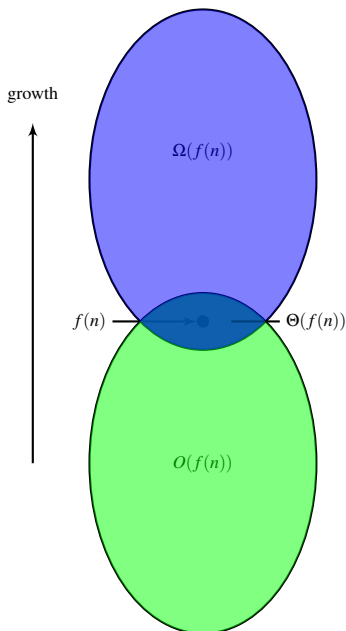
- Study execution time (or memory consumption) in function of the size of input data.
- When can we say that two algorithms have "similar effectiveness"?
- When can we say that an algorithm is better than an other?

	n	$\log_2 n$	n	$n \log_2 n$	n^2	2^n
Comparison between some elementary functions	2	1	2	2	4	4
	16	4	16	64	256	$6.5 \cdot 10^4$
	64	6	64	384	4096	$1.84 \cdot 10^{19}$

$1.84 \cdot 10^{19} \mu\text{seconds} = 2.14 \cdot 10^8 \text{ days} = 583.5 \text{ millennia}$

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How complexity can be specified?



- How does the complexity grow with the size n of input data?
- Asymptotic complexity — what happens when n grows to infinity?
- Much easier if we ignore constant factors
- $O(f(n))$ – grows at most as fast as $f(n)$
- $\Omega(f(n))$ – growth at least as fast as $f(n)$
- $\Theta(f(n))$ – grows as fast as $f(n)$

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Ordo-notation

f, g : grow from \mathbb{N} to \mathbb{R}^+

- $f \in O(g)$ if and only if it exists $c > 0, n_0 > 0$ such as $f(n) \leq c \cdot g(n)$ for all $n \geq n_0$ Intuition: ignoring the constant factor, f does not grow faster than g
- $f \in \Omega(g)$ if and only if it exists $c > 0, n_0 > 0$ such as $f(n) \geq c \cdot g(n)$ for all $n \geq n_0$ Intuition: ignoring the constant factor, f grows at least as fast as g
- $f(n) \in \Theta(g(n))$ if and only if $f(n) \in O(g(n))$ and $g(n) \in O(f(n))$ Intuition: ignoring the constant factor, f and g have similar growth

Note: Ω is the opposite of O , i.e. $f \in \Omega(g)$ if and only if $g \in O(f)$.

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3.2 Recursive algorithms

Execution time for recursive algorithms

- Characterize execution time with a recursive relation
- Find a solution in closed form the recursive relation
- If you do not recognize the recursive relation, you can
 - “Unroll” the relation a number of times to formulate a hypothesis for a possible solution of the form $T(n) = \dots$
 - Prove the hypothesis about $T(n)$ by induction. If it does not work, modify the hypothesis and try again. . .

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Example: Factorial function

```

function FACT( $n$ )
  if  $n = 0$  then return 1
  else return  $n \cdot \text{FACT}(n - 1)$ 

```

Execution time:

- time for comparison: t_c
- time for multiplication: t_m
- time for calls and returns: t_r

Total execution time $T(n)$. $T(0) = t_r + t_c$ $T(n) = t_r + t_c + t_m + T(n-1)$, if $n > 1$ Hence, for $n > 0$:

$$\begin{aligned}
 T(n) &= (t_r + t_c + t_m) + (t_r + t_c + t_m) + T(n-2) = \\
 &= (t_r + t_c + t_m) + (t_r + t_c + t_m) + (t_r + t_c + t_m) + T(n-3) = \dots = \\
 &= \underbrace{(t_r + t_c + t_m) + \dots + (t_r + t_c + t_m)}_{n \text{ ggr}} + t_r + t_c = n \cdot (t_r + t_c + t_m) + t_r + t_c \in \Theta(n)
 \end{aligned}$$

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Example: Binary search

```

function BINSEARCH(v[a,...,b],x)
  if a < b then
    m ← ⌊(a+b)/2⌋
    if v[m].key < x then
      return BINSEARCH(v[m+1,...,b],x)
    else return BINSEARCH(v[a,...,m],x)
  if v[a].key = x then return a
  else return 'not found'
  
```

Let $T(n)$ be the time, in the worst case, to search among n numbers with BINSEARCH.

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ T(\lceil \frac{n}{2} \rceil) + \Theta(1) & \text{if } n > 1 \end{cases}$$

If $n = 2^m$ we get

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ T(\frac{n}{2}) + \Theta(1) & \text{if } n > 1 \end{cases}$$

We can then conclude that $T(n) = \Theta(\log n)$.

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Master theorem

Sats 1 ("Master theorem"). Assume $a \geq 1, b > 1, d > 0$. The recursive relation

$$\begin{cases} T(n) &= aT(\frac{n}{b}) + f(n) \\ T(1) &= d \end{cases}$$

has the following asymptotic solution

- $T(n) = \Theta(n^{\log_b a})$ if $f(n) \in O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$
- $T(n) = \Theta(n^{\log_b a} \log n)$ if $f(n) \in \Theta(n^{\log_b a})$
- $T(n) = \Theta(f(n))$ if $f(n) \in \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$ and $af(\frac{n}{b}) \leq c \cdot f(n)$ for some constant $c < 1$ for all large enough n .

Examples:

- $T(n) = 9T(n/3) + n$
- $T(n) = T(2n/3) + 1$
- $T(n) = 3T(n/4) + n \log n$

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3.3 Common growth rates

Common growth rates

Growth	typical code	description	example	$T(2n)/T(n)$
1	a = b + c	instruction	add two numbers	1
$\log_2 n$	while (n > 1) { n = n / 2; ... }	divide in halves	binary search	≈ 1
n	for (int i = 0; i < n, i++) { ... }	loop	find maximum	2
$n \log_2 n$	see lecture on mergesort	divide and conquer	mergesort	≈ 2
n^2	for (int i = 0; i < n, i++) for (int j = 0; j < n, j++) { ... }	double loop	check all pairs	4
n^3	for (int i = 0; i < n, i++) for (int j = 0; j < n, j++) for (int k = 0; k < n, k++) { ... }	triple-loop	check all triples	8
2^n	see next lecture	total-search	check all subsets	$T(n)$

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