University of Ljubljana, Faculty of Computer and Information Science

# Analysis of Algorithms and Heuristic Problem Solving



Prof Dr Marko Robnik-Šikonja

Ljubljana, February 2023

### Lecturer

- Prof Dr Marko Robnik-Šikonja
- marko.robnik@fri.uni-lj.si
- FRI, Večna pot 113, room 2.06, 2<sup>nd</sup> floor, right from the elevator
- (01) 4798 241
- Contact hour (see webpage)
  - currently, Wednesdays, 13:00 14:00 or by arrangement, best to email me
- <a href="https://fri.uni-lj.si/en/employees/marko-robnik-sikonja">https://fri.uni-lj.si/en/employees/marko-robnik-sikonja</a>
- Research interests:

data science, machine learning, artificial intelligence natural language processing, network analytics, algorithms and data structures

### Assistant

- Dr Matej Pičulin <u>matej.piculin@fri.uni-lj.si</u>
- Laboratory for Cognitive Modeling
- tutorials mainly in the form of consultations; please, prepare questions!



### Objectives

- Students shall become acquainted with
  - the analysis of algorithms, at foremost computational complexity,
  - techniques for efficient solving of difficult problems, requiring optimization techniques and approximations.
- Practical use of theoretical knowledge on (almost) real-world problems.
- Increase the problem-solving toolbox with
  - new techniques for analysis of algorithms,
  - heuristic optimization algorithms.
- For a given optimization problem, students shall be able to
  - select one of the appropriate methods,
  - construct a solution prototype.



<sup>&</sup>quot;Thinking outside of the box is difficult for some people. Keep trying."

### Lectures and tutorials

- Lectures:
  - introduction to the topic, discussion,
  - some examples,
  - broader view of the topic.
- Tutorials:
  - exercises,
  - assignments motivated by practical use,
  - assistant presents the assignments, helps with tips, moderates discussion so...
  - ... come prepared and pose questions.
  - Introduce some problem solving tools and useful software.

### Syllabus

- 1<sup>st</sup> part:
  - computational complexity,
  - analysis of algorithms,
  - some problems turn out to be too difficult for solving exactly, so we need approximation methods and heuristic approaches,
- 2<sup>nd</sup> part:
  - heuristic programming,
  - introduction to some heuristic approaches using
    - operation research approaches,
    - population techniques
    - metaheuristics
  - how to approach real-world problems.

### More details

Lecture topics:

- 1. Analysis of recursive algorithms: recursive tree method, substitution method, solution for divide and conquer approach, Akra-Bazzi method.
- 2. Probabilistic analysis: definition, analysis of stochastic algorithms.
- 3. Randomization of algorithms.
- 4. Amortized analysis of algorithm complexity.
- 5. Solving linear recurrences.
- 6. Analysis of multithreaded, parallel and distributed algorithms.
- 7. Linear programming for problem solving.
- 8. Combinatorial optimization, local search, simulated annealing.
- 9. Metaheuristics and stochastic search: guided local search, variable neighbourhood search, and tabu search.
- 10. Memetic algorithms, particle swarm optimization, grey wolf, whales, bees, etc.
- 11. Differential evolution.
- 12. Machine learning for combinatorial optimization.
- 13. Many (almost) practical problems; interspersed within other topics

# Obligations

- 5 quizzes checking continuous work; obtaining at least 50% of points altogether is necessary,
- 5 assignments of different difficulty, practical and theoretical assignments, written reports, one assignment is in the form of competition and public presentation,
- written exam.

### Learning materials

- learning materials in the eClassroom <u>http://ucilnica.fri.uni-lj.si</u>
- practical work in open-source system R,
- optionally in Python, java, C/C++

### Readings



- T.H. Cormen, C.E. Leiserson, R.L. Rivest, C. Stein: *Introduction to Algorithms, 3<sup>rd</sup> or 4<sup>th</sup> edition*. MIT Press, 2009, 2022
- M. Gendreau, J-Y. Potvin (Eds.): *Handbook of Metaheuristics*, 2<sup>nd</sup> edition. Springer 2010

#### Further readings:

- R. Sedgewick, P. Flajolet: An Introduction to the Analysis of Algorithms. Addison-Wesley, 1995
- scientific papers, some on eClassroom

Review of existing knowledge on computational complexity

### Find the computational complexity

```
int i = n ;
int r =0 ;
while (i >1) {
    r = r+1 ;
    i = i / 2 ;
}
```

```
public static void loopRek(int m, int n)
{
    if (n == 1)
        System.out.println("+");
    else
    for (int i=0; i < m ; i++)
        loopRek(m, n-1);
}</pre>
```



### public static void infix(Node p)

```
if (p != null) {
    infix(p.left) ;
    System.out.print(p.key) ;
    infix(p.right) ;
```



first determine the parameter of complexity

```
max = a[1] ;
for (i=2 ; i <= n ; i++)
if (max < a[i])
max = a[i] ;
System.out.print(max) ;</pre>
```

```
max = a[1];
for (i=2; i <= n; i++)
  if (max < a[i]) {
     max = a[i];
     veryComplexOperation(max)
   }
  System.out.print(max);</pre>
```

```
void p(int n, int m) {
  int i,j,k ;
  if (n > 0) {
    for (i=0 ; i < m ; i++)
     for (j=0 ; j < m ; j++)
       if (i < j - a)
         for (k=0 ; k < m ; k++)
            System.out.println(i + j * k) ;
    p(n/m, m);
}
```

### Analysis of algorithms

- How complex is the algorithm?
- How many resources it requires?
- How much time, memory, etc. will the computer need?
- Resources: time, memory, network accesses, other hardware

# A simple model of computer - RAM

- RAM abstract uniprocessor machine with random access to the memory (RAM –Random-Access Machine)
- operations and their price (execution time, memory, etc.):
- typical operations: arithmetical and logical operations, memory operations, control
- each operation uses a constant amount of time
- integers and floating-point numbers
- numbers use a limited amount of memory; for example number n takes at most c log<sub>2</sub>(n) bits, where constant c >= 1 (what if it is not constant)
- we assume constant time for some other operations as well, e.g., logarithms, exponents, trigonometrical operations
- we do not consider parallelism, pipelines, memory hierarchies
- RAM is (good enough) approximation for real world computers

### Input size

- define for each problem separately
  - size of an input, e.g., array
  - number of bits in input
  - size of graph (nodes, edges)
  - number of steps taken,
  - etc.

### Execution time

- number of steps of the abstract machine
- for simpler analysis, we assume that each line of pseudocode requires a constant time (except function calls),
- so line i requires c<sub>i</sub> time

# An example: insertion sort

- execution time depends on input (number of elements, their initial positions)
- time: number of steps of abstract machine
- for the sake of simplicity, we assume a constant execution time for each line of pseudo-code, i.e., line i takes c<sub>i</sub> time, where c<sub>i</sub> is constant larger or equal zero
- idea: iteratively increase the sorted part of an array, by inserting unsorted elements into the already sorted part

### Pseudocode

InsertionSort(A) {

- 1 for j = 2 to A.length
- 2 key = A[j];
- 3 // insert A[j] into sorted array A[1..j-1]
- 4 i = j-1;
- 5 while i > 0 and A[i] > key

```
6 A[i+1] = A[i] ;
```

```
7 i = i -1 ;
```

8 A[i+1] = key ;

### Count the operations

In	SERTION-SORT $(A)$	cost	times
1	for $j = 2$ to A.length	$c_1$	п
2	key = A[j]	$C_2$	n-1
3	// Insert $A[j]$ into the sorted		
	sequence $A[1 \dots j - 1]$ .	0	n-1
4	i = j - 1	$C_4$	n-1
5	while $i > 0$ and $A[i] > key$	$C_5$	$\sum_{j=2}^{n} t_j$
6	A[i+1] = A[i]	$C_6$	$\sum_{j=2}^{n} (t_j - 1)$
7	i = i - 1	C7	$\sum_{j=2}^{n} (t_j - 1)$
8	A[i+1] = key	C 8	n-1

### Sum together

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8 (n-1).$$

• number of operations depends on the input

#### Best case

the best case is when the array is already sorted, then
 t<sub>j</sub> = 1, for j = 2,3,...,n and we get a linear dependency on n

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 (n-1) + c_8 (n-1)$$
  
=  $(c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8).$ 

### Worst case

 worst case occurs when the array is sorted in reversed order, then t<sub>j</sub> = j, for j=2,3, ..., n and we get

$$\sum_{j=2}^{n} j = \frac{n(n+1)}{2} - 1 \qquad \qquad \sum_{j=2}^{n} (j-1) = \frac{n(n-1)}{2}$$

n

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 \left(\frac{n(n+1)}{2} - 1\right) + c_6 \left(\frac{n(n-1)}{2}\right) + c_7 \left(\frac{n(n-1)}{2}\right) + c_8 (n-1) = \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right) n^2 + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8\right) n - (c_2 + c_4 + c_5 + c_8) .$$

which can be expressed as a quadratic dependency  $T(n) = an^2 + bn + c$ 

### Analysis

- we mostly analyze worst and average case complexities; why?
- we are rarely interested in actual constant and settle for the order of growth,
- in this case only the fastest growing terms are important, others are asymptotically unimportant,
- the worst case for the insertion sort is  $\Theta(n^2)$

	n		
	10		
	100		
	1.000		
	10.000		
	100.000		
	1.000.000		

√n	n		
3	10		
10	100		
31	1.000		
100	10.000		
316	100.000		
1.000	1.000.000		

log <sub>10</sub> (n)	√n	n		
1	3	10		
2	10	100		
3	31	1.000		
4	100	10.000		
5	316	100.000		
6	1.000	1.000.000		

log <sub>10</sub> (n)	√n	n	$n \cdot \log_{10}(n)$		
1	3	10	10		
2	10	100	200		
3	31	1.000	3.000		
4	100	10.000	40.000		
5	316	100.000	500.000		
6	1.000	1.000.000	6.000.000		

log <sub>10</sub> (n)	√n	n	$n \cdot \log_{10}(n)$	n <sup>2</sup>	
1	3	10	10	100	
2	10	100	200	10.000	
3	31	1.000	3.000	1.000.000	
4	100	10.000	40.000	108	
5	316	100.000	500.000	1010	
6	1.000	1.000.000	6.000.000	10 <sup>12</sup>	

log <sub>10</sub> (n)	√n	n	$\mathbf{n} \cdot \log_{10}(\mathbf{n})$	n <sup>2</sup>	n <sup>3</sup>	
1	3	10	10	100	1000	
2	10	100	200	10.000	1.000.000	
3	31	1.000	3.000	1.000.000	10 <sup>9</sup>	
4	100	10.000	40.000	108	10 <sup>12</sup>	
5	316	100.000	500.000	1010	1015	
6	1.000	1.000.000	6.000.000	10 <sup>12</sup>	10 <sup>18</sup>	

log <sub>10</sub> (n)	√n	n	$n \cdot \log_{10}(n)$	n <sup>2</sup>	n <sup>3</sup>	2 <sup>n</sup>
1	3	10	10	100	1000	1024
2	10	100	200	10.000	1.000.000	1.25· 10 <sup>30</sup>
3	31	1.000	3.000	1.000.000	10 <sup>9</sup>	10 <sup>301</sup>
4	100	10.000	40.000	108	10 <sup>12</sup>	$2 \cdot 10^{3.010}$
5	316	100.000	500.000	1010	1015	10 <sup>30.103</sup>
6	1.000	1.000.000	6.000.000	10 <sup>12</sup>	10 <sup>18</sup>	10 <sup>301.030</sup>