Principles of Computer Science II Computational Thinking

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Lecture 1

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to learn and use.

abstraction.

on concepts fundamental to computing.

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Wing, J. M. 2006 Computational thinking, CACM 49, 33-35 Computational thinking represents a universally applicable attitude and skill set everyone, not just computer scientists, would be eager

Wing, J. M. 2006 Computational thinking, CACM 49, 33-35 Thinking like a computer scientist means more than being able to

program a computer. It requires thinking at multiple levels of

The riddle of machine intelligence

Computational thinking confronts the riddle of machine intelligence:

- What can humans do better than computers?
- What can computers do better than humans? ▶ What is computable?

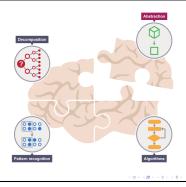
Computational Thinking

- Computers are here to help us.
- What do we need from computers?
 - What is our problem?
 - ► Computational Thinking allows us to understand what needs
 - to be solved. Four key techniques (cornerstones) to computational thinking:

within problems

- 1. Decomposition breaking down a complex problem or system into smaller, more manageable parts 2. Pattern Recognition - looking for similarities among and
- Abstraction focusing on the important information only. ignoring irrelevant detail
- 4. Algorithms developing a step-by-step solution to the problem, or the rules to follow to solve the problem

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Computational Thinking vs Programming

Thinking computationally is not programming.

- ...not even thinking as a computer.
- Programming tells computer what to do / how to do it.
- Computational thinking enables us to understand what we need to tell to computers.
- ...what to program.

Examples:

- Explain to a friend how to drive to your house
- ► Organize a party at the park
- ▶ Prepare your luggage
- ► Teach a kid addition/subtraction



Decomposition

Turn a complex problem into one we can easily understand.

- ... probably you already do every day.
 The smaller parts are easier to solve.
- ... we already know/have the solutions.

Examples:

- Brushing our teeth Which brush? How long? How hard? What toothpaste?
- Solving a crime What crime? When? Where? Evidence? Witnesses? Recent similar crimes?

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...

Pattern Recognition

We often find patterns among the smaller problems we examine.

The patterns are similarities or characteristics that some of the problems share.

Example: Cats

- All cats share common characteristics.
- they all have eyes, tails and fur.
- Once we know how to describe one cat we can describe others, simply by following this pattern.



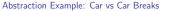


Abstraction

Hiding irrelevant details to focus on the essential features needed to understand and use a thing

- A compression process multiple different pieces of constituent data to a single piece of abstract data. e.g., "cat"
- Ambiguity multiple different references.
 e.g., "happiness", "architecture"
- ► Simplification no loss of generality
- e.g., "red" many different things can be red

Thought process wherein ideas are distanced from objects







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- Do we know how car breaks work?
- Do we know how to use them?

Filter out (ignore) the characteristics that we don't need in order to concentrate on those that we do.



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Algorithms

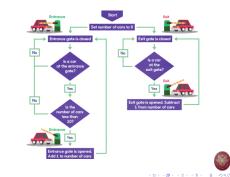
A plan, a set of step-by-step instructions to solve a problem.

In an algorithm, each instruction is identified and the order in which they should be carried out is planned.

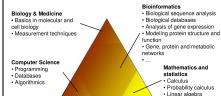








Bioinformatician's skill set



Where would you be in this triangle?

Prof. Juho Rousu, 2006

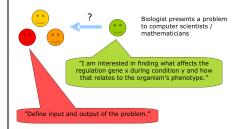
4 m x 4 m x 4 2 x 4 2 x 3 x 4 0 4 0 0



- Statistics, data analysis methods Lots of data

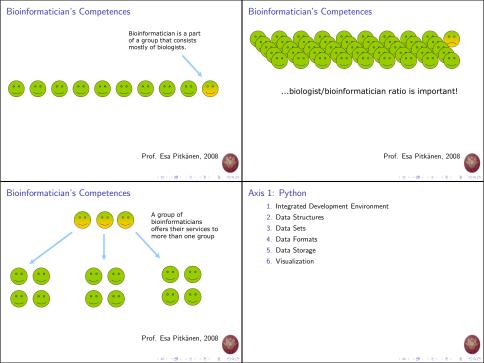
 - High noise levels, missing values
- ▶ #attributes ≫ #data points Programming languages
 - Scripting languages: Python, Perl. Ruby. . . .
 - Extensive use of text file formats: need parsers
 - Integration of both data and tools
- Data structures, databases
 - New measurement techniques produce huge quantities of biological data.
- Scientific computation packages
 - R. Matlab/Octave. . . .

Bioinformatician's Competences



Prof. Esa Pitkänen, 2008





Axis 2: Algorithms 2. Sorting Literature Jones Peyzner: An Introduction to Bioinformatics Algorithms. MIT Press, 2004

Axis 3: Cloud Computing 1. Complexity Analysis 3. Exhaustive Search 4. Branch-and-Bound Algorithms 5. Greedy Algorithms 6. Divide-and-Conquer Algorithms 7. Data Mining Algorithms

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1st Assignment

https://www.hackerrank.com/

 Complete all Python challenges under the following subdomains:

1. Cloud Storage

3. Elastic Compute

4. Handling Large Data Sets

Databases

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▶ Basic Data Types (6), Strings (10), Sets (13), Math (7),

IterTools (6), Collections (8)

► Total: 50

You can cooperate, You can search on the Internet, ... You need to write your own code ► Email ichatz@diag.uniroma1.it Subject: [PCS2] Homework 1

► Deadline: 16/October/2020, 23:59

A .zip or a .tar.gz file with your python solutions, for all challenges.

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