



University of
Central Lancashire
UCLan

CO2412

Computational Thinking

Module structure
What is computational thinking?

Where opportunity creates success

Resources

Recommended literature:

- K. Erciyes, *Discrete Mathematics and Graph Theory*, Cham: Springer (ISBN 978-3-03061114-9), **2021**.
- P. Sanders, K. Mehlhorn, M. Dietzfelbinger, R. Dementiev, *Sequential and Parallel Algorithms and Data Structures*, Cham: Springer (ISBN 978-3-03025208-3), **2021**.

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- K. Brennan, M. Resnick, "New frameworks for studying and assessing the development of computational thinking," in *Proceedings of AERA 2012*, Cambridge, MA: Academic Press, **2012**.

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Course website:

- <https://home.bawue.de/~horsch/teaching/co2412/>

All essential information will be made accessible through the course website.

What is computational thinking?

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Brennan & Resnick (2012) “have developed a definition of computational thinking that involves three dimensions:

- **computational concepts** (the concepts *designers* employ [...]),
- **computational practices** (the practices *designers* develop [...]), and
- **computational perspectives** (the perspectives *designers* form about the world around them and about themselves).”

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These concepts, practices, and perspectives are easy to develop even in children; e.g., for Scratch as evaluated by Brennan & Resnick (2012) at the time, “hundreds of thousands of young creators (mostly between the ages of 8 and 16)”.

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Programming practice

The official programming language of the module is Python.

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However, Computational Thinking is a theoretical course.

It does not matter what programming languages or environments you use to implement the considered data structures and algorithms.

Any code that is well written and documented will be accepted as part of the solution to a problem from our practical/tutorial sessions.

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Any code that is well written and documented will be accepted as part of the solution to a problem from our practical/tutorial sessions. If it does not run on my system (a rather average Linux installation) or if there are any other issues due to the employed environment, I may ask for clarification, e.g., by demonstrating your solution on the system where it was implemented.

In extreme, unexpected cases you may be required to rewrite code in Python.

Programming practice

Algorithms are best discussed using **pseudocode**:

```
INSERTION-SORT(A)
1  for j = 2 to A.length
2      key = A[j]
3      // Insert A[j] into the sorted sequence 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
```

Since the analysis of an algorithm should occur at a level of abstraction higher than that of its practical implementation (as code), simple generally comprehensible notations are often more suitable than syntactically correct code in any given programming language.

Learning outcomes

Upon successful completion of this module, a student will be able to:

- 1) Use methods including logic and probability to reason about algorithms and data structures;
- 2) Compare, select, and justify algorithms and data structures for a given problem;
- 3) Analyse the computational complexity of problems and the efficiency of algorithms;
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Module structure

On the topic of **program analysis**, we will:

- Consider the space (memory) and time efficiency of algorithms;
- Describe asymptotic scaling behaviour using Landau $O(n)$ notation;
- Analyse algorithms formally via pre-/postconditions of statements;
- Review concurrency and scalability for massively-parallel computing.

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Module structure

On the topic of **algorithm design**, we will:

- Compare and apply algorithm design strategies such as recursion, divide-and-conquer, greedy algorithms, dynamic programming;
- Look at common data structures and their specification and implementation;
- Apply algorithm design to *sorting* as a highly relevant use case.

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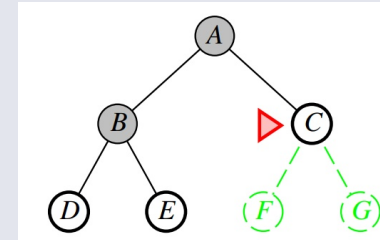
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Module structure

On the topic of **graphs and trees**, we will:



- Introduce graph theory and its basic definitions and concepts, including trees as a special case;
- Address basic tasks/problems when dealing with graphs, e.g., computing the shortest paths or connected components, strategies for graph traversal, and the application of trees to sorting and searching;
- Discuss numerical and mathematical representations of graphs.

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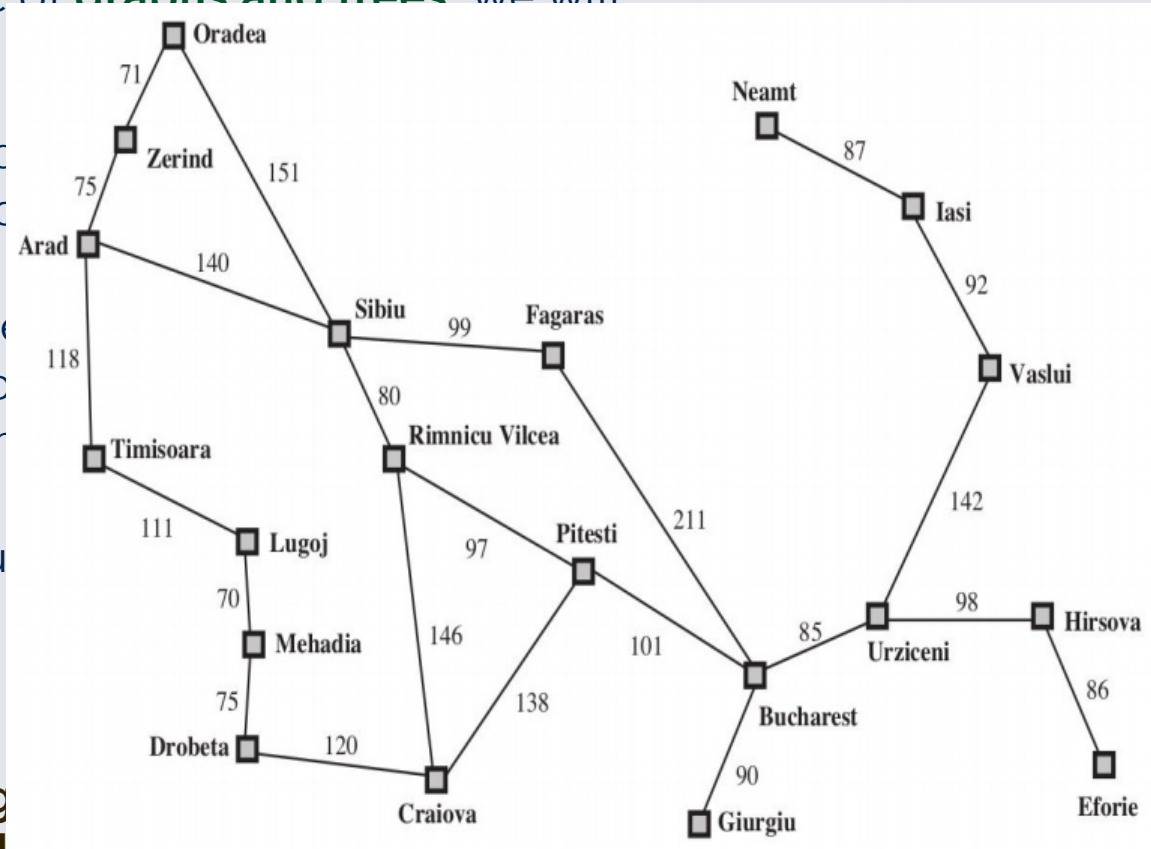
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Module structure

On the topic of **logic**, we will:

- Introduce propositional logic and the semantics of logical formulas;
- Consider deductive logical reasoning by inference and address basic problems for logical expressions such as their satisfiability;
- Introduce first-order logic as a powerful formalism by which most application scenarios, including from program analysis, can be covered.

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Module structure

On the topic of **formal languages**, we will:

- Formalize the definition of a computational problem in terms of a language and the word problem for a given language;
- Introduce ways for specifying formal languages, e.g., by regular expressions and more generally by generative grammars;
- Introduce finite automata and the Turing machine, a simple theoretical computer model for purposes of decidability and complexity analysis.

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On the topic of **complexity**, we will:

- Remove ourselves from considering specific algorithms by introducing complexity, i.e., the efficiency of the best possible algorithm for a given problem, as an additional layer of abstraction;
- Introduce the hierarchy of complexity classes, NP-complete problems, and the $P = NP$ problem;
- Characterize the complexity of common problems from graph theory and from logic.

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Module structure

On the topic of **randomness and probability**, we will:

- Introduce and/or review basic concepts from probability theory;
- Apply statistics and discrete mathematics to probability;
- Discuss randomness and random number generators;
- Consider randomized algorithms that can help at addressing computationally challenging problems.

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Grading

Number of Assessments	Form of Assessment	% weighting	Size of Assessment/ Duration/ Wordcount	Category of assessment	Learning outcomes being assessed
1	Examination	40%	1.5 hours	Written exam	1,2,3,4
1 (split into parts)	Practical work involving the selection, implementation and evaluation of algorithms and data structures	60%	2,000 words equivalent	Coursework	1,2,3,4,5

To pass this module, you must achieve a grade of 40% or above aggregated across all the assessments.



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