

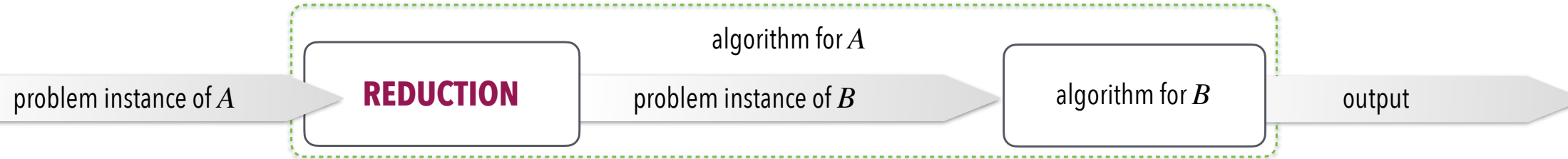
Towards Computer-Aided Teaching of Reductions in Theoretical Computer Science

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Reductions – A Formal Definition

- A problem A is called reducible onto B , if there is a computable function f such that $x \in A \Leftrightarrow f(x) \in B$ (f.a. x from the domain of A).
- Essentially, the above definition expresses the ability to **use a solution for B to solve problem A** .



Motivation & Research Questions

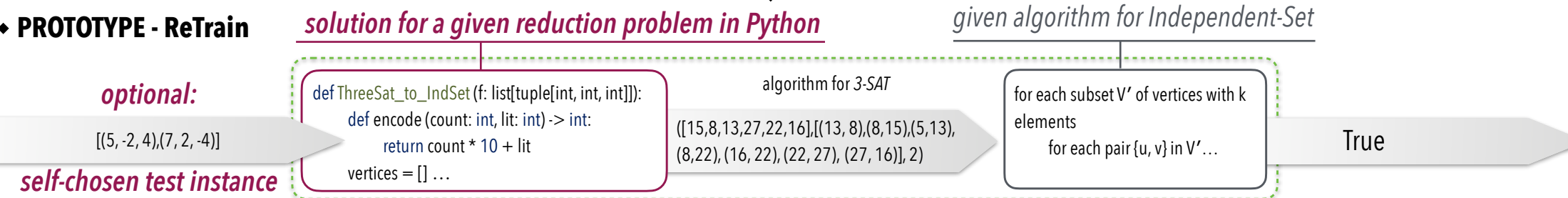
- "Reduction is an important method for solving problems in mathematics and science in general and is one of the main principles of computer science".¹
- There are different tools to support teaching and learning, mostly relying on reductions given in a domain-specific language. e.g.^{2,3}
- Our approach: Reductions as a programming exercise in a common and widespread language.

How can the abstraction level of reductions be minimized by testing worked examples in and with Python?

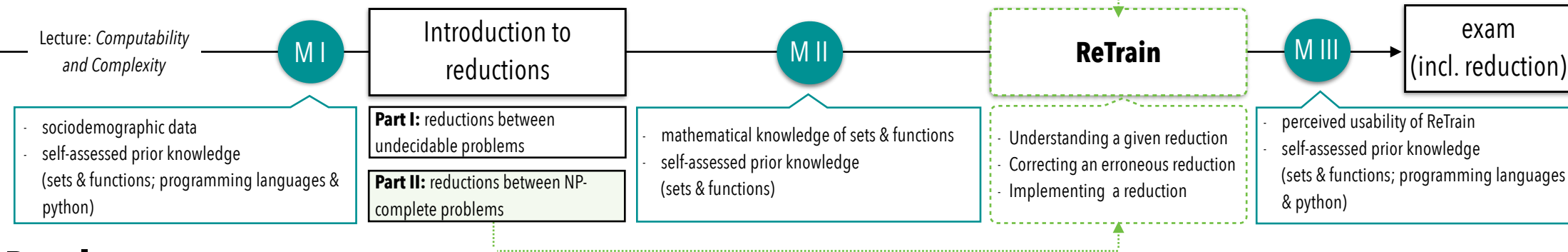
How and to what extent should automated feedback be provided to support learners in their learning process of reductions?

Tool-Based Teaching of Reductions – A Pilot Study

PROTOTYPE - ReTrain



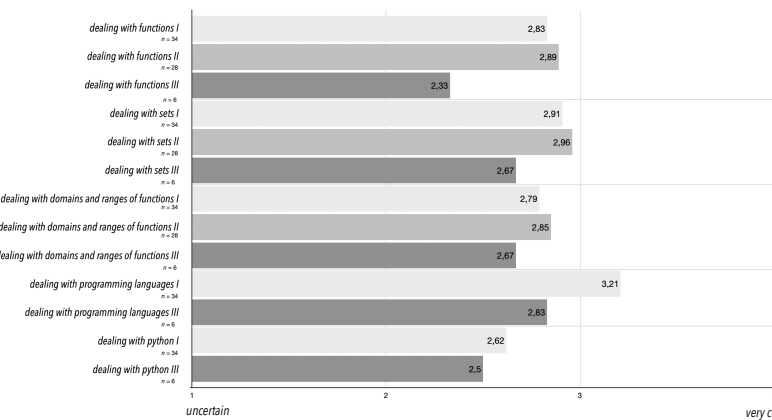
LEARNING ENVIRONMENT & METHOD



Results

$n = 41$; 1st- 3rd semester = 47.1 %; 4th-7th semester = 26.5 %; >7th semester = 26.5 %

- At the measurement points students assessed their mathematical (MI - MIII) and programming skills (MI & MIII) in different ways.



- The students demonstrated a range of proficiency in mathematical skills, with specific attention to reductions.

(1= incorrect; 2= correct)

- knowledge of sets: $M = 1.50$; $SD = .23$
- knowledge of functions: $M = 1.67$; $SD = .13$

- There is a partial correlation between the self-assessed prior knowledge, (i.e. dealing with sets) and mathematical skills.

	knowledge of sets	knowledge of functions
dealing with functions	$r = -.028, p = .893$	$r = .34, p = .103$
dealing with sets	$r = .051, p = .804$	$r = .35, p = .043, d = .24$
dealing with domains and ranges of functions	$r = -.008, p = .971$	$r = .38, p = .048, d = .28$

"sometimes a bit fiddly"
(student 4th - 7th semester)

"For me personally, the code window was
sometimes not displayed at all."
(student 1st - 3rd semester)

"I think it is very good that there is the possibility of testing reductions
using pre-defined queries. However, for some problems it is quite
complicated to specify the reduction, as many transformations have to be
made due to the encoding."
(student 1st - 3rd semester)

Results of telemetry data and qualitative feedback:

- testing of instances is perceived as helpful but troublesome due to abstract encodings
- only a few self-chosen examples were tested
- time consuming implementation of reductions

A Future Learning Tool for Reductions

EXTENSIBILITY

simplified generation of new reduction tasks

- support via pre-defined libraries and reusability of data from existing problems
- easy-to-understand and uniform description language for decision procedures cf.²

support of undecidable problems

- only limited feedback via testing (e.g. for semi-decidable problems)
- further research needed (e.g. identification of "didactic" subclasses that permit certain automatic feedback methods)

USABILITY

focus on side-aspects such as problem encodings needs to be reduced

- encodings/data-structures should be pre-defined and if possible visualized (e.g. graph representation)

more problem-specific feedback is needed

- type hints for instances of reduction problems
- (semi-)automatic assessment of correctness (e.g. by static code analysis)
- further research needed on typical classes of errors cf.⁴

individualized and simplified instance testing

- automatic generation of examples and provision of edge cases

Literature

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³ Zhang, C.; Hartline, J. D.; Dimoulas, C.: Karp: A Language for NP Reductions. In: Association for Computing Machinery (Eds.) Proceedings of the 43rd ACM SIGPLAN International Conference on Programming Language Design and Implementation. PLDI 2022, pp. 762–776, 2022.

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Acknowledgements

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