

Software technology for automated feedback generation

Bastiaan Heeren, ICS Colloquium December 17, 2020

Short bio

- Associate professor at the Open University of the Netherlands
- Head of OU's Computer Science Department
- At Utrecht University:
 - PhD in Software Technology (2000-2005)
 - Lecturer (2005-2007)
 - Guest researcher with the Software Technology for Learning and Teaching research group



Bastiaan Heeren

Intelligent Tutoring Systems

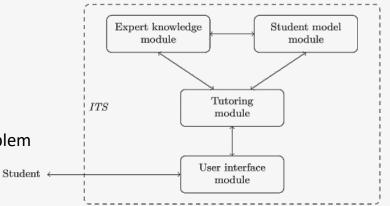
Intelligent Tutoring Systems (ITS): computer systems that provide immediate and customized feedback to learners

Structure:

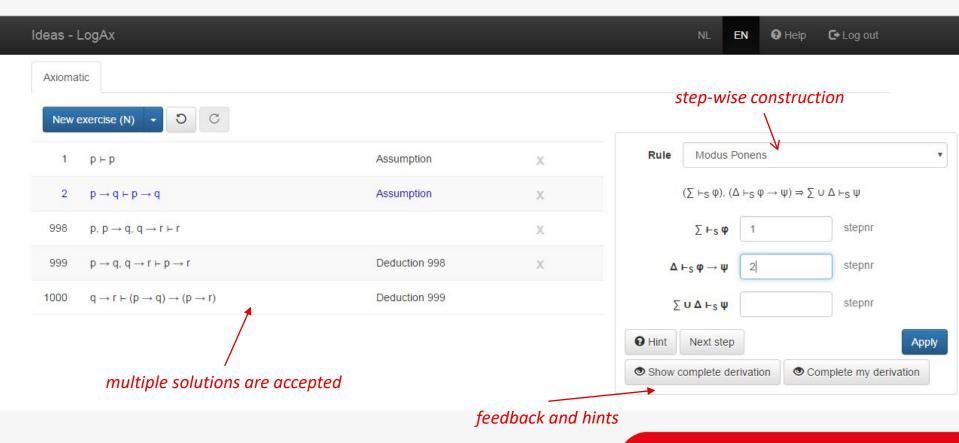
- Classical architecture with four components

Behaviour:

- Outer loop: solving one task after another
- Inner loop: the steps for solving one complex, multi-step problem



Example: axiomatic proofs



Research motivation

- 1. Simplify construction of ITSs (which are complex software systems)
- 2. Represent expert domain knowledge explicitly (for better feedback)
- 3. Apply approach to a wide range of problem domains

Approach: use software technology for automated feedback generation

Techniques in this presentation (outline):

- Rewrite strategies for automated feedback (basics)
- Light-weight rewrite rules
- Generic traversals

Problem domains



Four recent PhD theses, for different problem domains, all based on the same approach

Ideas framework

Generic framework for constructing domain reasoners

- Developed in Haskell
- Size: 12,397 LOC
- Open source
- Independent of problem domain
- http://ideas.cs.uu.nl/tutorial/

💡 Ideas

Interactive Domain-specific Exercise Assistants

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ideas: Feedback services for intelligent tutoring systems

[apache, education, library] [Propose Tags]

Ideas (Interactive Domain-specific Exercise Assistants) is a joint research project between the Open University of the Netherlands and Utrecht University. The project's goal is to use software and compiler technology to build state-of-the-art components for intelligent tutoring systems (ITS) and learning environments. The ideas software package provides a generic framework for constructing the expert knowledge module (also known as a domain reasoner) for an ITS or learning environment. Domain knowledge is offered as a set of feedback services that are used by external tools such as the digital mathematical environment (DME), MathDox, and the Math-Bridge system. We have developed several domain reasoners based on this framework, including reasoners for mathematics, linear algebra, logic, learning Haskell (the Ask-Elle programming tutor) and evaluating Haskell expressions, and for practicing communication skills (the serious game Communicate!).

Modules

[Index]

- Common
 - Ideas.Common.Classes Ideas.Common.Constraint Ideas.Common.Context Ideas.Common.Derivation Ideas.Common.DerivationTree

Versions [faq] 0.5.8, 0.6, 0.7, 1.0, 1.1, 1.2, 1.3, 1.3.1, 1.4, 1.5, 1.6, 1.7, **1.8**

Change log CHANGELOG.txt

Dependencies

base (>=4.8 & & <5), blaze-builder (>=0.4), bytestring, case-insensitive, containers, Diff, directory, filepath, HDBC, HDBC-sqlite3, http-types, mtl, network, parsec, QuickCheck (>=2.8 & & <2.12), random, semigroups (==0.18.⁵), streaming-commons (<0.2), time, uniplate, wai, wi-pyrint (details)

License

Apache-2.0

Copyright

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Category Education

Home page http://ideas.cs.uu.nl/

Source repo head: git clone https://github.com/ideasedu/ideas.git



Rewrite strategies

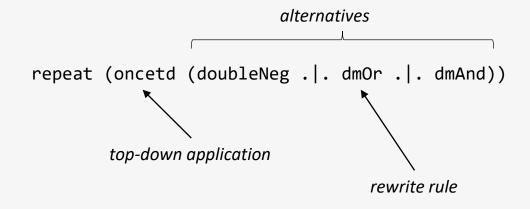
Rewrite strategies for automated feedback

- Domain-specific language for specifying problem-solving procedures:
 - describe sequences of rule applications that solve a particular task
 - are formalized by a trace-based semantics (CSP)
 - allow new composition operators (interleaving, topological sorts)
- Problem-solving procedures are used for feedback generation:
 - recognizing the solution strategy
 - detecting detours
 - suggesting subgoals
 - providing next-step hints
 - providing worked-out examples

Example

Goal: rewrite proposition into negation normal form (NNF)

 $\neg ((p \lor q) \land \neg (p \land r))$ $\Leftrightarrow De Morgan$ $\neg (p \lor q) \lor \neg \neg (p \land r)$ $\Leftrightarrow De Morgan$ $(\neg p \land \neg q) \lor \neg \neg (p \land r)$ $\Leftrightarrow Double Neg$ $(\neg p \land \neg q) \lor (p \land r)$ Rewrite strategy for NNF:



Strategy combinators

p.*.q	sequence: first p, then q
succeed	always succeeds
p. .q	choice: p or q
p ./. q	preference: p is preferred over q
p > q	left-biased choice: p or else q
fix	fixed-point combinator

Derived combinators:

try s = s |> succeed

```
repeat s = try (s .*. repeat s)
```

Finite representation with explicit recursion:

repeat s = fix \$ \x ->
 try (s .*. x)

Advantages:

- Extract rules from strategy
- Customize strategy
- Document/visualise strategy



Light-weight rewrite rules

Proposition logic

data Logic = Logic :&&: Logic	conjunction
Logic : : Logic	disjunction
Not Logic	negation
Var String	variable

Representation can be more complex, with nested and parameterised datatypes, e.g.:

3x + 9 = 0 V x = 1

1	p ⊢ p	Assumption
2	$p \to q \vdash p \to q$	Assumption
998	$p,p \to q,q \to r \vdash r$	
999	$p \to q,q \to r \vdash p \to r$	Deduction 998
1000	$q \to r \vdash (p \to q) \to (p \to r)$	Deduction 999

$\neg \neg \varphi \iff \varphi$ $\neg (\varphi \land \psi) \iff \neg \varphi \lor \neg \psi$ $\neg (\varphi \lor \psi) \iff \neg \varphi \land \neg \psi$

Rewrite rules

doubleNeg = rewriteRule "doubleNeg" \$

\phi -> Not (Not phi) :~> phi

dmAnd = rewriteRule "dmAnd" \$
 \phi psi -> Not (phi :&&: psi) :~> Not phi :||: Not psi
 //
meta-variables are
introduced by lambdas left-hand side right-hand side

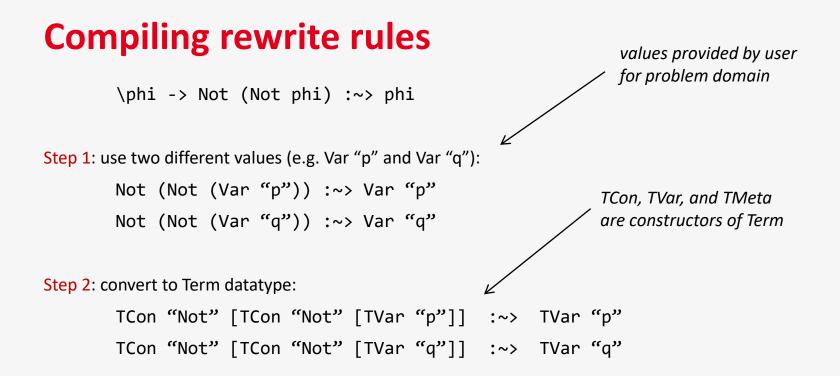
How to use such rewrite rules?

Embedding-projection pair

Approach: conversion from/to a generic Term datatype with support for meta-variables

- toTerm :: Logic -> Term
- fromTerm :: Term -> Maybe Logic
- From/to should be inverse functions (intuitively)
- Conversion allows generic functions, such as unification and zippers
- Pair can be derived automatically from the datatype definition

Note: more powerful generic programming libraries exist that can guarantee more type safety, with less overhead



Step 3: find meta-variables by comparing left-hand sides and right-hand sides

TCon "Not" [TCon "Not" [TMeta 0]] :~> TMeta 0

Applying rewrite rules

Rewrite rule:

Not (Not (Var "p" :&&: Var "r"))

TCon "Not" [TCon "Not" [TMeta 0]]
:~> TMeta 0

Step 1: convert to Term datatype:

```
TCon "Not" [TCon "Not" [TCon "And" [TVar "p", TVar "q"]]]
```

Step 2: match with rule's left-hand side:

0 = TCon "And" [TVar "p", TVar "q"]

Step 3: substitute in rule's right-hand side:

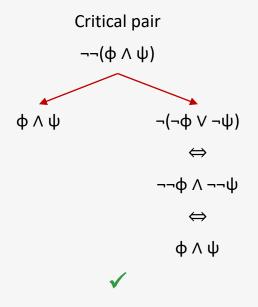
```
TCon "And" [TVar "p", TVar "q"]
```

Step 4: convert back to Logic:

Var "p" :&&: Var "r"

Knuth-Bendix completion

Use case for explicit representation: search for missing rewrite rules (and reach confluence)



$\neg \neg \varphi$	\Leftrightarrow	φ	
$\neg(\phi \land \psi)$	\Leftrightarrow	$\neg \varphi \lor \neg \psi$	

Missing rule: $\neg(\phi \lor \psi) \iff \neg \phi \land \neg \psi$

Light-weight rewrite rules

Advantages of explicit representation:

- Knuth-Bendix completion (analysis)
- AC-rewriting
- Rule inversion
- Automated testing
- Documentation (pretty-printing)

Summary for rewrite rules:

- Simplify construction (light-weight embedding)
- Explicit representation (for better feedback)
- Many problem domains

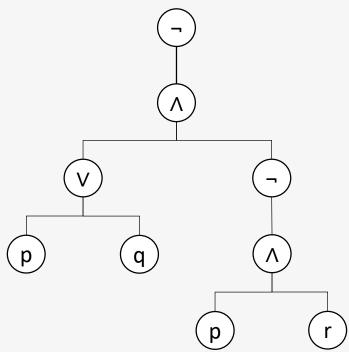


Generic traversals



Tree representation

 $\neg((p \lor q) \land \neg(p \land r))$

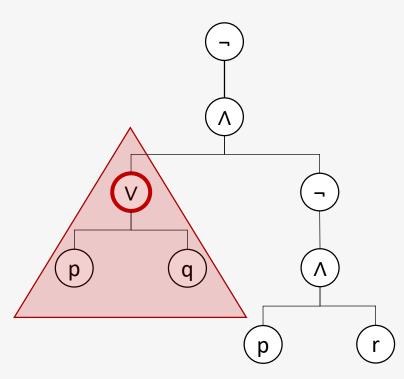




Point of focus

 $\neg((\underline{p \lor q}) \land \neg(p \land r))$

- Implemented as a so-called zipper over the generic Term datatype
- Stored in a Context



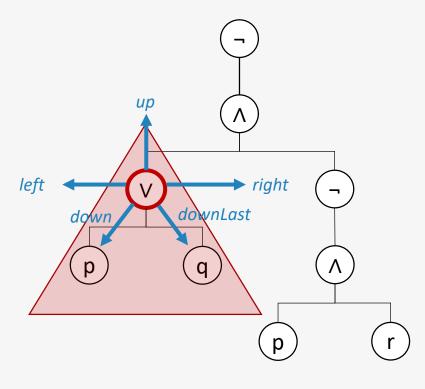
Navigation

Five navigational actions:

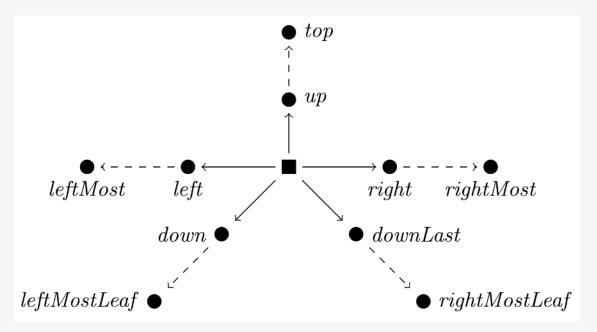
- up
- left
- right
- down
- downLast
- Actions may fail
- Many useful laws, e.g.:

left \circ right \approx id

 $up \circ down \approx id$



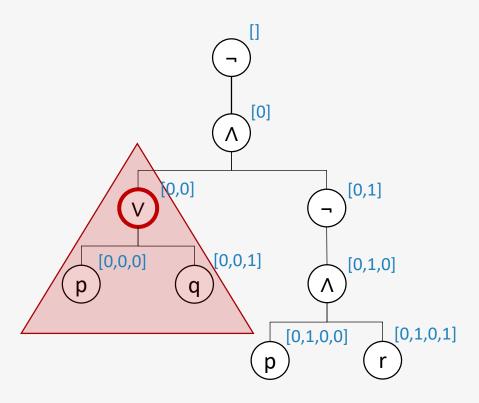
Navigation (extended)



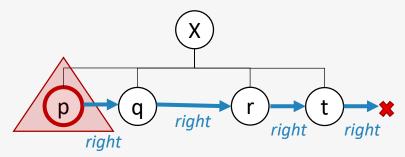
From: <u>Traversals with Class</u>. In Jurriaan Hage and Atze Dijkstra, editors, *Een Lawine* van Ontwortelde Bomen: Liber Amicorum voor Doaitse Swierstra, pages 62-75. 2013.

Position

- Zippers keep a position for the point of focus
- Position information is useful for generating feedback



Horizontal visits



visitOne s = fix $x \rightarrow s$. (right .*. x)

visitFirst s = fix $x \rightarrow s > (right .*. x)$

visitAll s = fix $(x \rightarrow s \cdot (not right \rightarrow (right \cdot x)))$

Approach: define traversals as (normal) strategy combinators

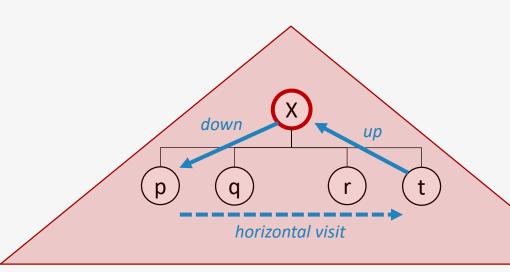
Idea: also parameterize "next" function to also support right-to-left visits



One-layer visits

layer s = down .*. s .*. up

layerOne s = layer (visitOne s)



Traversals

somewhere s = fix $x \rightarrow s$. |. layerOne x

oncetd s = fix \$ \x -> s |> layerOne x -- top down

oncebu s = fix \$ \x -> layerOne x |> s -- bottom up

Also: full traversals, spine traversals, innermost, outermost, etc.

Example trace

 $\neg((p \lor q) \land \neg(p \land r))$ $\Leftrightarrow De Morgan$ $\neg(p \lor q) \lor \neg\neg(p \land r)$ $\Leftrightarrow De Morgan$ $(\neg p \land \neg q) \lor \neg\neg(p \land r)$ $\Leftrightarrow Double Neg$ $(\neg p \land \neg q) \lor (p \land r)$

De Morgan	at []	
down		
De Morgan	at [0]	
up		
down		
right		
Double Neg	at [1]	
up		

Corresponding trace:

Summary for traversals:

- Simplify construction (traversals are first-class strategy combinators)
- Explicit representation (for better feedback)
- Many problem domains



Conclusion

Trends and challenges

- Authoring intelligent tutoring system
 - Literature reports 200-300 authoring hours for 1 hour of instruction
 - We believe software technology can help
- Data-driven intelligent tutoring system
 - Use AI techniques to generate feedback from collected data
 - Raises questions about the role of expert domain knowledge
- Further adaptation and personalization
 - Models for mastery learning (e.g. Bayesian knowledge tracing)
- Designing tools for less-structured problem domains
 - For example, domains of software design and learning languages

Conclusion

- Rewrite strategies are used for feedback generation
- Rewrite rules can be embedded by using datatype-generic programming techniques
- Generic traversals can be composed from navigational actions and strategy combinators
- The presented approach can be applied to a wide range of problem domains

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Project website: http://ideas.cs.uu.nl/

Related publications

- Thomas van Noort, Alexey Rodriguez, Stefan Holdermans, Johan Jeuring, and Bastiaan Heeren. <u>A Lightweight Approach to</u> <u>Datatype-generic Rewriting</u>. In *Proceedings of the ACM SIGPLAN Workshop on Generic Programming*, WGP '08, pages 13-24, 2008. ACM.
- Johan Jeuring, José Pedro Magalhães, and Bastiaan Heeren. <u>Generic Programming for Domain Reasoners</u>. In Zoltán Horváth, Viktória Zsók, Peter Achten, and Pieter W. M. Koopman, editors, *Proceedings of the Tenth Symposium on Trends in Functional Programming, TFP 2009*, pages 113-128, 2009. Intellect.
- Thomas van Noort, Alexey Rodriguez yakushev, Stefan Holdermans, Johan Jeuring, Bastiaan Heeren, and José Pedro Magalhães. <u>A Lightweight Approach to Datatype-generic Rewriting</u>. J. Funct. Program., 20(3-4):375-413, 2010.
- Bastiaan Heeren and Johan Jeuring. <u>Feedback services for stepwise exercises</u>. Science of Computer Programming, 88:110-129, 2014. Software Development Concerns in the e-Learning Domain.
- Josje Lodder, Bastiaan Heeren and Johan Jeuring. <u>A Domain Reasoner for Propositional Logic</u>. Journal of Universal Computer Science, 22(8):1097-1122, 2016.
- Bastiaan Heeren and Johan Jeuring. <u>An Extensible Domain-Specific Language for Describing Problem-Solving Procedures</u>. In Elisabeth André, Ryan Baker, Xiangen Hu, Ma. Mercedes T. Rodrigo, and Benedict du Boulay, editors, *Proceedings of Artificial Intelligence in Education: 18th International Conference, AIED 2017*, pages 77-89, 2017. Springer International Publishing.
- Josje Lodder, Bastiaan Heeren, Johan Jeuring, and Wendy Neijenhuis. <u>Generation and Use of Hints and Feedback in a</u> <u>Hilbert-Style Axiomatic Proof Tutor</u>. *International Journal of Artificial Intelligence in Education*, 2020.