



Learning from Expert's proof

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The project

- Newcastle University
 - Cliff Jones, Leo Freitas, and Andrius Velykis
- University of Edinburgh
 - Alan Bundy, Gudmund Grov, and Yuhui Lin
- DEPLOY and other industrial partners
 - Bosch, Siemens, *etc.*
 - Praxis, US NRL
- 4 years – started April 2010

Objectives

- Reduce verification bottlenecks
 - Avoid rework on (structurally) similar proofs
 - Not aiming at general mathematical problems
 - Focus on POs from development in industry
- Use machine learning for proof mining
 - Lemma suggestion and generalization
 - Inference of induction principles
 - Reasoned proof critics and plans
- Domain knowledge acquisition
 - Investigation of failed proof attempts
 - Proof family identification

Context and Proof Support

- Source of proof obligations
 - Top-down : C by C, posit + prove (gluing invariants)
 - Bottom-up : post-facto verification (code-styles, static analysis)
 - Both : abstraction guided documentation
- Decompose verification effort with key abstractions
 - Profit from structural / conceptual similarities
- Alan's learning scenarios
 - Proof chunking (*i.e.*, term taxonomy / hierarchy)
 - “n-proofs” (*i.e.*, proof versions: declarative, programmed, *etc.*)
 - Anti-unification (*i.e.*, find least-general common generalization)
 - Cut-rules (*i.e.*, lemmas suggestion and inference)
 - Meta-tagging (*i.e.*, defined function, constructor, type, *etc.*)

AI4FM Approach

- Hypothesis: learning from proof processes of an expert
 - On a specific class of problems
 - Lemma suggestion and problem decomposition
 - Tool based learning from proof failures
- Rationale: proof influences modeling decisions
 - But through counter-examples / something new
 - Avoid model fiddling (just) to increase levels of proof automation
- Machine learning techniques envisaged
 - Proof planning and critics (*e.g.*, IsaPlanner)
 - Top-down formal development (*e.g.*, VDM, Z, B)
 - Bottom-up code-level verification (*e.g.*, Boogie, Spark)
- Find “toy-problems”
 - Like lab mice in pharmaceutical research
 - Use strategy from toy to solve original problem

AI4FM Approach

- **Proof expert role will still be key**
- Create strategy language
 - Beyond simply sequential tactical language
 - Take into account taxonomy of terms and their use
 - Specification method independent as much as possible
 - We are currently investigating AI algorithms/ideas to this effect
- Investigate industry-relevant proof data
- Proof data under consideration now / near future
 - Praxis' Tokeneer ID station (*e.g.*, Ada, Spark)
 - Bosh cruise control (*e.g.*, Event-B, Rodin, Pro-B)
 - NRL Xenon High-Assurance Hypervisor (*e.g.*, CodeSonar, C++)
- Take inspiration from various sources
 - Event-B: various layers of abstraction and refinement
 - Boogie: targeted (to Z3/SMT) abstract intermediate language (ATP-like)
 - ACL2/Z-Eves: guess and prove, lemma generalization, and toy problems

Finally

- We are at the beginning
 - More info at <http://www.ai4fm.org>
 - AI4FM mailing list is open: ai4fm-info@jiscmail.ac.uk
 - We have meetings planned for sharing results / ideas
- We would love to hear your feedback / criticisms
 - What do you like about the idea?
 - What would you do differently?
 - Goal: reduce residual / repetitive POs

Questions?