Isabelle/RL Project Proposal: Reinforcement learning on the Isabelle proof assistant

Jonathan Julián Huerta y Munive <u>huertjon@cvut.cz</u> Czech Technical University in Prague September, 2024





Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union. The European Union cannot be held responsible for them.

Summary

The AITP conference is the forum for discussing how to get to our inevitable future of largescale semantic processing and strong computer assistance of mathematics and science

Discussion points:

- Isabelle is one of the most developed interactive theorem provers (ITPs)
- Machine learning (ML) methods have been successfully applied to other ITPs
- Many users are still not benefiting from tool integrations of ML and ITPs
- Reinforcement learning (RL) approaches with conjecturing have not been sufficiently attempted in these integrations
- There is still much to learn with respect to ML and ITPs
- We are currently working towards filling these gaps
- This is an open invitation to contribute to this project

Overview of the Isabelle proof assistant

An Isabelle apply-style proof

Tactics for backward reasoning from goal to hypothesis (tactics inputs: terms and premises)



Hammers

Calls to automated theorem provers (ATPs) to reduce the number of premises needed to automatically prove the current proof obligation

lemma

```
fixes P :: "('a::finite) \Rightarrow nat \Rightarrow bool"
                                                                                              proof (prove)
assumes "∀i. ∃N::nat. ∀n > N. P i n"
                                                                                              goal (1 subgoal):
shows "\exists N. \forall i. \forall n > N. P i n"-
                                                                                               1. \exists N. \forall i n. N \leq n \longrightarrow P i n
apply (rule tac x="Max {Inf {N. \forall n \ge N. P i n} |i::'a. i \in UNIV}" in exI)
apply clarify
subgoal for i n
  using wellorder Inf lemma[of "\lambdan. P i n"] assms
     Max ge[of "{Inf {N. \forall n \ge N. P i n} |i:::'a. i \in UNIV}", OF finite image]
  apply -
  apply (erule tac x=i in allE)
  apply (subgoal tac "\forall n \ge Inf \{N. \forall n \ge N. P i n\}. P i n")
  sledgehammer
Sledgehammering...
vampire found a proof...
spass found a proof...
vampire: Try this: apply (metis (mono tags, lifting) UNIV I dual order.trans mem Collect eg) (90 ms)
spass: Try this: apply (metis (mono tags, lifting) iso tuple UNIV I mem Collect eq order trans) (40 ms)
```

Done

An Isabelle apply-style proof

Iterate based on feedback and finish the proof



An Isabelle Isar-style Proof



An Isabelle Isar-style Proof



Previous machine learning approaches on interactive theorem provers

Previous ML approaches on ITPs: TacticToe

Uses k-nearest neighbours (kNN) and Monte-Carlo tree search (MCTS) to predict the next proof-step (proves 66.4% of HOL4 standard library)

lemma



Previous ML approaches on ITPs: "Draft, Sketch, Prove" vs Magnushammer



Combined approach: Proof-rate of 71% on publicly available benchmark of 183K Isabelle proofs from the AFP

Other previous ML approaches on ITPs

- TacticZero for HOL4: It is an RL approach using NNs. It achieves ~49% success on a test set of 268 problems.
- **Tactician for Coq:** It is similar to TacticToe but extends the approach with a graph neural network (GNN) and compares it with the kNN. It also uses an LLM that is not context aware. Intersection between kNN and GNN is thin. GNN almost fully covers CoqHammer's results while proving some that CoqHammer does not prove.
- HOL-list environment for HOL Light: It is a light-weight tool for testing different approaches for tactic and premise selection. It has tested GNNs and an RL algorithm. The RL approach solves 56.3% of a set of 3225 theorems of the HOL Light's mathematical library.
- Abduction prover, smart induct, and SeLFiE for Isabelle: Combinations of proof strategy languages and conjecturing heuristics specialised to induction proofs, where hammers are weaker.
- LLMs for Lean: Many recent articles on using LLMs for autoformalization, premise selection, tactic suggestion, and proof search.

Questions

Questions after reading the literature

- TacticToe for HOL4 TacticZero for HOL4 \bullet
- Tactician for Coq \bullet
- Abduction prover for Isabelle
- Magnushammer for Isabelle
- LLMs for Lean (at least three research groups)

Observations:

- Isabelle is the only one that does not have an ML-focused integration
- Only the Tactician and Abduction prover seem to conjecture
- Only the Tactician offers comparisons of different ML models AND comparisons to hammers

- How do the kNN and the NN-based RL approaches compare?
- Does context-awareness improve the LLM results?
- HOL-list environment for HOL Light \rightarrow How does it compare with LLM-approaches?
 - How does it compare with ML approaches?
 - Can it be integrated as a tool into Isabelle? Does context-awareness improve the results?
 - Do any of these methods perform well on new problems?

Progress

Project's current outputs



Data generation example output as JSON attributes

'term': '(dvd.dvd times (numeral.numeral one plus (num.Bit0 num.One)) x
\\<Longrightarrow> dvd.dvd times (numeral.numeral one plus (num.Bit0 num.One)) (power.power one times x 2))
\\<Longrightarrow> (dvd.dvd times (numeral.numeral one plus (num.Bit0 num.One)) x
\\<Longrightarrow> dvd.dvd times (numeral.numeral one plus (num.Bit0 num.One)) (power.power one times x 2))",

```
"thms": [
    {"thm": {"name": "Pure.protectI", "term": "PROP ?A \\<Longrightarrow> (PROP ?A)"}},
    {"thm": {"name": "Pure.protectD", "term": "(PROP ?A) \\<Longrightarrow> PROP ?A"}},
    {"thm": {"name": "HOL.allE", "term": "\\<forall>x. ?P x \\<Longrightarrow> (?P ?x \\<Longrightarrow> ?R) \\<Longrightarrow> ?R"}},
    {"thm": {"name": "HOL.notE", "term": "\\<not> ?P \\<Longrightarrow> ?P \\<Longrightarrow> ?R"}},
    {"thm": {"name": "HOL.allI", "term": "\\<not> ?P x) \\<Longrightarrow> \\<forall>x. ?P x"}},
```

```
'keywords": [
 {"name": "\\<proof>"},
 {"name": "sorry"},
 {"name": "by"},
 {"name": ".."},
 {"name": "."},
 {"name": "unfolding"},
 {"name": "including"},
 {"name": "using"},
 {"name": "prefer"},
 {"name": "defer"},
 {"name": "apply"},
 {"name": "back"},
 {"name": "done"},
 {"name": "oops"},
 {"name": "proof"},
 {"name": "subgoal"}
```

- More lemmas found per .thy file than in previous Isabellefocused approaches
- More context-aware data generation: syntax, keywords, methods and theorems are userextensible, and the datageneration accounts for that

methods"	ſ
methous :	L
{"name":	"HOL.split"},
{"name":	"HOL.force"},
{"name":	"HOL.cases"},
{"name":	"HOL.blast"},

{"name":	"HOL.auto"},
{"name":	"SMT.smt"},
{"name":	"SAT.sat"},
{"name":	"Pure"}
1	

Next steps

Next steps

- Currently working on retrieving Data.T in Scala (we can already retrieve Isabelle/ML terms t:term, and we have already accessed Isabelle functions in Scala from Python)
- Reproduce premise selection, tactic suggestion and proof-search results
- Use these results for generating "conjecturing" data
- Combine these approaches into various RL algorithms, similar to those used for playing Go or StarCraft, and compare them
- Create an Isabelle tool suggesting "next steps" and/or "complete (legible) proofs" for a given proof obligation

Conclusion

There are **many things to learn** from integrations of ITPs and machine learning methods. This is an **open invitation** to Isabelle or machine learning experts to collaborate on the implementation of the project or to provide constructive feedback that accelerates its completion.

Thanks!

- Isabelle is one of the most developed interactive theorem provers (ITPs)
- Machine learning (ML) methods have been successfully applied to other ITPs
- Many users are still not benefiting from tool integrations of ML and ITPs
- Reinforcement learning (RL) approaches with conjecturing have not been sufficiently attempted in these integrations
- There is still much to learn with respect to ML and ITPs
- We are currently working towards filling these gaps
- This is an open invitation to contribute to this project

Jonathan Julián Huerta y Munive <u>huertjon@cvut.cz</u> Czech Technical University in Prague September, 2024