## Learning to Prove with Tactics

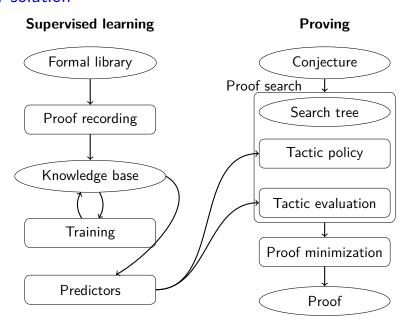
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March 28, 2018

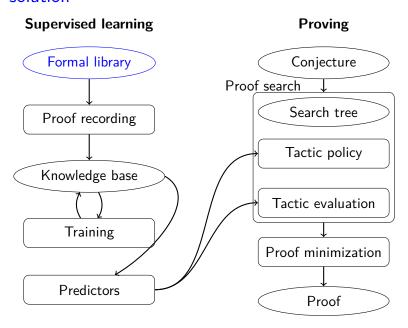
## **Problem**

Can we formally prove mathematical formulas automatically?

### Our solution



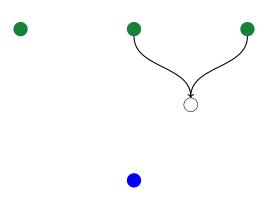
### Our solution



# Formal library: reasoning with inference rules

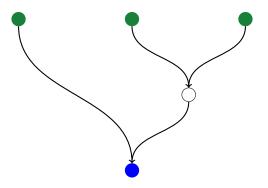
- axiom
- conjecture
- ightarrow rule

# Formal library: reasoning with inference rules



- axiom
- conjecture
- $\to \mathsf{rule}$
- Olemma

# Formal library: reasoning with inference rules

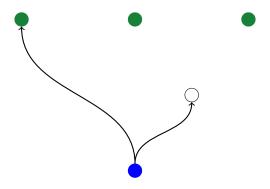


- axiom
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# Formal library: reasoning with tactics

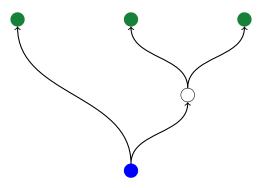
- axiom
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- $\bigcirc$  goal

# Formal library: reasoning with tactics



- axiom
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# Formal library: reasoning with tactics



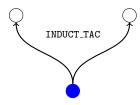
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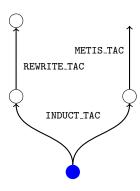
# Formal library: tactics

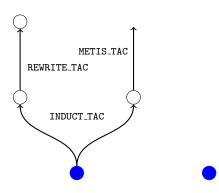
REWRITE\_TAC

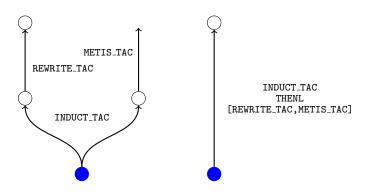
INDUCT\_TAC

METIS\_TAC



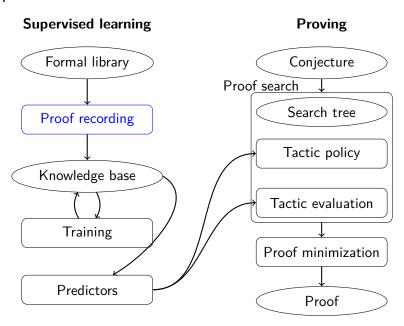






## Demo

## Plan



## Recording

### Original proof:

INDUCT\_TAC THENL [REWRITE\_TAC, METIS\_TAC]

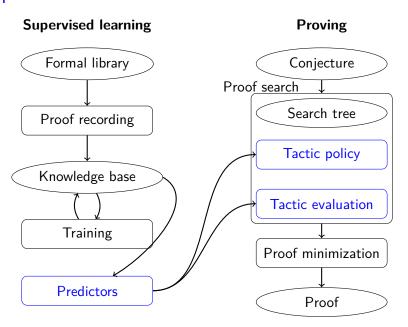
### Modified proof:

```
(R numLib.INDUCT_TAC) THENL
  [R boolLib.REWRITE_TAC, R metisLib.METIS_TAC]
```

#### Database of tactics:

```
R (f n) (f (SUC n)) \Rightarrow transitive R: INDUCT_TAC n * m \leq n * p \Rightarrow (n = 0) V m \leq p : REWRITE_TAC INJ f U(:num) s \Rightarrow INFINITE s : METIS_TAC ...
```

## Plan



## Prediction algorithm

### Algorithm:

Nearest neighbor weighted by TF-IDF heuristics

#### Effect:

Order goals from the database according to their distance to a target goal.

#### Remark:

This is algorithm performs premise selection. How do we adapt it to predict tactics?

## Policy: choosing a tactic

Database of tactics is a map from goals to tactics.

```
R (f n) (f (SUC n)) \Rightarrow transitive R: INDUCT_TAC n * m \leq n * p \Rightarrow (n = 0) V m \leq p : REWRITE_TAC INJ f U(:num) s \Rightarrow INFINITE s : METIS_TAC ...
```

An order on goals induces an order on tactics.

### New goal appearing during proof search:

```
LENGTH (MAP f 1) = LENGTH 1
```

### Policy for the new goal:

```
Rank Tactic Policy
1 REWRITE_TAC 0.5
2 METIS_TAC 0.25
...
4 INDUCT_TAC 0.0625
```

# Value function: provability of a list of goals

### Database of lists of goals:

- ▶ Positive examples: appears in human proofs.
- Negative examples: produced during TacticToe search but do not appear in the final proof.

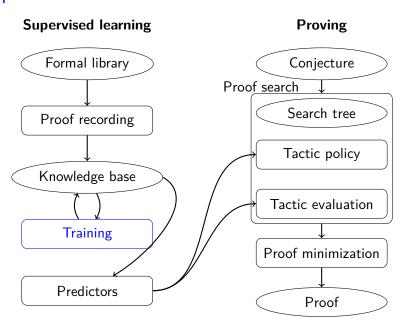
#### Value function:

Percentage of positives in the 10 closest lists of goals of a target list of goals.

#### Future work:

Estimate the number of steps needed to prove a list of goals.

## Plan



# **Training**

Improve recorded data to create better predictions during search.

# Training: orthogonalization

Issue: Many tactics are doing the same job on a goal g.

Solution: Competition for g where the most popular tactic wins.

# Training: orthogonalization

### Recorded goal-tactic pair:

```
LENGTH (MAP f l) = LENGTH l: INDUCT_TAC
```

### Competition:

	Progress	Coverage
INDUCT_TAC	Yes	136
REWRITE_TAC	No	2567
METIS_TAC	Yes	694

#### Added to the database:

```
LENGTH (MAP f 1) = LENGTH 1: METIS_TAC
```

Result: 6 % improvement.

## Training: abstraction

Issue: Some theorems are never used inside tactics.

Solution: Abstract all lists of theorems in a tactic and instantiate them depending on the target goal.

## Training: abstraction

### Abstraction algorithm:

Original : REWRITE\_TAC [T1,T2]

Abstraction : REWRITE\_TAC X

Instantiation: REWRITE\_TAC [T67, T1, T43, ...]

Question: Dow we keep the original or the abstraction?

Answer: Let them compete during orthogonalization.

## Training: preselection

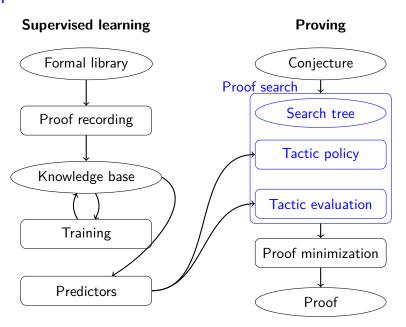
Issue: Predictions are too slow during proof search.

Solution: Preselect 1000 suitable tactics using dependencies.

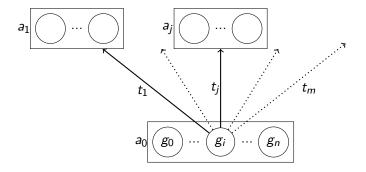
Dependency: Appear in the same proof.

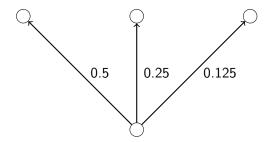
Result: 15% improvement.

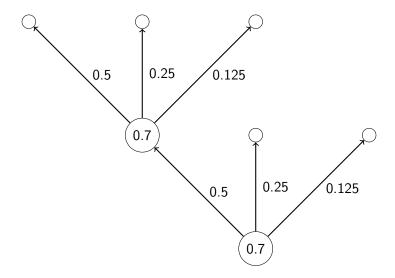
## Plan

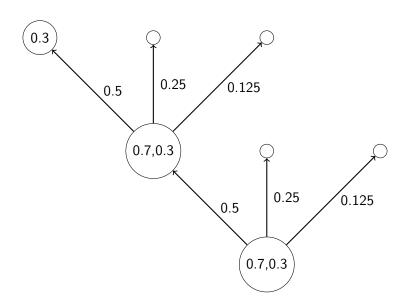


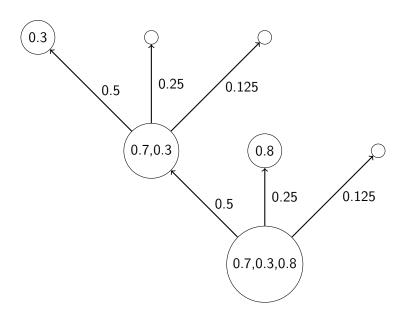
# Proof search: search tree









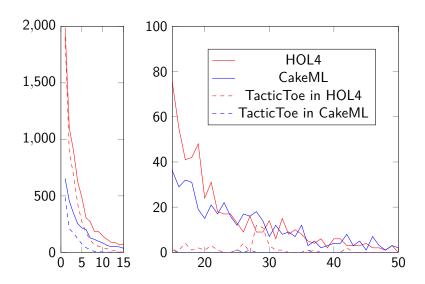


## Re-proving: results

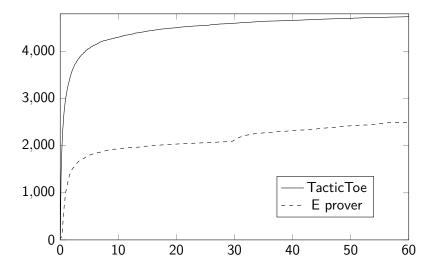
Evaluation is "fair". (not totally if you ask Freek)
Only previous proofs are available for training.
TacticToe does not call external provers.

	HOL4: 7164, 60s	CakeML: 3329, 15s
E prover TacticToe	2472 (34.5%) 4760 (66.4%)	1161 (34.9%)
Total	4946 (69.0%)	

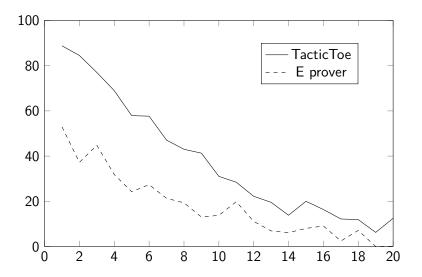
# Re-proving: proofs of size x



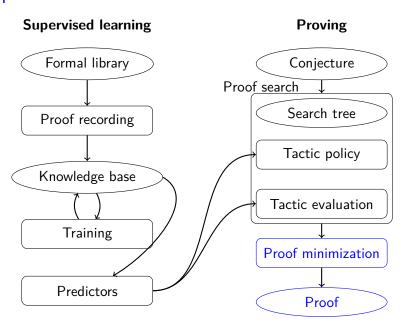
# Re-proving: HOL4 proofs found in less than x seconds



# Re-proving: percentage of solved HOL4 proof of size x



## Plan



### Minimization and embellishment

### Raw proof:

```
boolLib.REWRITE_TAC [DB.fetch "list" "EVERY_CONJ",...]

THEN

BasicProvers.Induct_on [HolKernel.QUOTE "l"]

THENL

[BasicProvers.SRW_TAC [] [],

simpLib.ASM_SIMP_TAC (BasicProvers.srw_ss ())

[boolLib.DISJ_IMP_THM, DB.fetch "list" "MAP",

DB.fetch "list" "CONS_11", boolLib.FORALL_AND_THM]]
```

### Processed proof:

```
Induct_on 'l' THENL
  [SRW_TAC [] [],
  ASM_SIMP_TAC (srw_ss ())
  [DISJ_IMP_THM, FORALL_AND_THM]]
```

### Conclusion

Summary: TacticToe learns from human proofs to solve new goals.

Advantages over ATPs (E prover) for ITP (HOL4) users:

- ▶ Includes domain specific automation found in the ITP.
- Generated proofs are human-level proofs.
- No translation or reconstruction needed.

### Future work

Enlarge the action space: parameter synthesis, sequence of tactics, forward proofs.

Train tactics by evaluating input/output pairs.

Conjecture intermediate lemmas.