

Using AI for Theorem Proving

Project Midway Presentation
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Theorem Proving in Lean

```
variables x y z w : ℕ
```

```
example (h1 : x = y) (h2 : y = z)  
(h3 : z = w) : x = w :=
```

```
begin
```

```
  rw [h1],
```

```
  rw [h2],
```

```
  assumption
```

```
end
```

Training Data

```
[{"goal": x y z w: nat  
  h1: x = y  
  h2: y = z  
  h3: z = w  
  |- x = w  
  "policy": 0.33  
  "tactic": "rw [h1]"},  
 {"goal": x y z w: nat  
  h1: x = y  
  h2: y = z  
  h3: z = w  
  |- y = w  
  "policy": 0.66  
  "tactic": "rw [h2]"},  
 {"goal": x y z w: nat  
  h1: x = y  
  h2: y = z  
  h3: z = w  
  |- z = w  
  "policy": 1.0  
  "tactic": "assumptions"}]
```

Tactic Generator Model

Model: ByT5-Small with 300M trainable parameters with 12 encoder and 4 decoder layers

Training objective:

```
Input: '''x y z w: nat
(goal)  h1: x = y
        h2: y = z
        h3: z = w
        |- x = w'''
```

```
Output: "rw [h1]"
        (tactic)
```

Policy Model

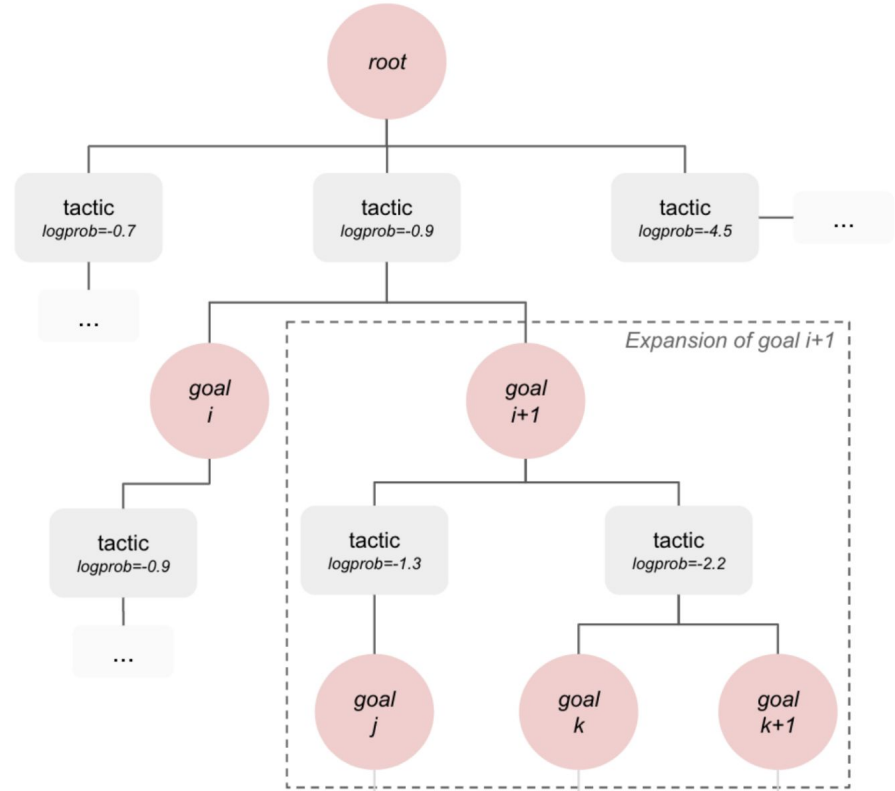
Model: RoBERTa-base with 125M trainable parameters

Training Objective:

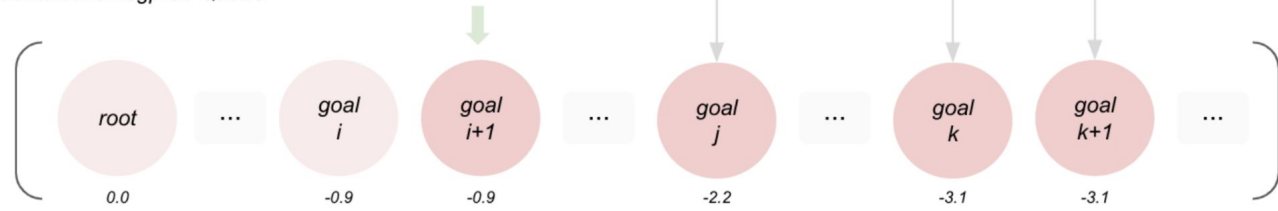
Input: '''x y z w: nat	Output: 0.33
(goal) h1: x = y	(policy)
h2: y = z	
h3: z = w	
- x = w'''	

Proof Search Tree

Proof Search Tree



Cumulative Logprob Queue



Results

Results (Pass@1) of proving theorems within a time limit of 10 minutes:

Method	Mathlib	MiniF2F
tactic log probability	55.0	28.6
tactic log probability + policy	55.4	29.0

Table 1. Table summarizing the effect of using tactic log probability & policy for guiding proof search.

Future Plan

- Integrate the policy model with Dynamic Sampling
- Limit on number of nodes to search instead of time limit
- Generate the data of success rate of goals and use it for training a new policy model

References:

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- Liu, Y., Ott, M., Goyal, N., Du, J., Joshi, M., Chen, D., Levy, O., Lewis, M., Zettlemoyer, L., and Stoyanov, V. Roberta: A robustly optimized bert pretraining approach. *arXiv preprint arXiv:1907.11692*, 2019
- Polu, S. and Sutskever, I. Generative language modeling for automated theorem proving. *arXiv preprint arXiv:2009.03393*, 2020

Thank You