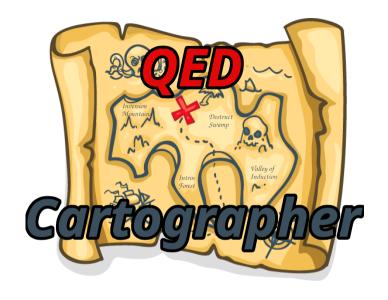


**Alex Sanchez-Stern**, Abhishek Varghese, Zhanna Kaufman, Dylan Zhang, Talia Ringer, Yuriy Brun



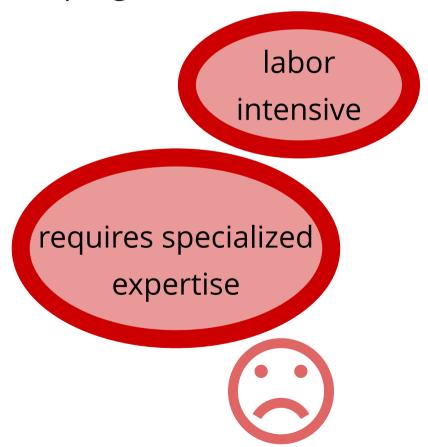
Automating Formal Verification with Reward-Free Reinforcement Learning

#### **Formal Verification**

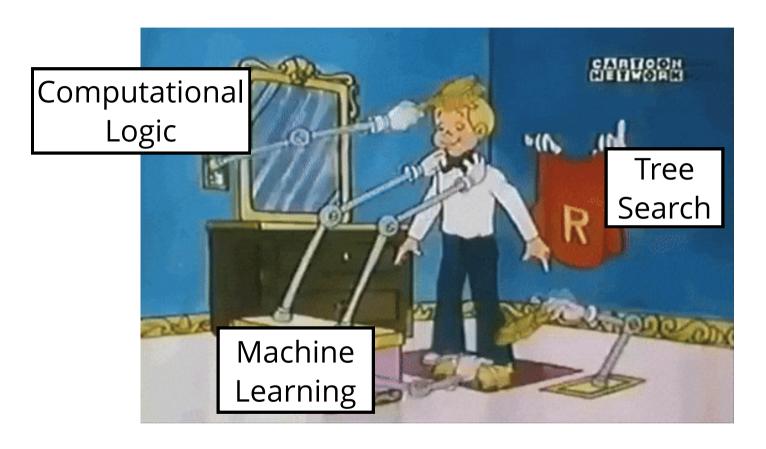
writing proofs about programs!

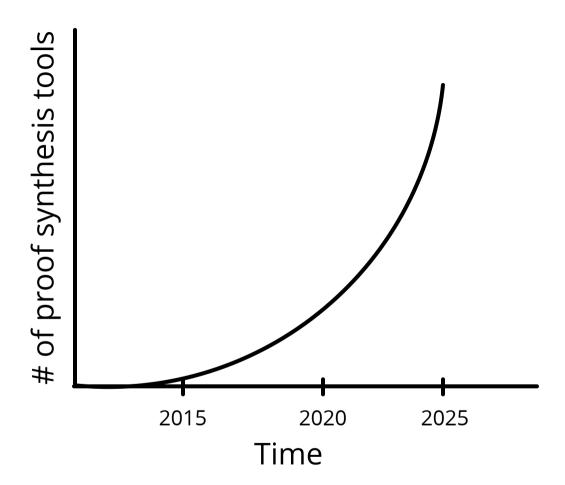


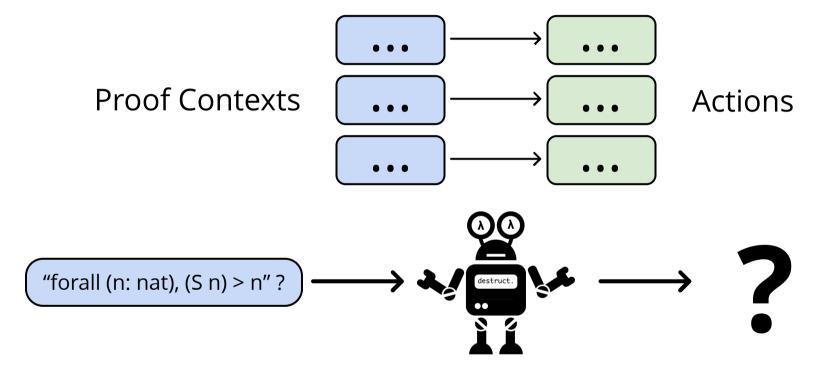


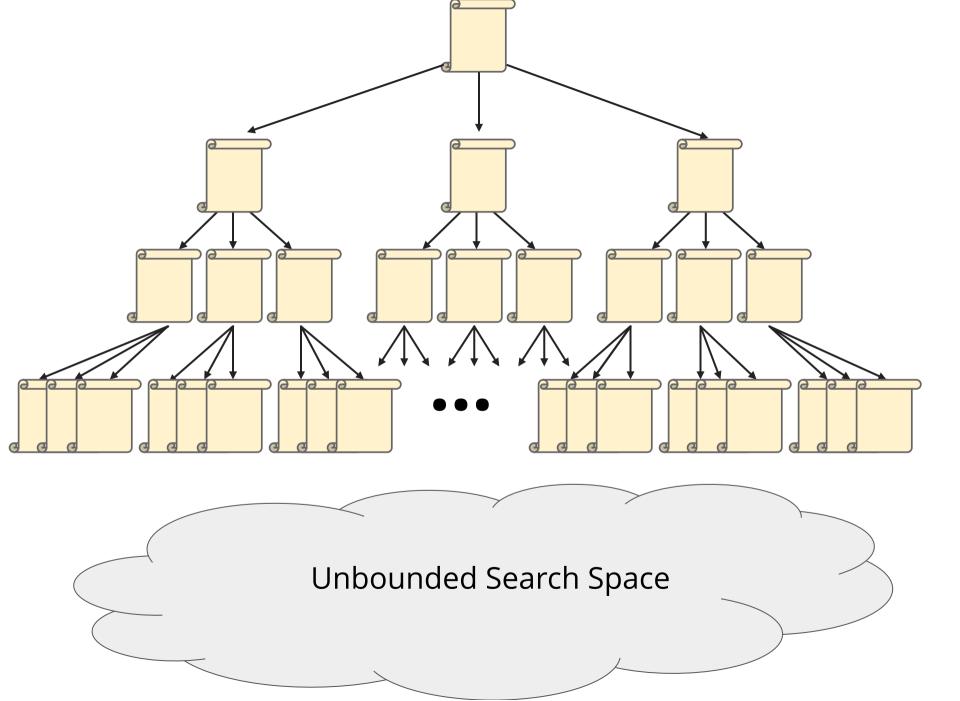


## **Automating** Formal Verification





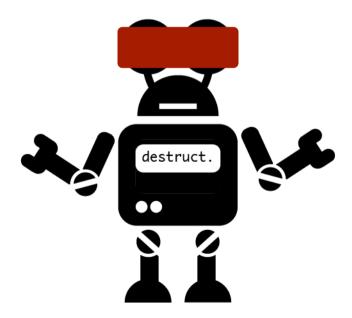




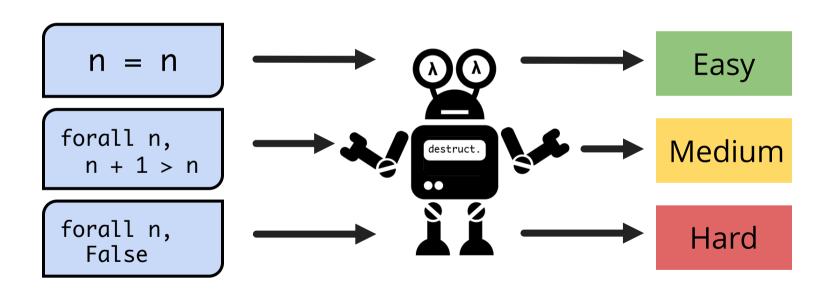
# There are some techniques that can help us prune the search tree

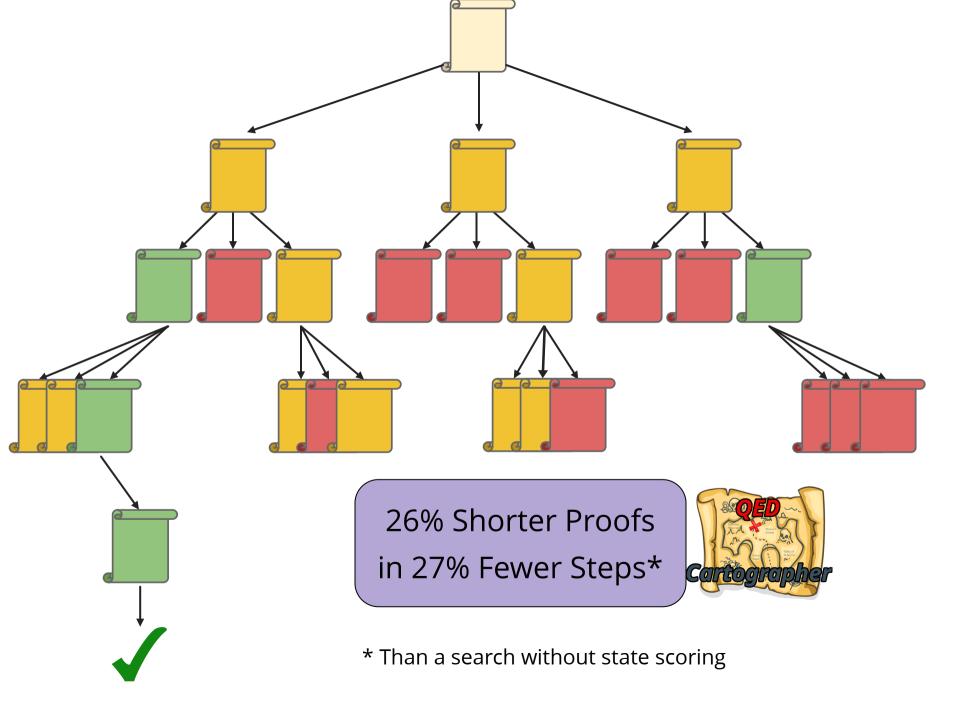


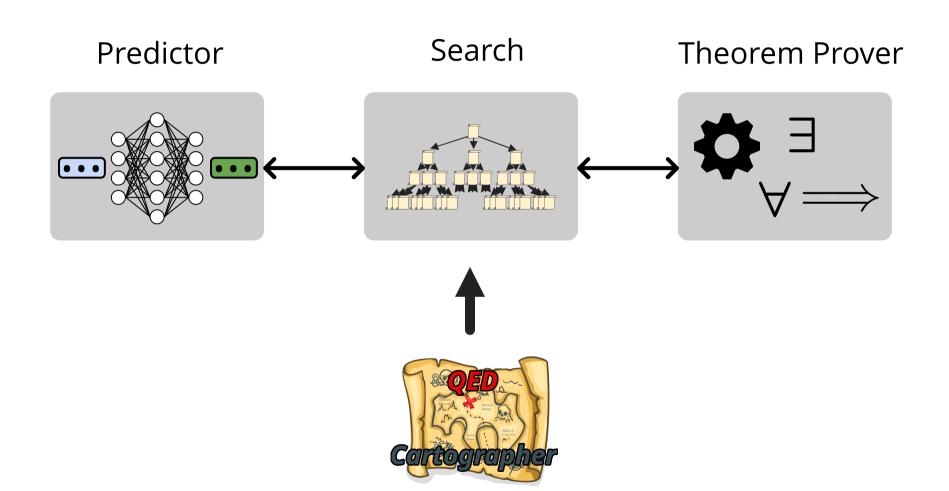
It's hard to explore when you don't know where you are!



# We can search more efficiently if we can evaluate proof states





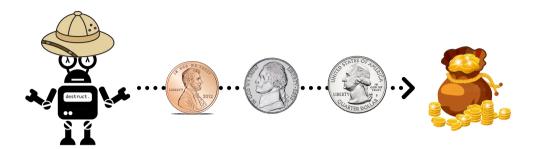


"Reward-free Reinforcement Learning"

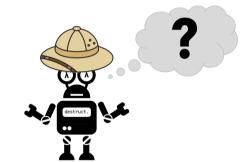
In particular, V-learning

## This Talk

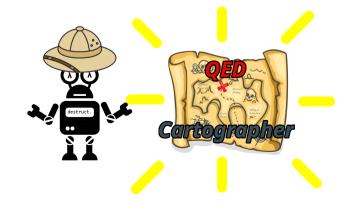
V-Learning



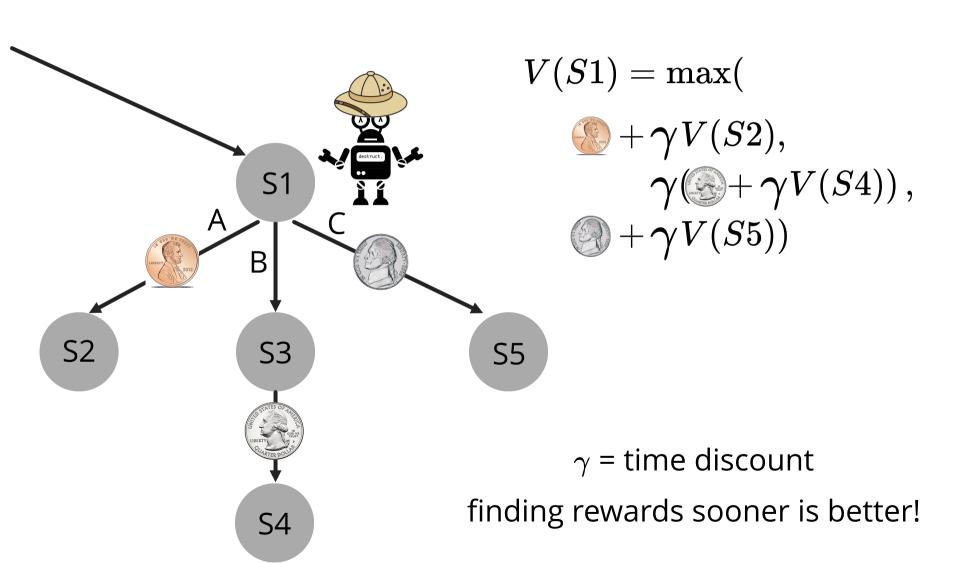
Limitations in Proofs



Adapting to Proofs



## Classic V-Learning

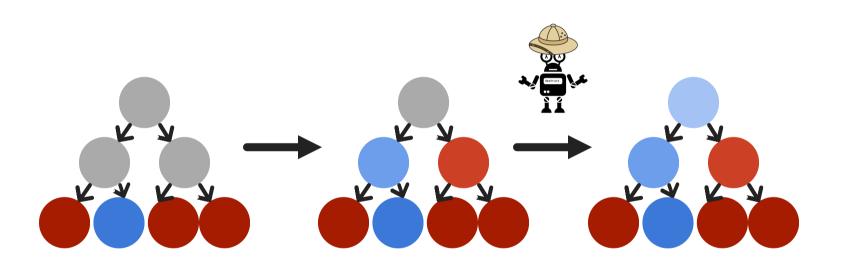


$$V(S1) = \max(\bigcirc + \gamma V(S2), \ \gamma \bigcirc + \gamma V(S4)), \bigcirc + \gamma V(S5))$$

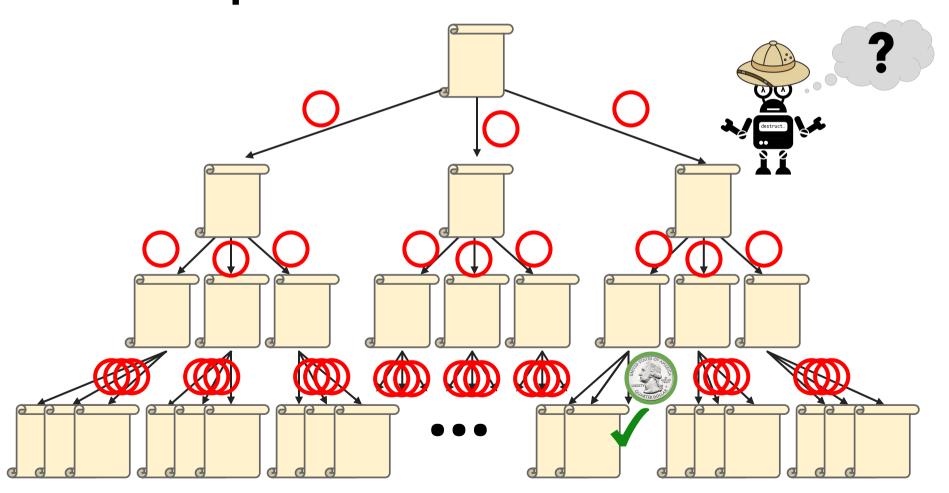


$$V(S) = \max_{a \in \operatorname{actions}(S)} \left( R(S, a) + \gamma V(\operatorname{next-state}(S, a)) \right)$$

# V-Learning in Practice: Iterative Updates

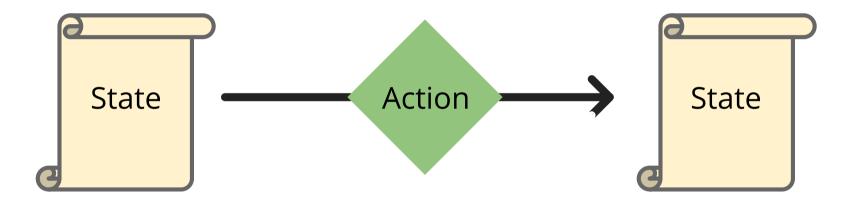


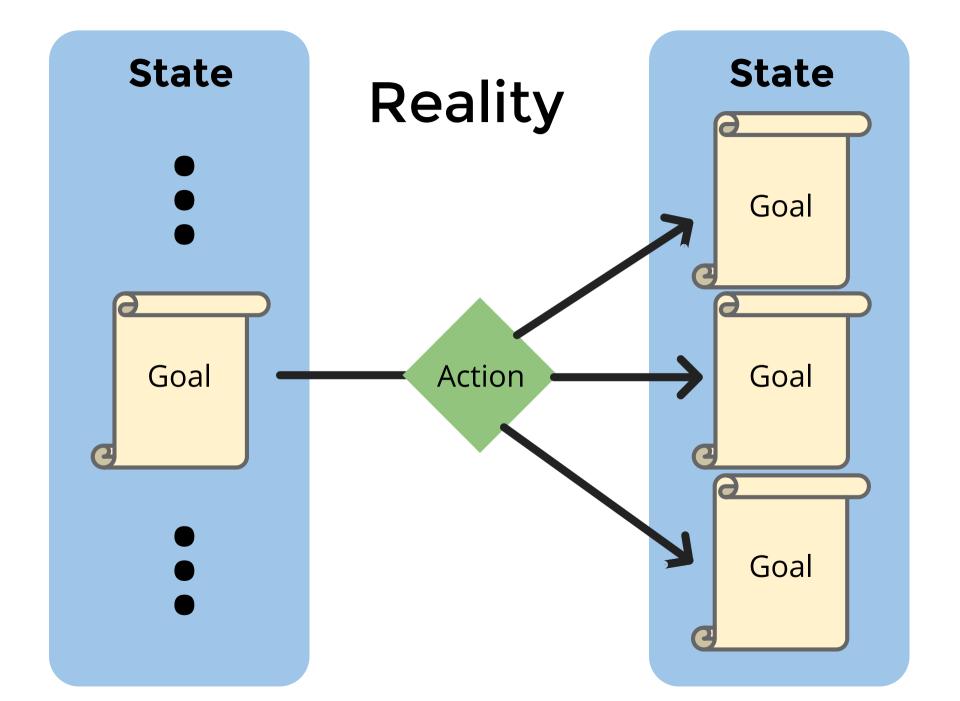
## V-Learning in Proofs: The Sparse Reward Problem

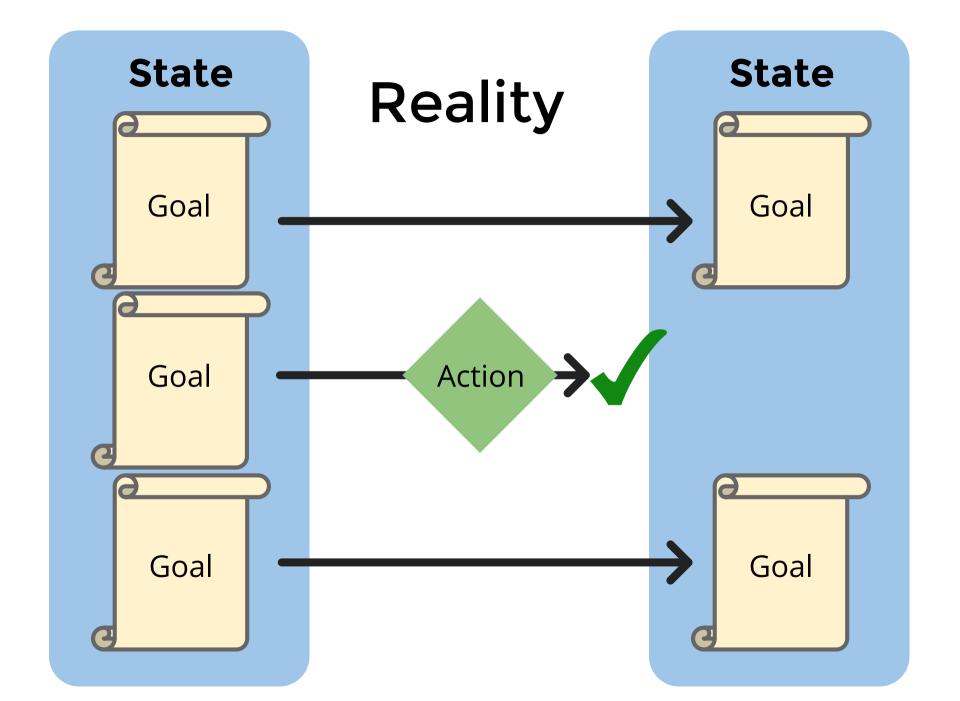


# Insight: Proofs have Useful Structure

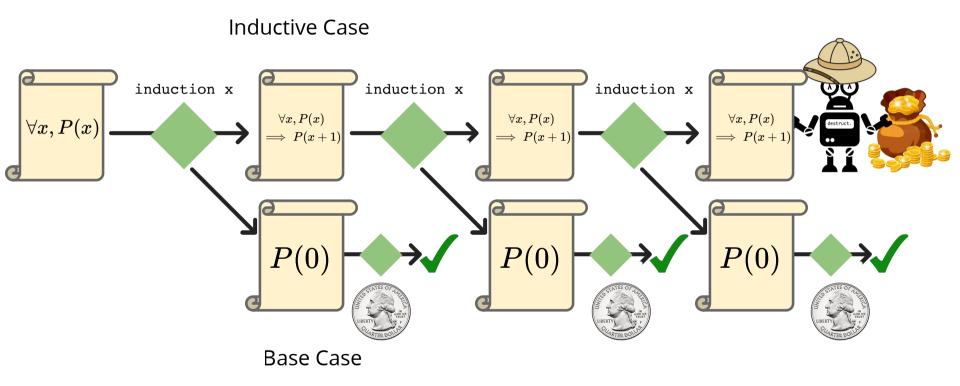
## **Abstraction**







#### Adding extra rewards can lead to bad behavior



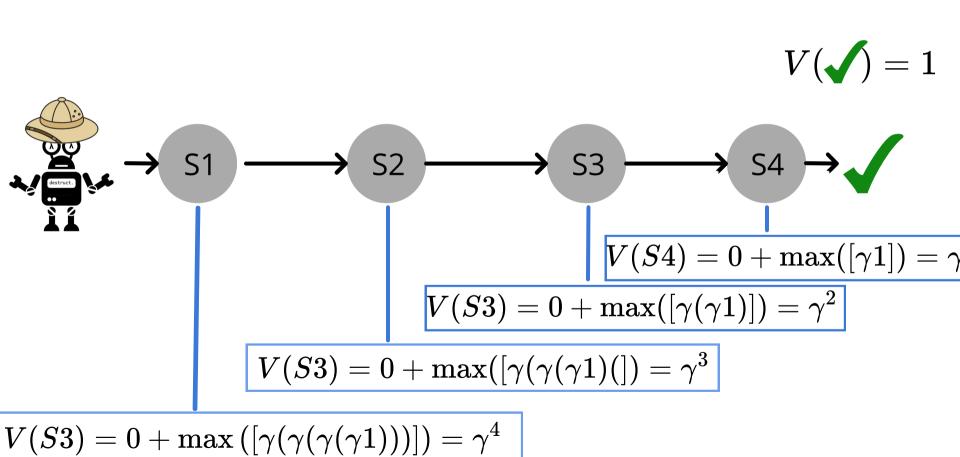
Reward-free doesn't have this problem!

## What We Need

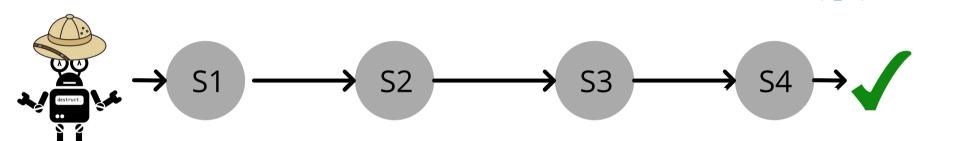
A new update equation that accounts for the branching structure of proofs

#### **Assumptions:**

• The state of a completed proof has value 1



## Assumption: The state of a completed proof has value 1



$$V(S) = \gamma^{
m (number\ of\ steps\ left)}$$

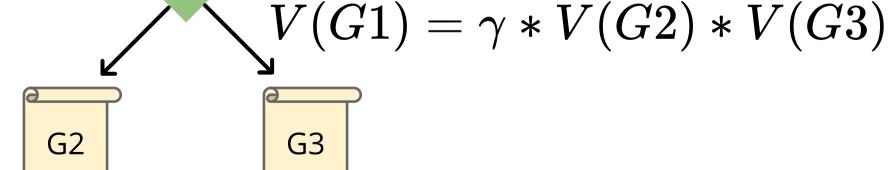


G1

m = Steps to complete proof from G2

n = Steps to complete proof from G3

Steps to complete proof from G1 = m + n + 1



$$V(G) = \overline{\max_{a \in \operatorname{actions}(G)}} \Big( \gamma \prod_{G' \in \operatorname{next-states}(G,a)} V(G') \Big)$$

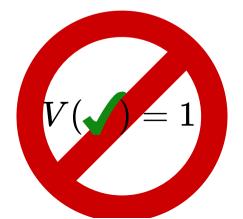
$$V(G) = \max_{a \in \operatorname{actions}(G)} \left( \gamma \prod_{G' \in \operatorname{next-states}(G,a)} V(G') 
ight)$$

$$V(G) = \max_{a \in \operatorname{actions}(G)} \left( \gamma \overline{\prod_{G' \in \operatorname{next-states}(G,a)} V(G')} 
ight)$$

$$V(G) = \max_{a \in \operatorname{actions}(G)} \left( \overline{\gamma} \prod_{G' \in \operatorname{next-states}(G,a)} V(G') 
ight)$$

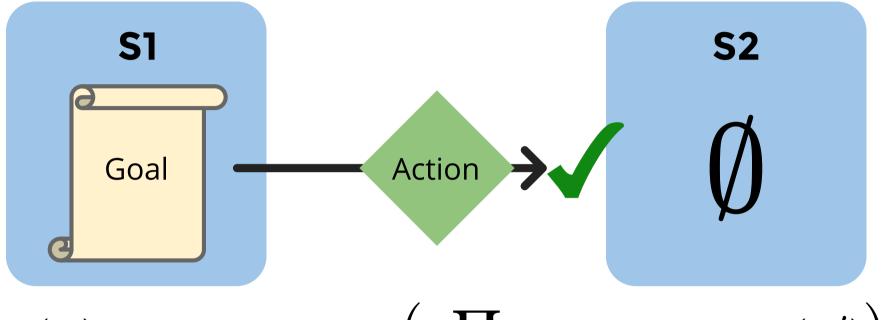
$$V(G) = \max_{a \in \operatorname{actions}(G)} \left( \gamma \prod_{G' \in \operatorname{next-states}(G,a)} V(G') 
ight)$$

#### Where's the Reward?

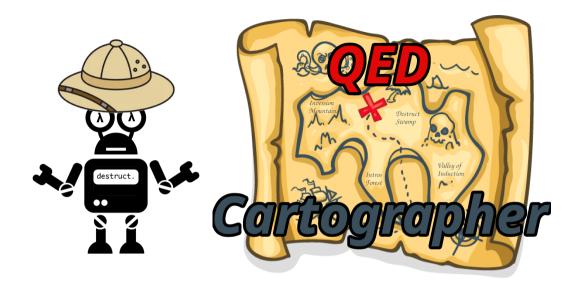




## Don't need one!



$$egin{aligned} V(G) &= \max_{a \in ext{actions}(G)} \left( \gamma \prod_{G' \in ext{next-states}(G,a)} V(G') 
ight) \ & \gamma \prod_{G' \in \emptyset} V(G') \ & \gamma(1) \end{aligned}$$



# Automating Formal Verification with Reward-Free Reinforcement Learning

26% Shorter Proofs in 27% Fewer Steps

## Benchmark: CoqGym



124 Coq Projects

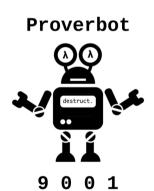
68,501 Theorems

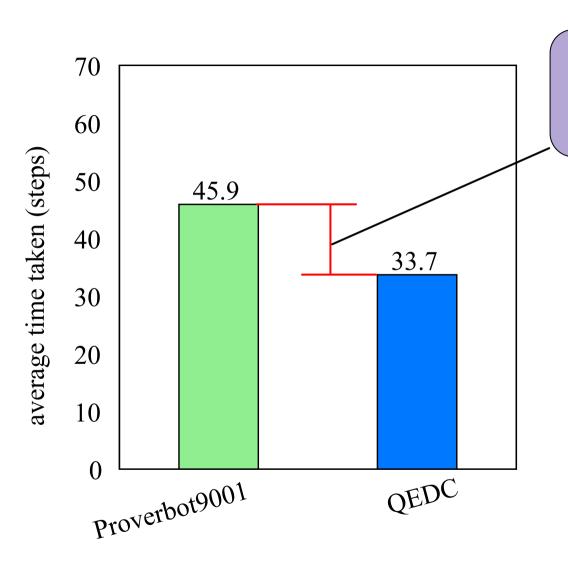
85/15 train-test split

### **Baseline: Proverbot9001 (updated)**

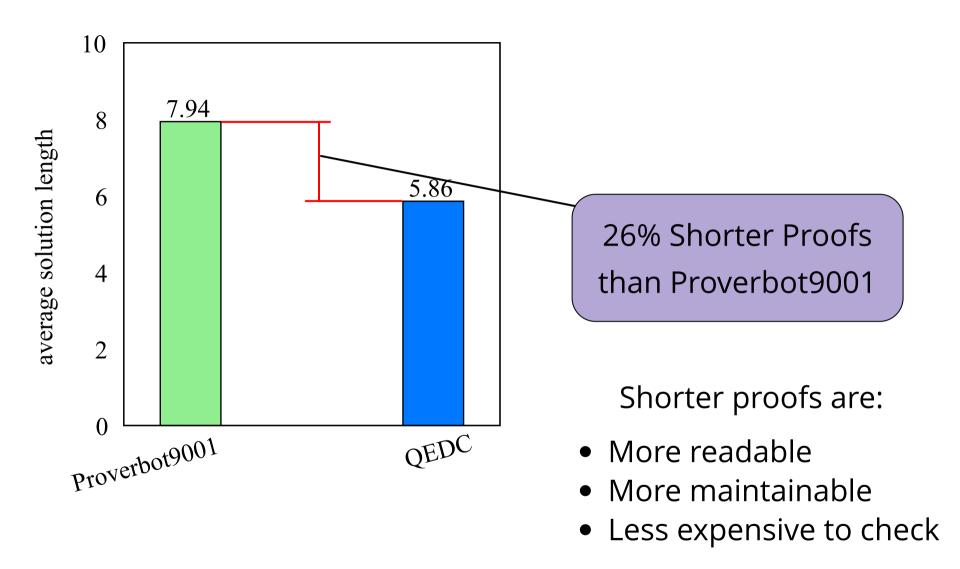
QEDCartographer, except without state scoring-based search

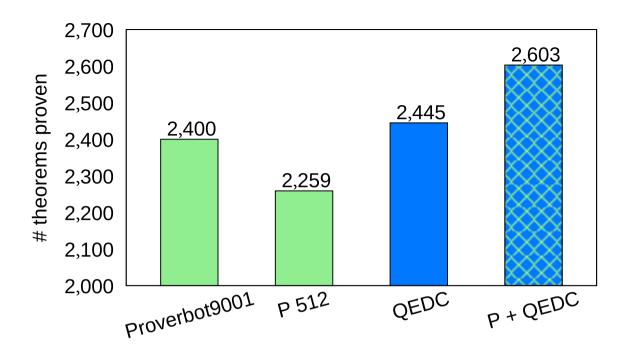
Uses a variant of depth-first search instead



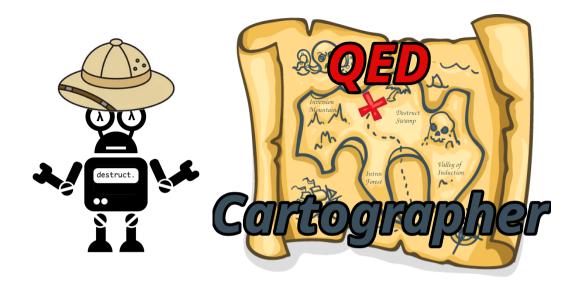


27% faster than Proverbot9001





Proves slightly more theorems, and proves complementary theorems



Automating Formal Verification with Reward-Free Reinforcement Learning

Uses a new V-value equation for branching goal structure

Makes producing verified-correct code easier and faster

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