



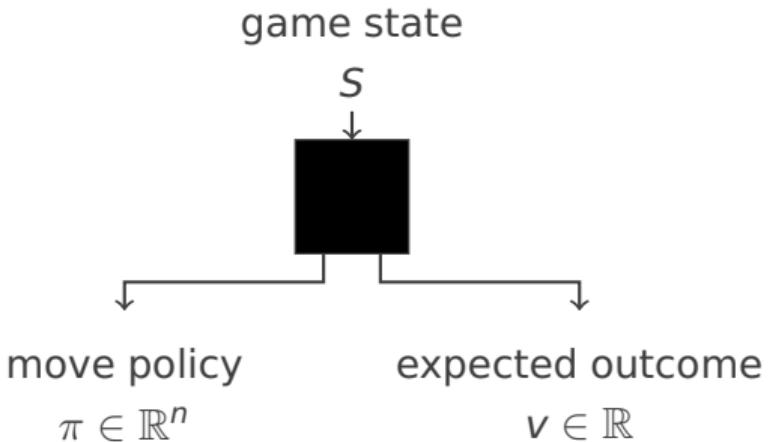
Learning theorem proving through self-play

Stanisław Purgał

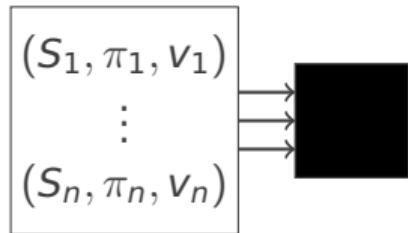
Overview

- AlphaZero
- Proving game
- adjusting MCTS for proving game
- some results

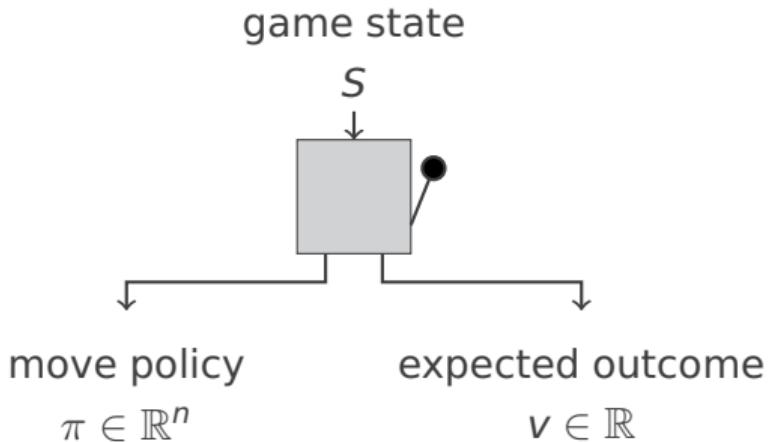
Neural black box



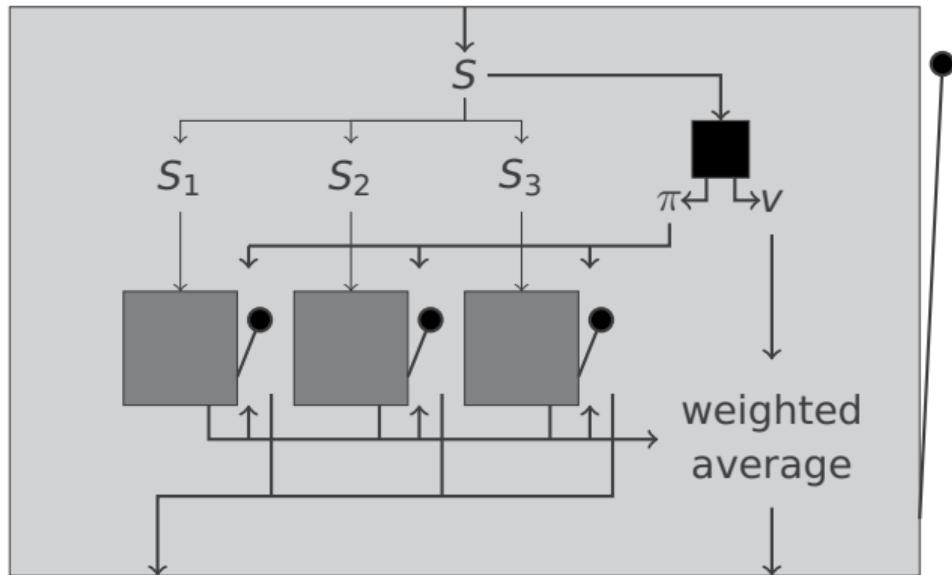
Neural black box



Monte-Carlo Tree Search



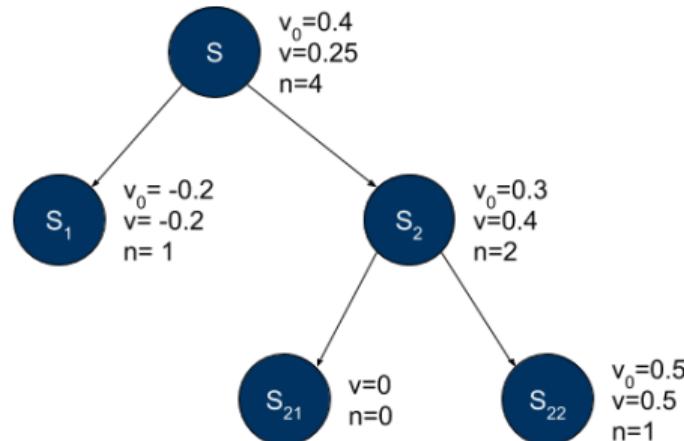
Monte-Carlo Tree Search



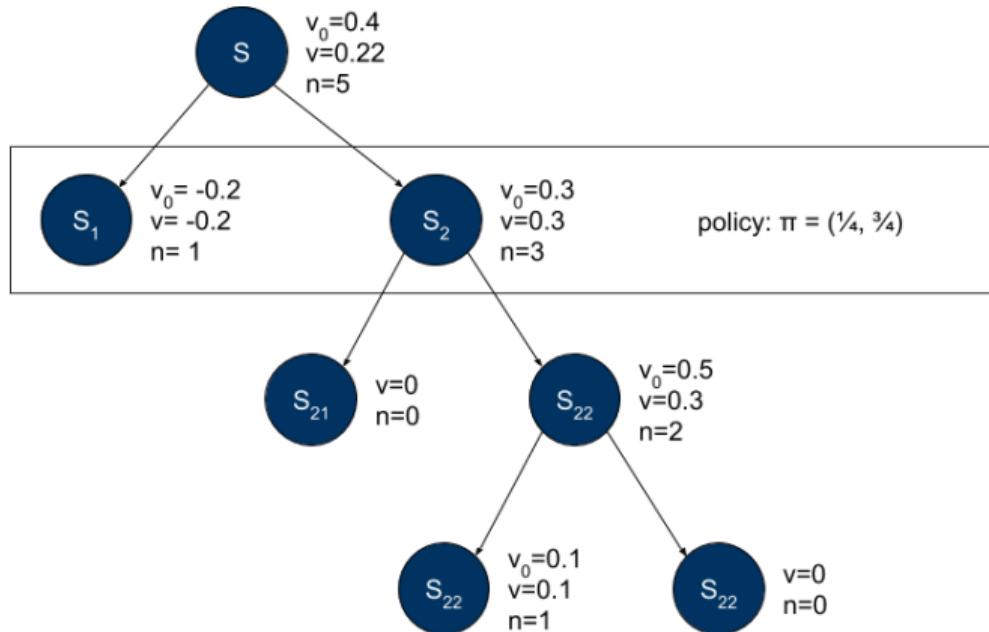
choose a child according
to the formula:

$$c = \left(\log \frac{n+c_{base}+1}{c_{base}} + c_{init} \right)$$
$$c_{base} = 19652$$
$$c_{init} = 1.25$$

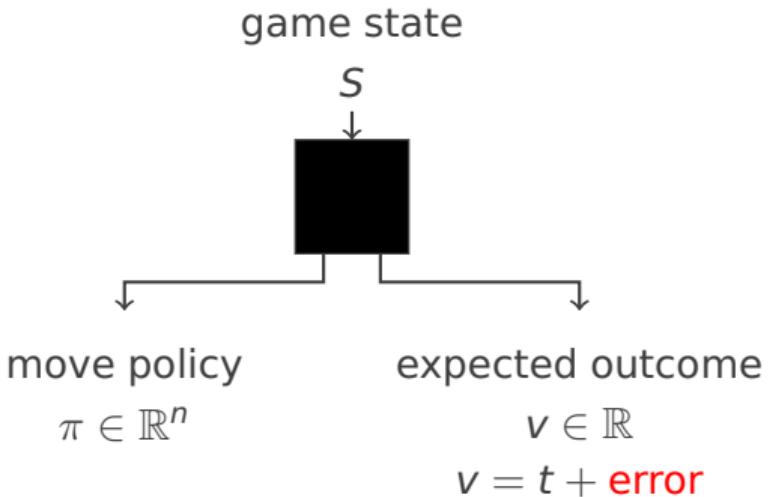
Monte-Carlo Tree Search



Monte-Carlo Tree Search



Why not maximum?



Why not maximum?

$$v_1 = t_1 + \text{error}$$

$$v_2 = t_2 + \text{error}$$

$$v_3 = t_3 + \text{ERROR}$$



min / max



$$v = t + \text{ERROR}$$

Why not maximum?

$$v_1 = t_1 + \text{error}$$

$$v_2 = t_2 + \text{error}$$

$$v_3 = t_3 + \text{ERROR}$$



average

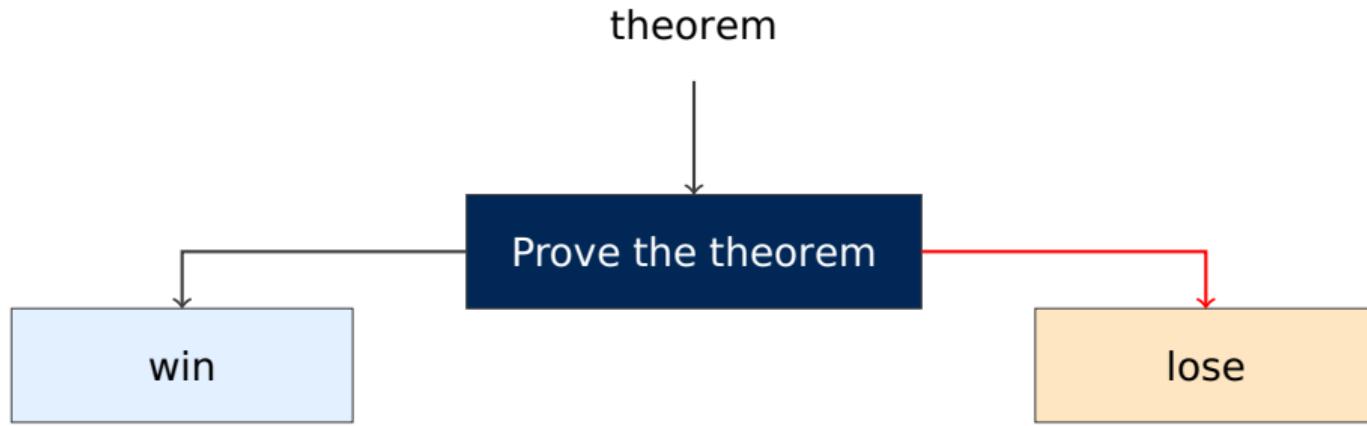


$$v = t + \frac{\sum \text{error}}{n}$$

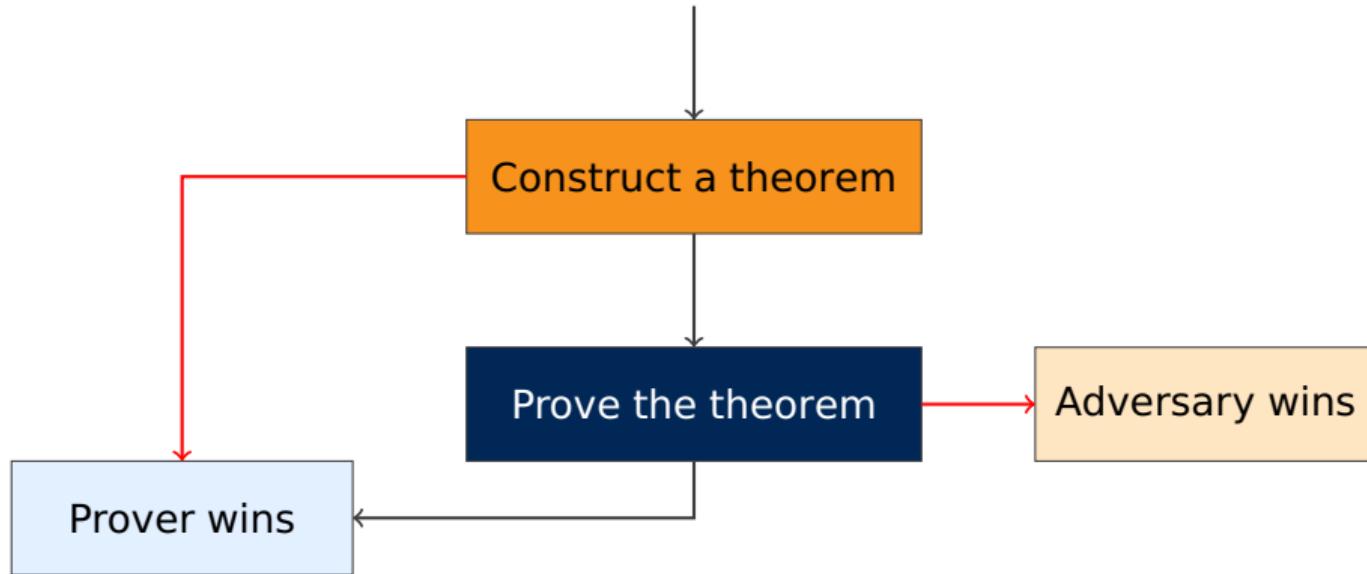
Closing the loop

- play lots of games
- choose moves randomly, according to MCTS policy
- use finished games for training:
 - desired value in the result of the game
 - desired policy is the MCTS policy
- also add noise to neural network output to increase exploration

Proving game



Proving game



Prolog-like proving

$$\frac{A \vdash X \quad A \vdash Y}{A \vdash X \wedge Y} \quad (1)$$

`holds(A, and(X, Y)) :- holds(A, X), holds(A, Y)` (2)

Prolog-like proving

[$X:A \vdash X \wedge \neg\neg X$, ...]



$A \vdash X \wedge Y :- A \vdash X, A \vdash Y$



$X:A \vdash X \wedge \neg\neg X :- X:A \vdash X, X:A \vdash \neg\neg X$



[$X:A \vdash X, X:A \vdash \neg\neg X$, ...]

Prolog-like proving

[$X:A$, $\text{and}(X, \text{not}(\text{not}(X)))$, ...]



$\text{holds}(A, \text{and}(X, Y)) :- \text{holds}(A, X), \text{holds}(A, Y)$



$\text{holds}(X:A, \text{and}(X, \text{not}(\text{not}(X)))) :- \text{holds}(X:A, X), \text{holds}(X:A, \text{not}(\text{not}(X)))$



[$\text{holds}(X:A, X)$, $\text{holds}(X:A, \text{not}(\text{not}(X)))$, ...]

Prolog-like theorem constructing

[$\text{holds}(X:A, \text{and}(X, \text{not}(\text{not}(X)))) , \dots]$



$\text{holds}(X:A, \text{and}(X, \text{not}(\text{not}(X)))) :- \text{holds}(X:A, X), \text{holds}(X:A, \text{not}(\text{not}(X)))$



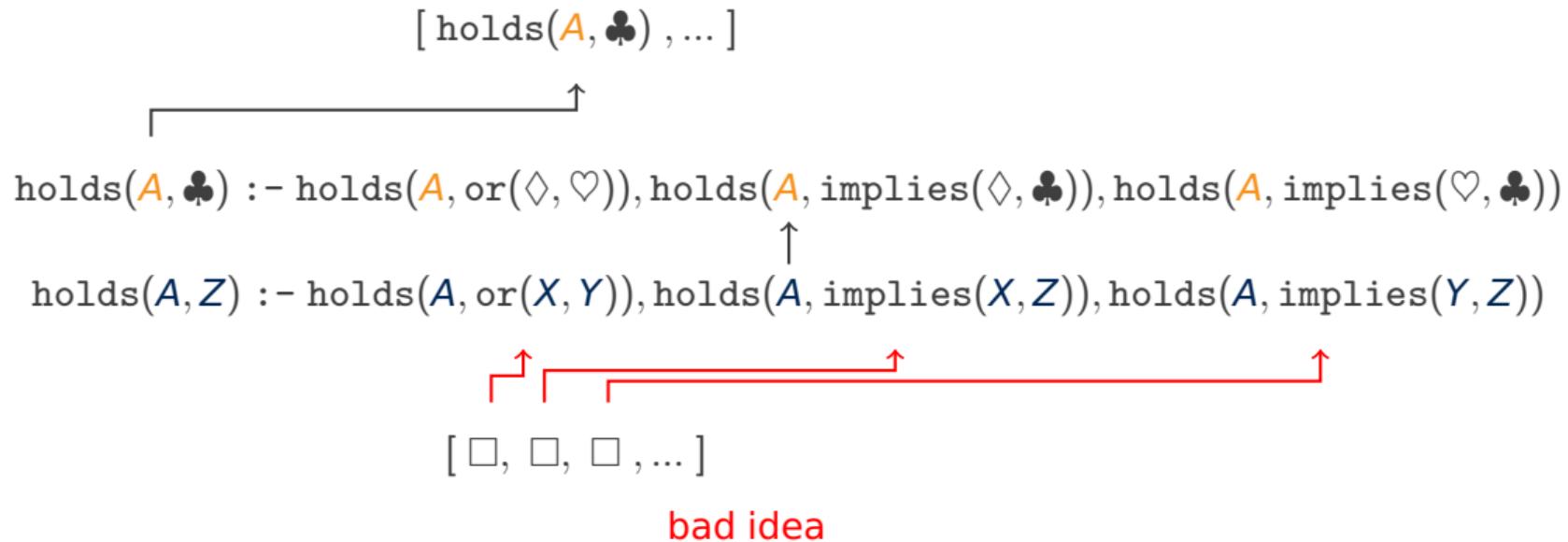
$\text{holds}(A, \text{and}(X, Y)) :- \text{holds}(A, X), \text{holds}(A, Y)$



[$\text{holds}(X:A, X), \text{holds}(X:A, \text{not}(\text{not}(X))) , \dots]$

bad idea

Prolog-like theorem constructing



Prolog-like theorem constructing

[T]



$\text{holds}(A, \text{and}(X, Y)) :- \text{holds}(A, X), \text{holds}(A, Y)$



$\text{holds}(A, \text{and}(X, Y)) :- \text{holds}(A, X), \text{holds}(A, Y)$



[$\text{holds}(A, X), \text{holds}(A, Y)$]

Prolog-like theorem constructing

T
↓
`holds($X:A$, and(X , not(not(X))))`
↓
`holds(x:a, and(x, not(not(x))))`

Forcing termination of the game

Step limit:

- ugly extension of game state
- strategy may depend on number of steps left
- even if we hide it, there is a correlation:
large term constructed \sim few steps left \sim will likely lose

Forcing termination of the game

Sudden death chance:

- game states nicely equal
- no hard limit for length of a theorem

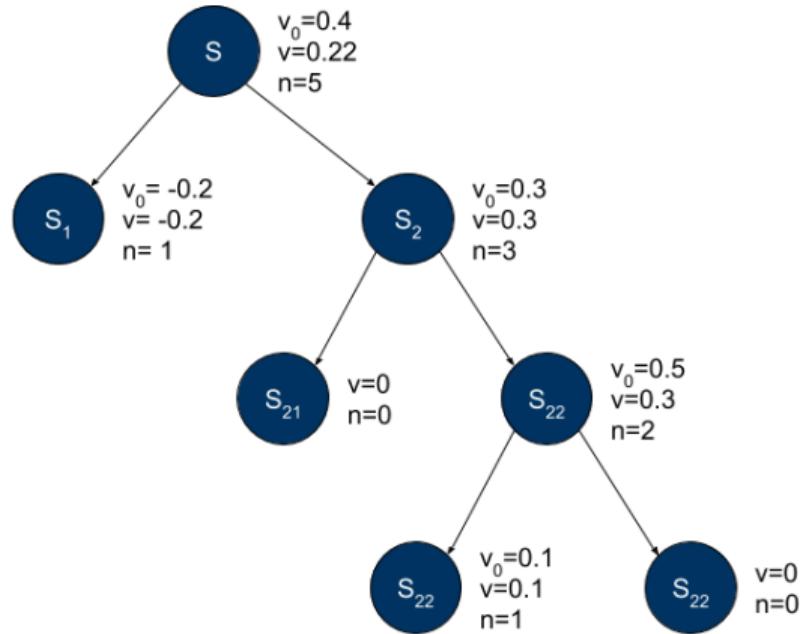
During training playout, randomly terminate game with chance p_d .

In MCTS, adjust value $v' = (-1) \cdot p_d + v \cdot (1 - p_d)$.

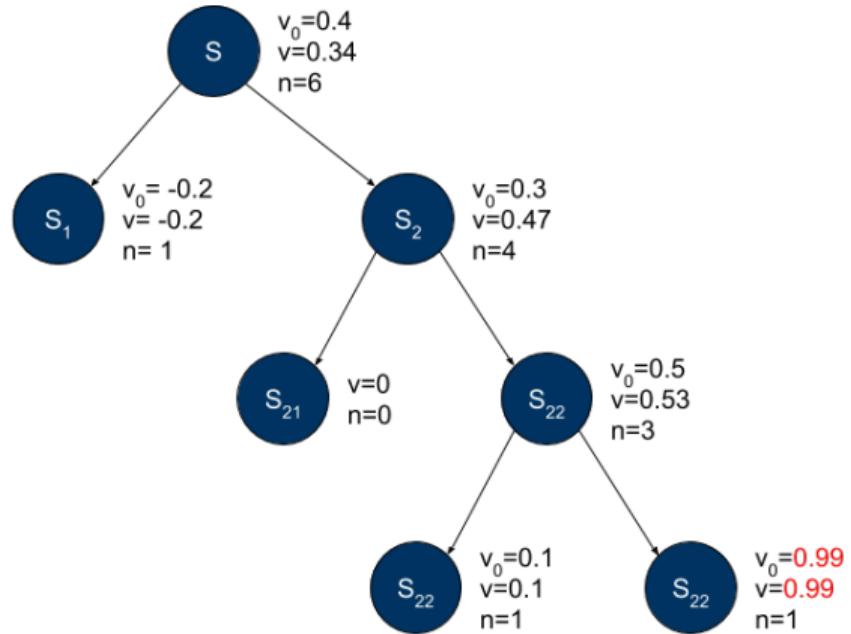
Disadvantages of this game

- two different players - if one player starts winning every game, we can't learn much
- proof use single inference steps - inefficient
- players don't take turns - MCTS not designed for that situation

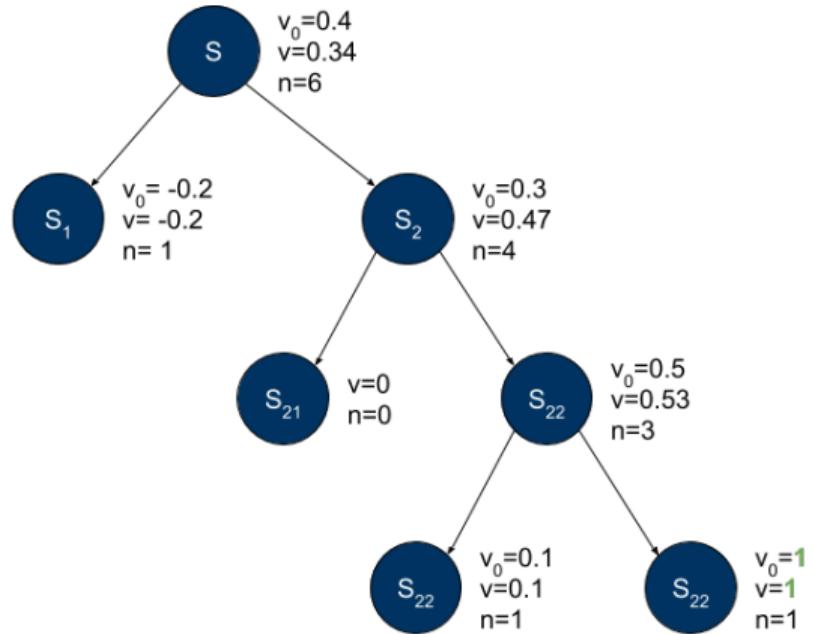
Not using maximum



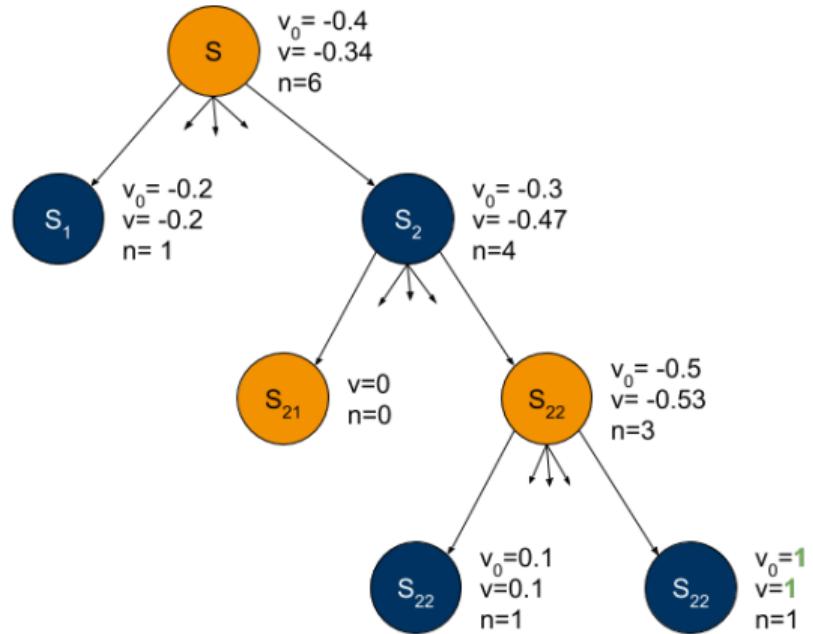
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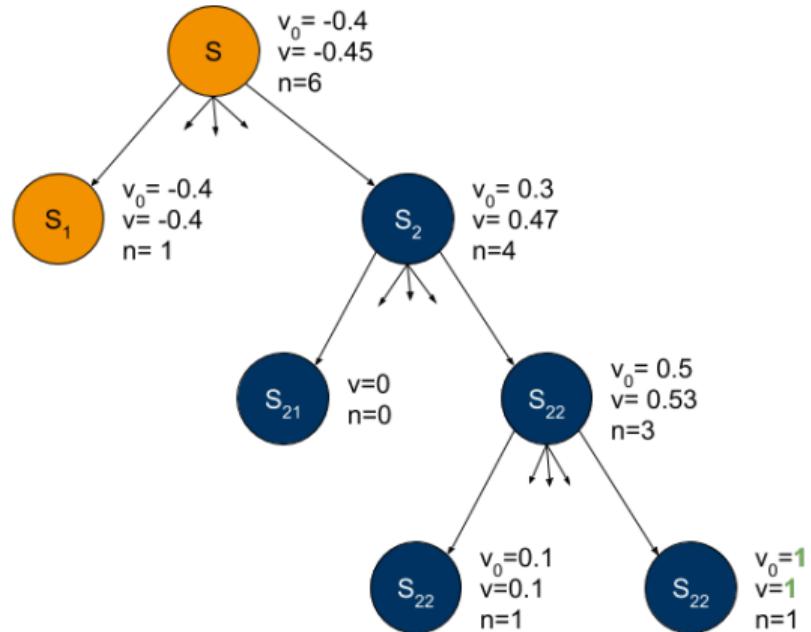
Not using maximum



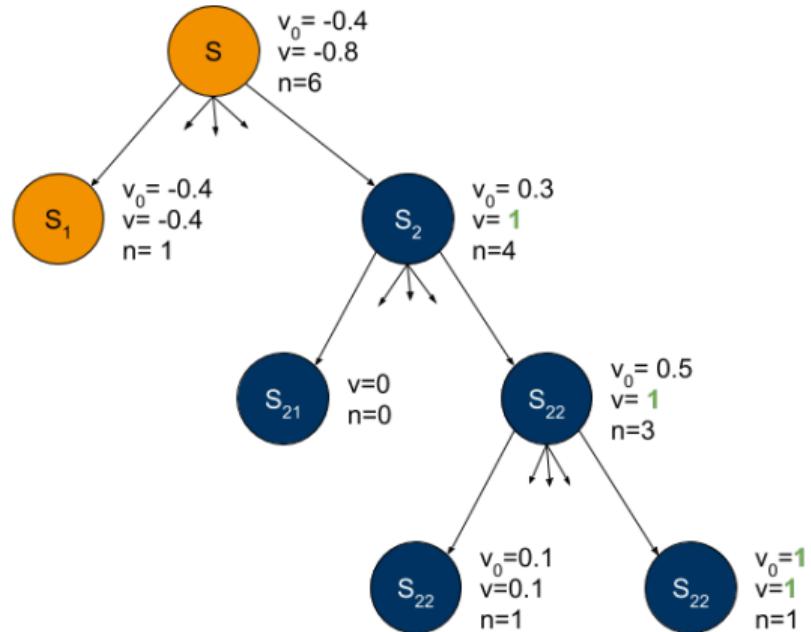
Not using maximum



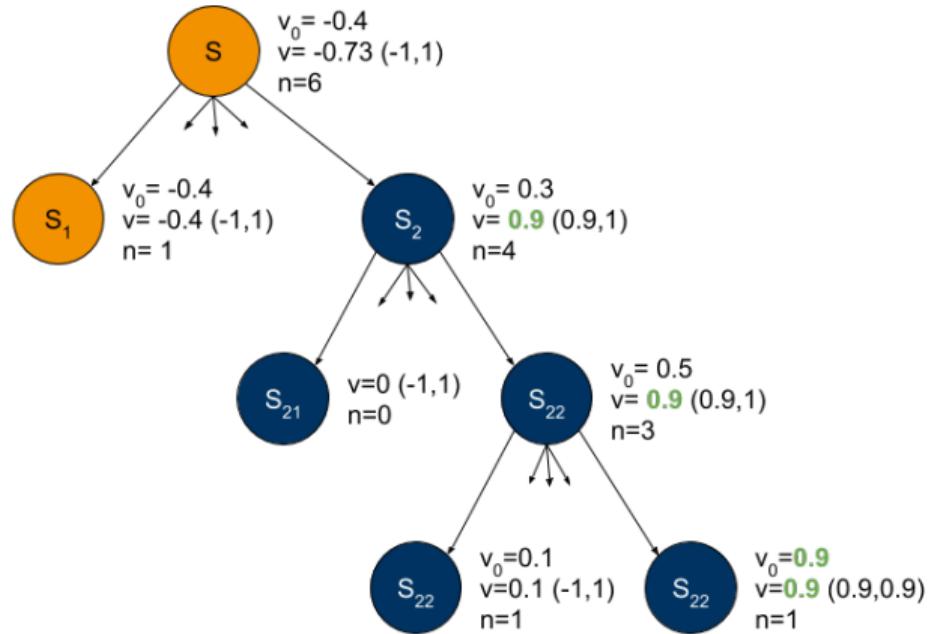
Certainty propagation



Certainty propagation



Certainty propagation



Certainty propagation

for uncertain leafs:

$$v = \blacksquare$$

$$a = \blacksquare$$

$$l = -1$$

$$u = 1$$

for certain leafs:

$$v = \text{result}$$

$$a = \text{result}$$

$$l = \text{result}$$

$$u = \text{result}$$

recursively:

$$v = \min(u, \max(l, a))$$

$$a = \frac{\blacksquare + \sum v_i \cdot n_i}{n+1}$$

$$l = \max_i l_i$$

$$u = \max_i u_i$$

when player changes:

- values and bounds flip
- lower and upper bound switch places

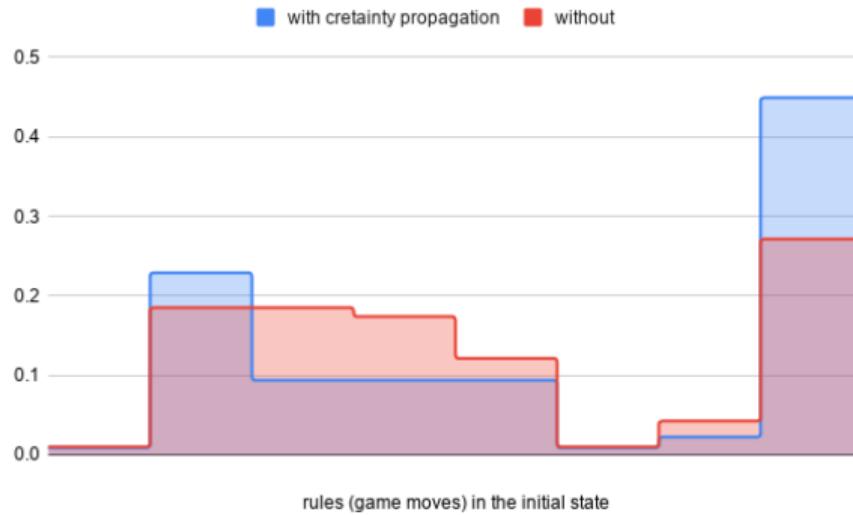
Toy problem

```
ablist([]).  
ablist([a|L]) :- ablist(L).  
ablist([b|L]) :- ablist(L).  
ablist([c|L]) :- ablist(L).  
ablist([d|L]) :- ablist(L).  
rev3([], L, L).  
rev3([H|T], L, Acc) :- rev3(T, L, [H|Acc]).  
revablist(L) :- ablist(T), rev3(L, T, []).
```

Toy problem evaluation

```
ablist([a,b,a,b,a,b,b]),  
revablist([]),  
revablist([a]),  
revablist([b]),  
revablist([c,d]),  
revablist([c,a,b]),  
revablist([a,d,c,b]),  
revablist([a,d,c,a,a]),  
revablist([a,b,c,d,b,d]),  
revablist([d,b,c,a,d,a,b]),  
revablist([a,c,b,a,c,a,d,d])
```

Certainty propagation effect

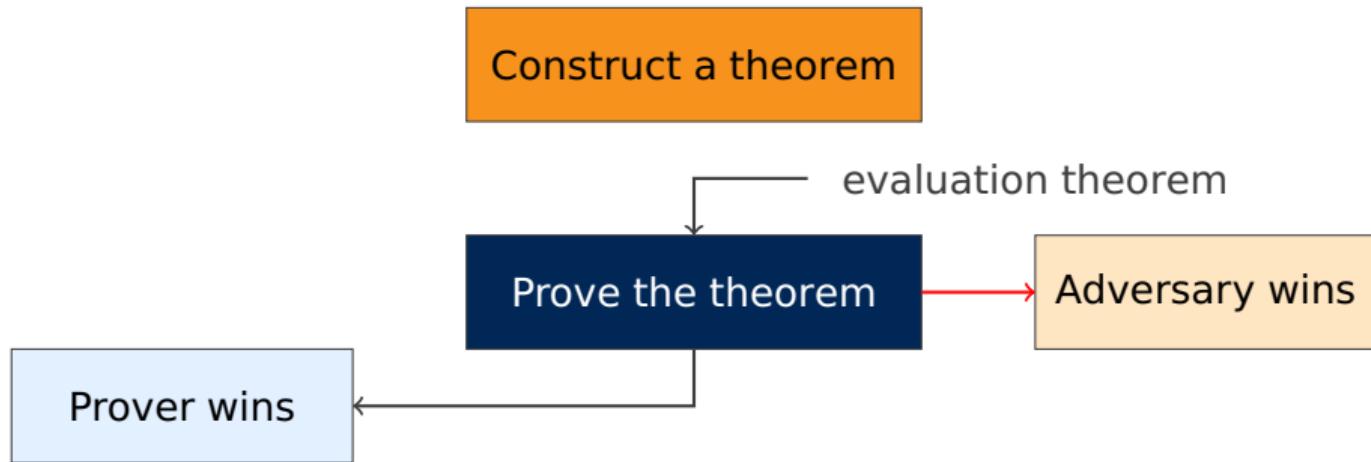


Learning the proving game

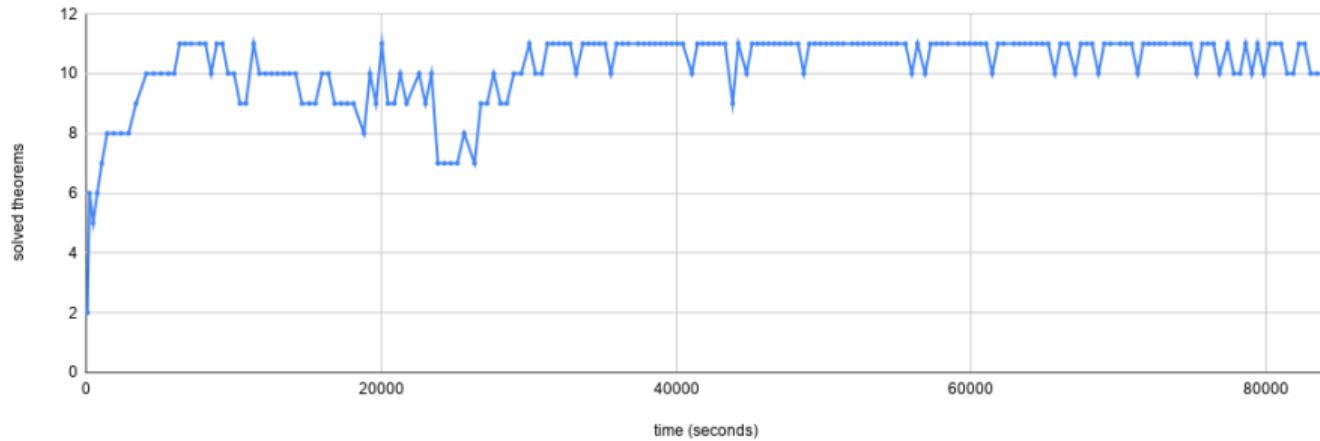
Like AlphaZero, with few differences:

- using Graph Attention Network for ■
- for theorems that prover failed to prove, show proper path with additional policy training samples
- during evaluation, greedy policy and step limit instead of sudden death

Proving game evaluation



Learning toy problem



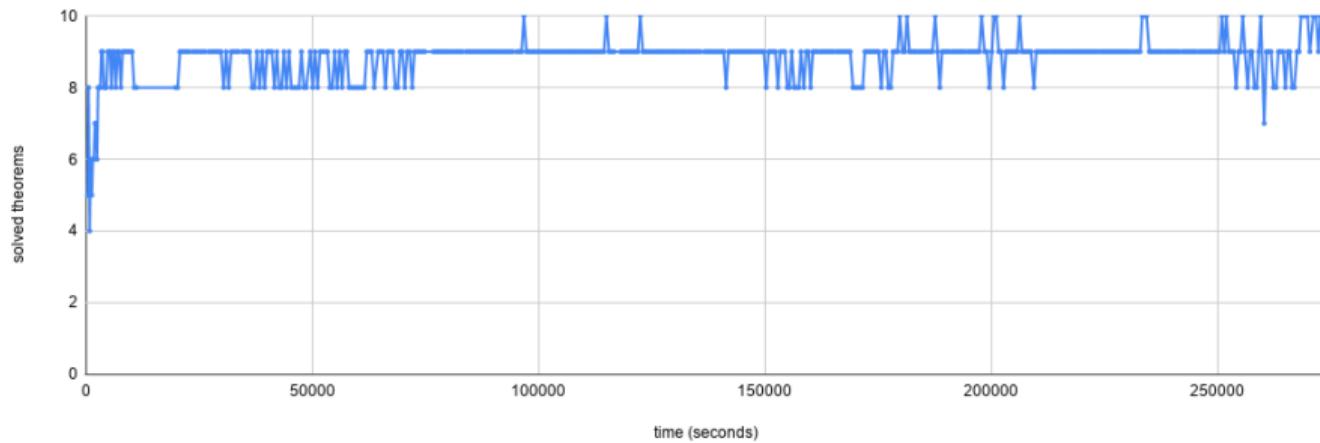
Intuitionistic propositional logic

```
holds([A|T], A).  
holds(T, A) :- holds([B|T], A), holds(T, B).  
holds([H|T], A) :- holds(T, A).  
holds(T, impl(A, B)) :- holds([A|T], B).  
holds(T, B) :- holds(T, A), holds(T, impl(A, B)).  
holds(T, or(A, B)) :- holds(T, A).  
holds(T, or(A, B)) :- holds(T, B).  
holds(T, C) :- holds(T, or(A, B)), holds([A|T], C), holds([B|T], C).  
holds(T, and(A, B)) :- holds(T, A), holds(T, B).  
holds(T, A) :- holds(T, and(A, B)).  
holds(T, B) :- holds(T, and(A, B)).  
holds([false|T], A).
```

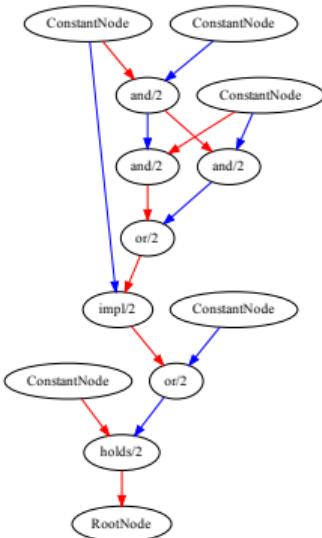
Classical propositional logic

```
holds([A|T], A).  
holds(T, A) :- holds([B|T], A), holds(T, B).  
holds([H|T], A) :- holds(T, A).  
holds(T, impl(A, B)) :- holds([A|T], B).  
holds(T, B) :- holds(T, A), holds(T, impl(A, B)).  
holds(T, or(A, B)) :- holds(T, A).  
holds(T, or(A, B)) :- holds(T, B).  
holds(T, C) :- holds(T, or(A, B)), holds([A|T], C), holds([B|T], C).  
holds(T, and(A, B)) :- holds(T, A), holds(T, B).  
holds(T, A) :- holds(T, and(A, B)).  
holds(T, B) :- holds(T, and(A, B)).  
holds([false|T], A).  
holds(T, A) :- holds([impl(A, false)|T], false).
```

Learning classical propositional logic



Constructed theorem example



$\vdash (((d \wedge b \wedge c) \vee (b \wedge c \wedge d)) \Rightarrow b) \vee e$
 $\perp \vdash a \vee b \vee c$
 $\vdash (((((a \wedge \perp \wedge b) \Rightarrow c) \Rightarrow d) \Rightarrow d)$
 $((a \wedge b) \Rightarrow a) \Rightarrow (\perp \wedge c) \vdash d$
 $((a \Rightarrow \perp) \Rightarrow b), c, (a \Rightarrow b) \vdash b$

First-order logic

%some classical logic

```
neq(var([a|_]), var([b|_])).  
neq(var([b|_]), var([a|_])).  
neq(var([_|A]), var([_|B])) :- neq(var(A), var(B)).  
  
repl(var(A), R, var(A), R).  
repl(var(A), R, var(B), var(B)) :- neq(var(A), var(B)).  
repl(var(A), R, op(0, X1, Y1), op(0, X2, Y2)) :- repl(var(A), R, X1, X2), repl(var(A), R, Y1, Y2).  
repl(var(A), R, q(0, var(A), P), q(0, var(A), P)).  
repl(var(A), R, q(0, var(B), P1), q(0, var(B), P2)) :- neq(var(A), var(B)), repl(var(A), R, P1, P2).  
repl(var(A), R, false, false).  
repl(var(A), R, [], []).  
repl(var(A), R, [H1|T1], [H2|T2]) :- repl(var(A), R, H1, H2), repl(var(A), R, T1, T2).  
  
holds(T, q(forall, var(A), Phi)) :- repl(var(A), var(B), Phi, PhiBA), repl(var(B), false, [Phi|T], [Phi|T]), holds(T, PhiBA).  
holds(T, Phi) :- holds(T, q(forall, var(A), PhiA)), repl(var(A), B, PhiA, Phi).  
  
holds(T, q(exists, var(A), Phi)) :- repl(var(A), R, Phi, PhiR), holds(T, PhiR).  
holds(T, P) :- holds(T, q(exists, var(A), Phi)), repl(var(B), false, Phi, Phi), repl(var(B), var(B), Phi, PhiB), holds([PhiB|T], P).
```

Future work

- better rule representation?
- proper prover with a different construction mechanism?
- different use cases?
- more computational power?



Thank you for your attention!

Stanisław Purgał