

Learning Dynamic Polynomial Proofs

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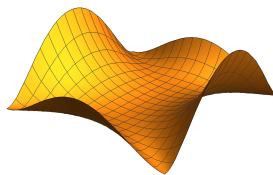


Proving with polynomials

For a multivariate polynomial p , consider the following task:

Show that $p(x_1, \dots, x_n) \geq 0$ for all $x \in [0, 1]^n$

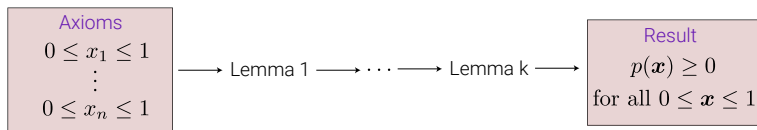
Many hard problems can be formulated as such; e.g., SAT, TSP, stable set, max-cut, etc.



Goal

Train an agent using Reinforcement Learning to prove the non-negativity of polynomials.

Proof system



- **Inference rules:**

- $h \geq 0 \implies x_j h \geq 0,$
- $h \geq 0 \implies (1 - x_j)h \geq 0,$
- $h_i \geq 0 \implies \sum_i \lambda_i h_i \geq 0, \forall \lambda_i \geq 0.$

- **Proof of $p \geq 0$:** corresponds to the **composition** of inference rules, which yields exactly the polynomial p .

Learning dynamic proofs of polynomials

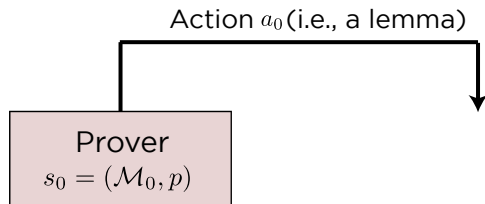
Prover

$$s_0 = (\mathcal{M}_0, p)$$

\mathcal{M}_0

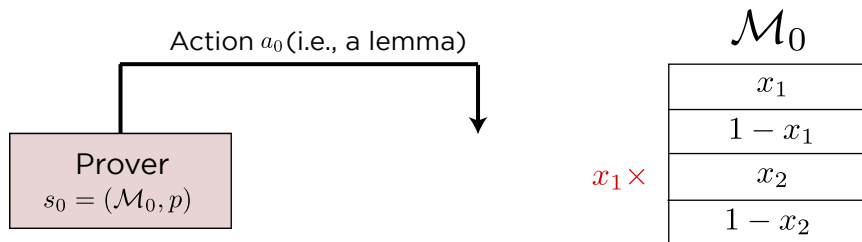
x_1
$1 - x_1$
x_2
$1 - x_2$

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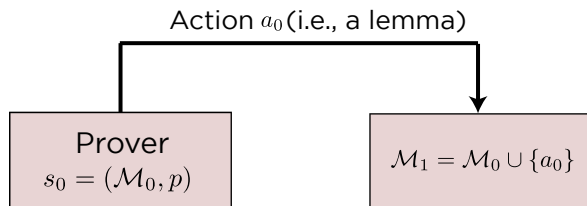

$$\mathcal{M}_0$$

x_1
$1 - x_1$
x_2
$1 - x_2$

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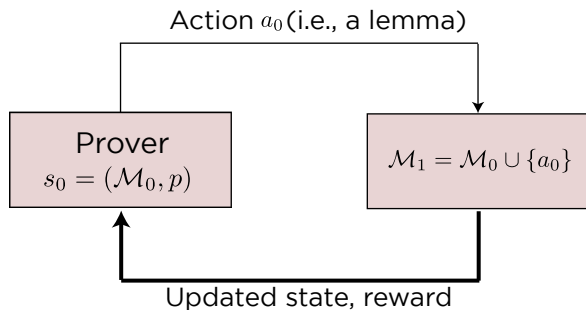
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\mathcal{M}_1

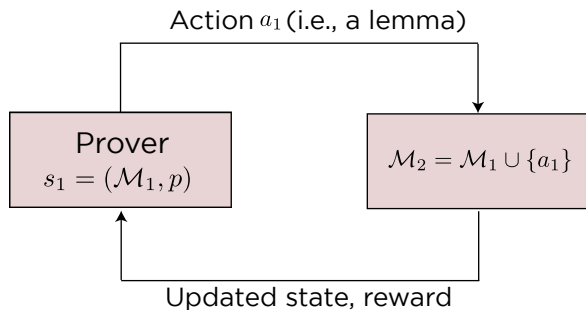
x_1
$1 - x_1$
x_2
$1 - x_2$
$x_1 x_2$

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$$\mathcal{M}_1$$

x_1
$1 - x_1$
x_2
$1 - x_2$
$x_1 x_2$

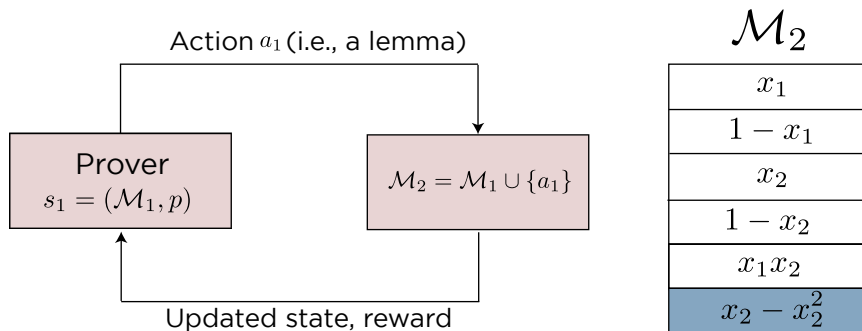
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\mathcal{M}_2

x_1
$1 - x_1$
x_2
$1 - x_2$
$x_1 x_2$
$x_2 - x_2^2$

Learning dynamic proofs of polynomials



We use DQN to train the prover. Two important ingredients:

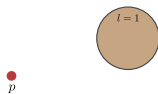
- We use **dense, unsupervised** rewards.
- We incorporate important **symmetries** in the Q-network.

Comparison to static approach

Static approach. Inference rules are unrolled for n steps.

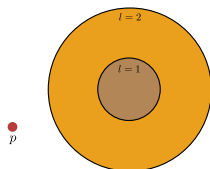
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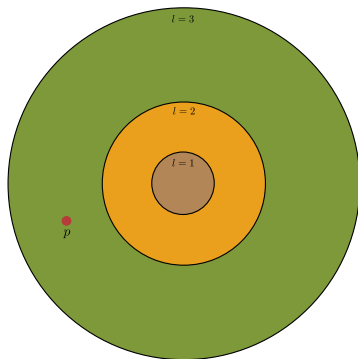
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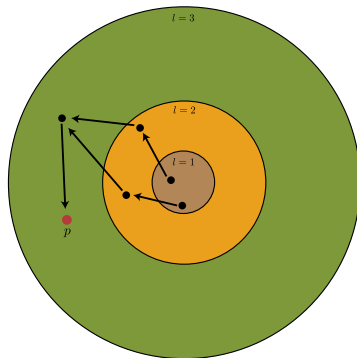
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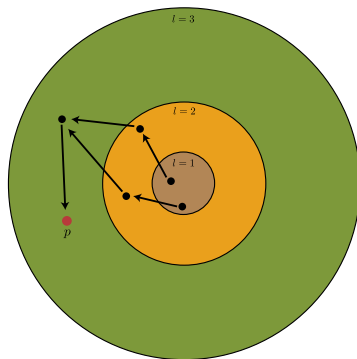
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Comparison to static approach

Static approach. Inference rules are unrolled for l steps.



Key result. Reduction of the size of the linear program by several orders of magnitude compared to the static approach.

Come see our poster #120 for more details!