

Computational Complexity of Games

Kyle Burke

October 23, 2017

**Games and Graphs
Workshop**

- Combinatorial Game Theory
- Graph Game Parameters
- Complexity of Games

With tutorials by:

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- Rebecca Milley
- Richard Nowakowski

October 23-25, 2017
Lyon, France

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The Big Question

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- ▶ CS: Fastest algorithm to answer this for entire ruleset?

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 - ▶ $10n^2 + 37.4n + 124 \rightarrow 10n^2 \rightarrow n^2$
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 - ▶ $O(n^3 + n^2 + n + 50) = O(n^3)$

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Note: No fixed-size rulesets!

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- ▶ $O(n \log(m))$ steps.

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... are EXPTIME-complete.

EXPTIME-complete Rulesets?

Yes, there are rulesets that require exponential time to solve!

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How do we know there's no faster algorithm?

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How does this show hardness?

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 - ▶ B solves CHESS !
 - ▶ B solves CHESS in faster-than-exponential time!
 - ▶ Now CHESS is not EXPTIME-hard $\rightarrow \leftarrow$

What about Short Games?

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- ▶ ... and polynomial options.

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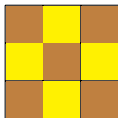
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↓ Left's potentially-winning options

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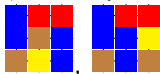
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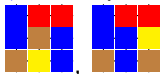
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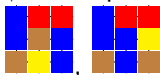
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$\in \mathcal{R} \cup \mathcal{P}$ (Right wins)

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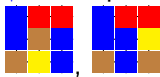
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↓ Right's potentially-winning options ✓

So... previous move not a winner for Left. Back up and try the next one.

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No Right move!

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No Right move! (Left wins)

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↓ Left's potentially-winning options



↓ Right's potentially-winning options



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No Right move! (Left wins) Go back and change Right's last move...

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- ▶ Continue to decide $G \in \mathcal{L} \cup \mathcal{N}$

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- ▶ Continue to decide $G \in \mathcal{L} \cup \mathcal{N}$
- ▶ How many rows of boards do I need?

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- ▶ How many rows of boards do I need? (polynomial)

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- ▶ Continue to decide $G \in \mathcal{L} \cup \mathcal{N}$
- ▶ How many rows of boards do I need? (polynomial)
- ▶ How many boards in a row?

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- ▶ Continue to decide $G \in \mathcal{L} \cup \mathcal{N}$
- ▶ How many rows of boards do I need? (polynomial)
- ▶ How many boards in a row? (polynomial)
- ▶ How much "workspace" do I need?

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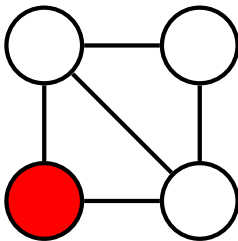
PSPACE-complete: Both PSPACE-hard and in PSPACE.

Example: IMPARTIAL COL is PSPACE-hard

IMPARTIAL COL: 2-coloring placement game

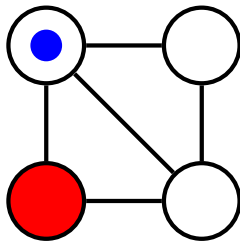
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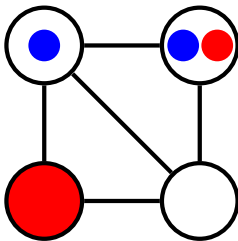
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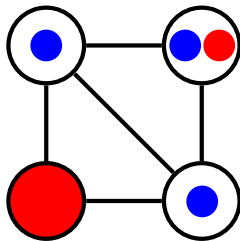
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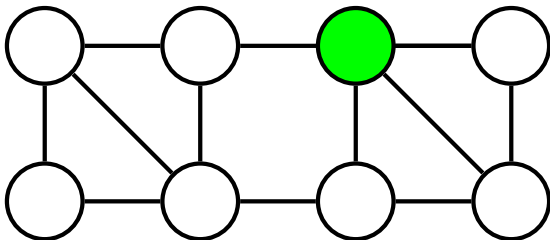
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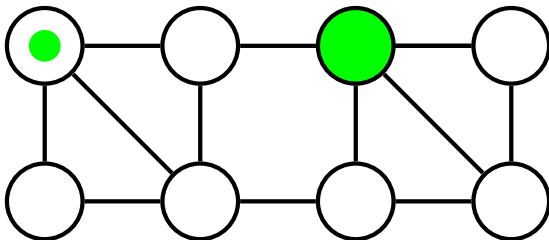
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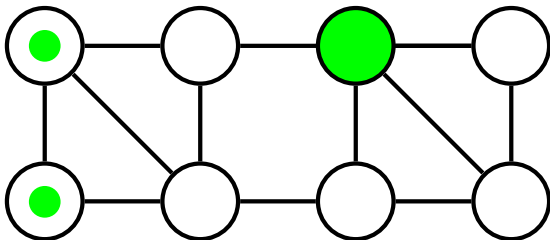
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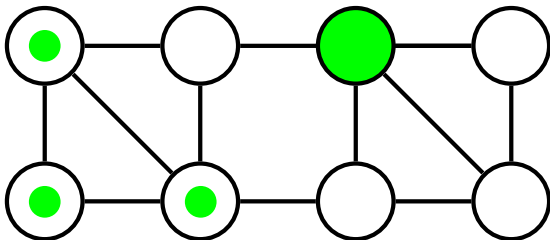
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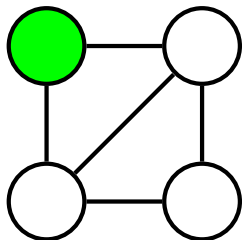


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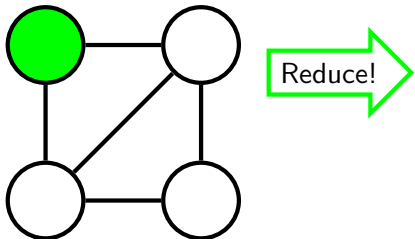
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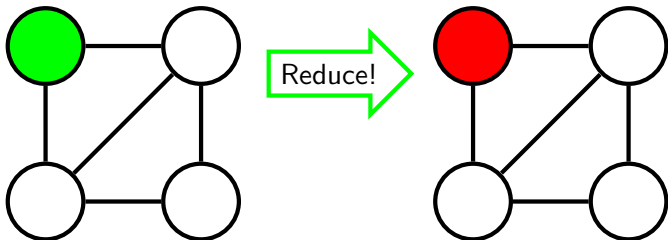
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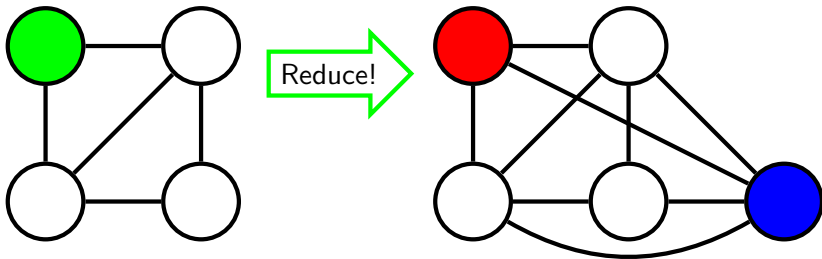
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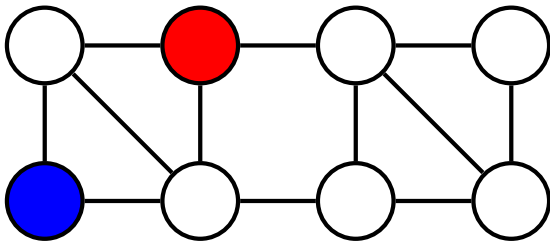


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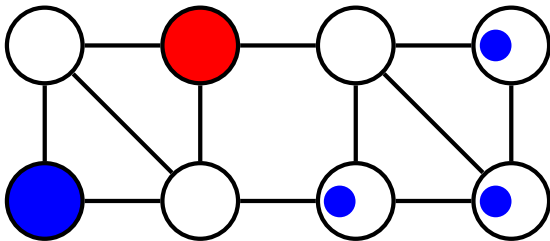
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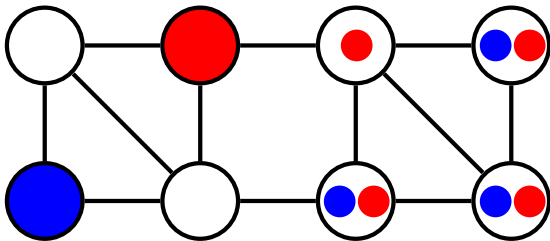
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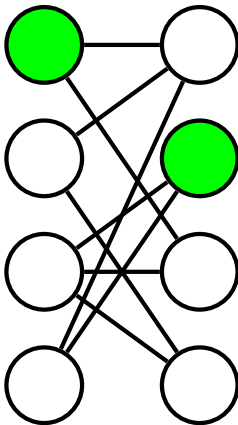
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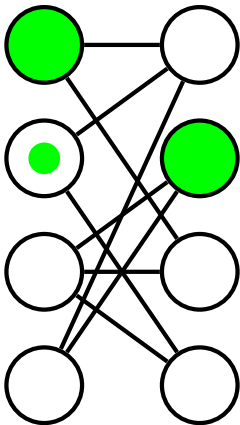
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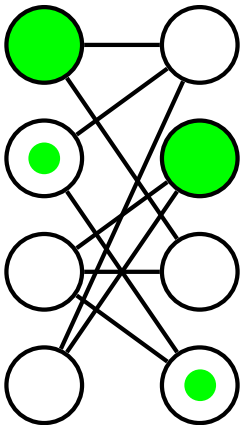
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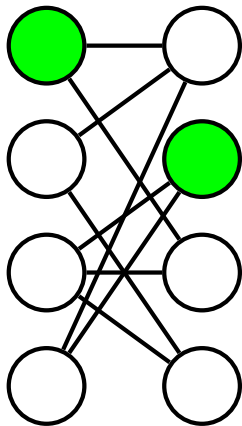
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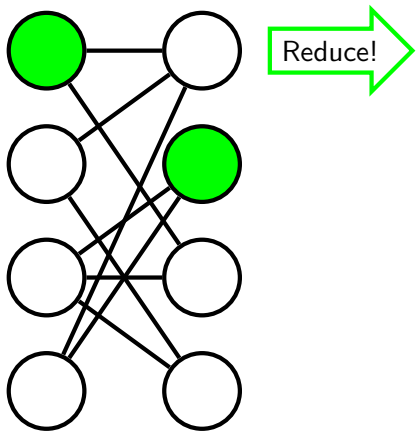
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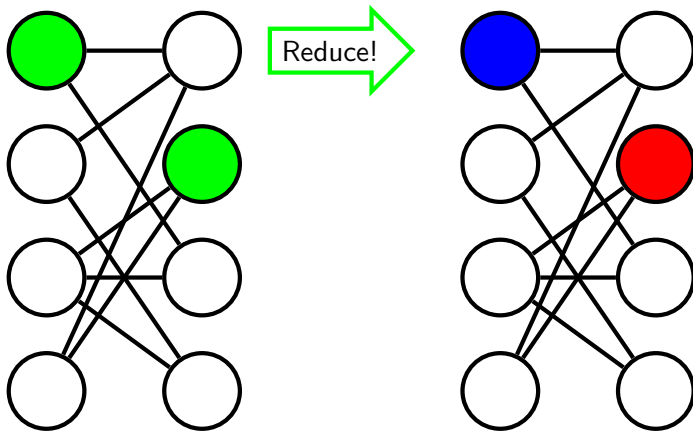
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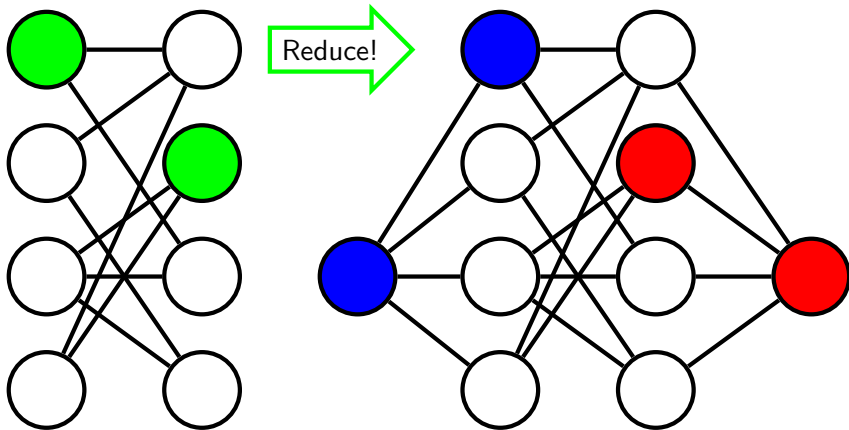
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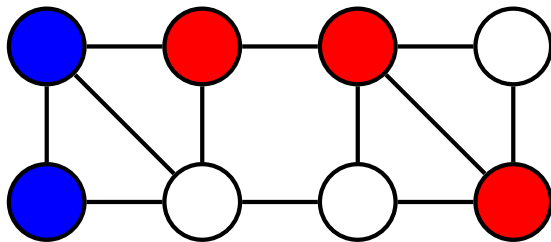


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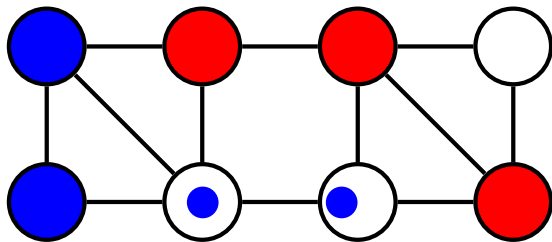
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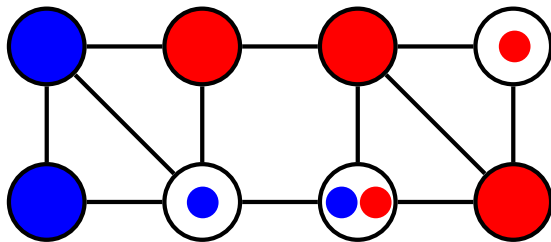
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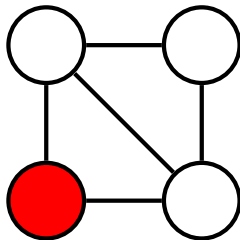
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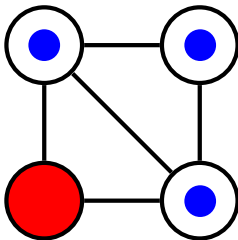
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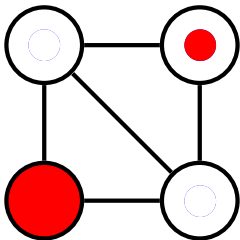
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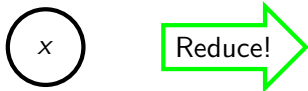


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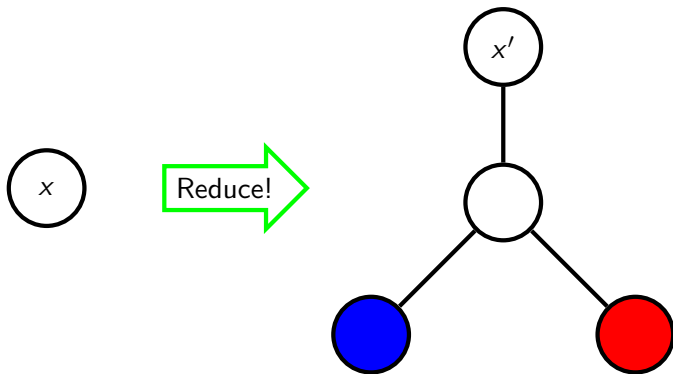


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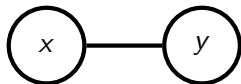


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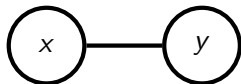
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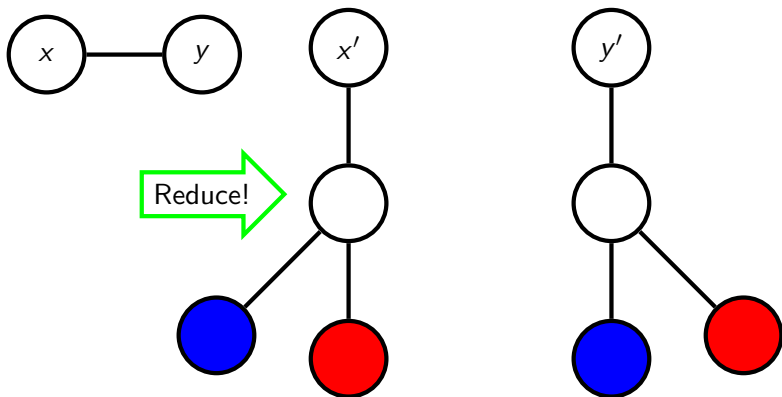
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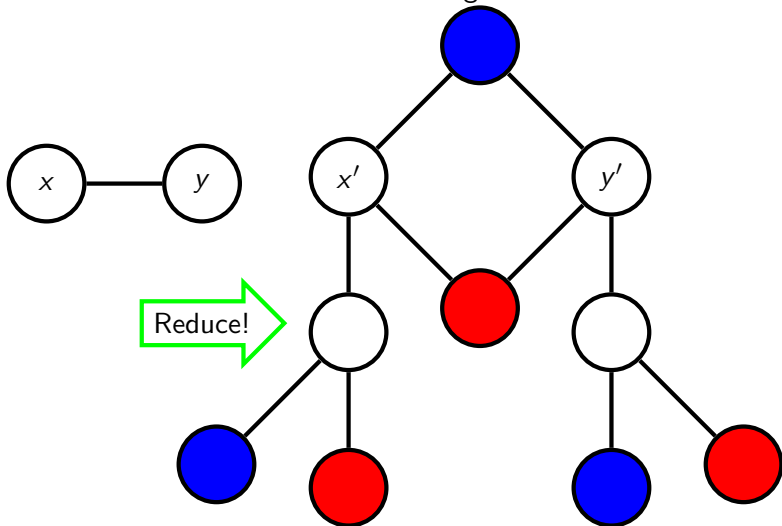
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Some of the first PSPACE-hard games were these, proven from QSAT.

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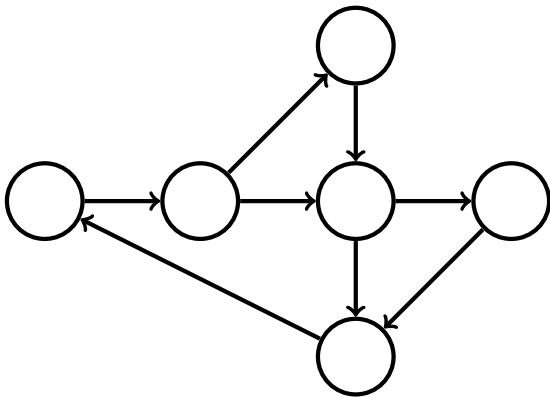
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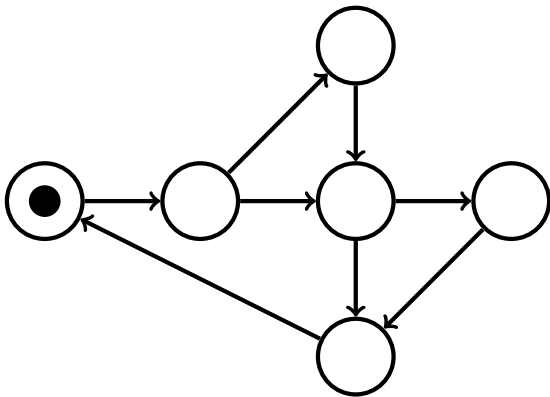
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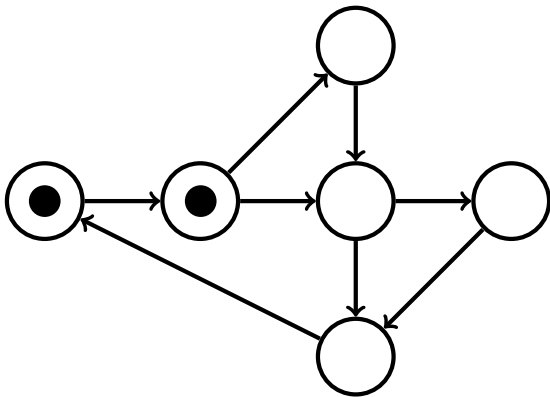
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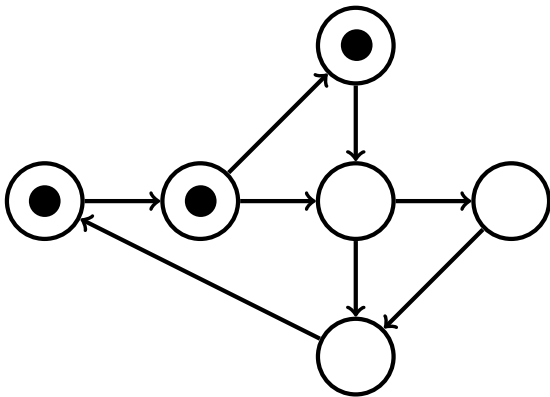
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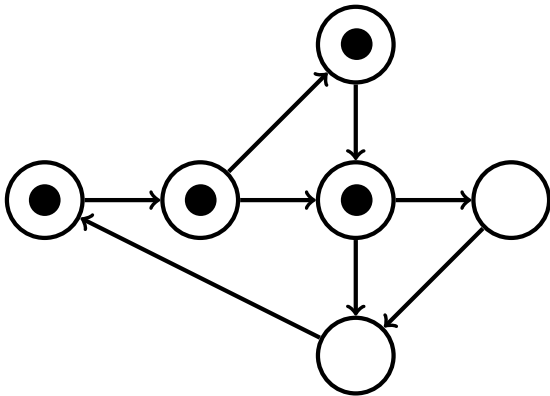
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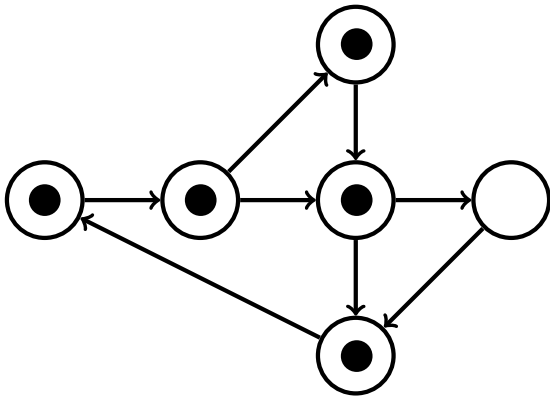
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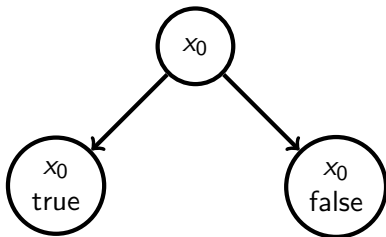
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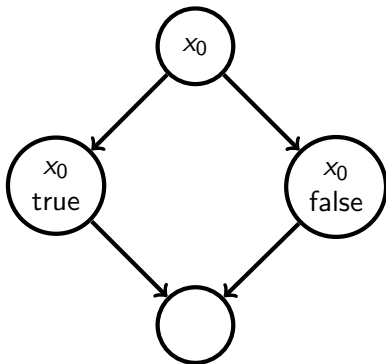
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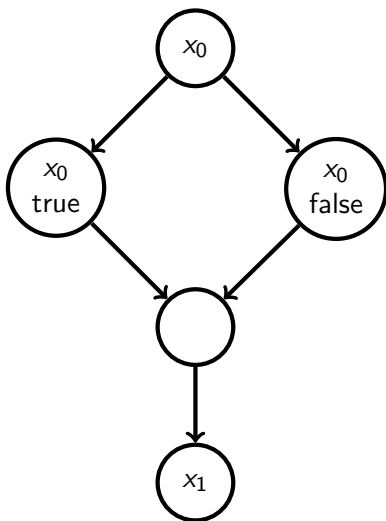
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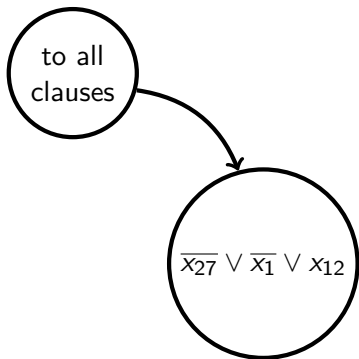
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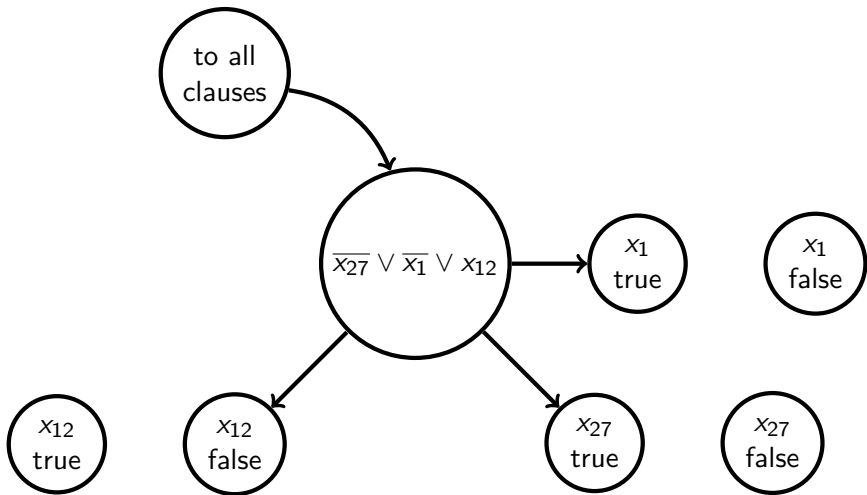
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Let's address the starting positions problem first.

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- ▶ Usually, starting positions are easy.

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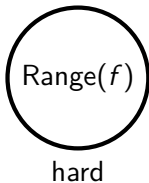
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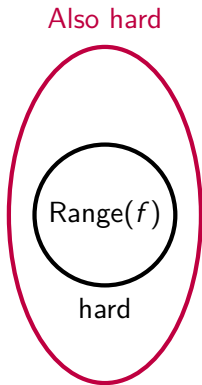
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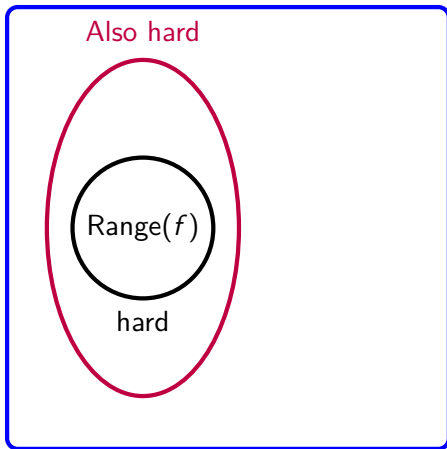
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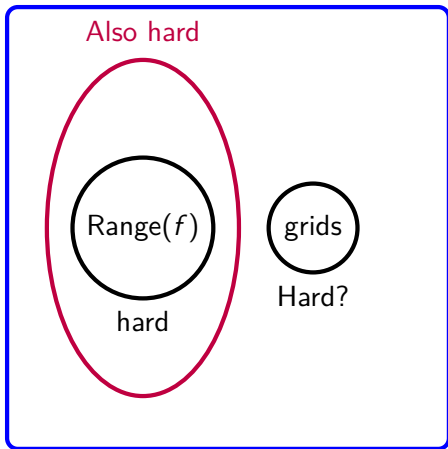
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Find the hardness, then use AI.

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Extra thanks to Dan Burgess and Matt Ferland for proof-watching
early versions of this talk.