



Scalable and Provable Distributed SAT Solvers

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Talk Takeaways

SAT solvers are general purpose **reasoning engines** that can be used for both **verification** and **optimization** purposes.

These solvers are the basis for **many of the automated reasoning tools** in use today.

Recent work in distributed **Clause Sharing Portfolio Solvers** has led to **dramatic performance improvements** over state-of-the-art sequential SAT solvers.

We have created a scalable approach for **proofs for distributed solvers** to ensure the correctness of solver results.

What is SAT?

Boolean satisfiability: Assign true/false to variables to make formula true

Some formulas don't have such an assignment

"standard" representation uses numbers for variables

$$(a \vee \neg b) \wedge (a \vee c \vee \neg d)$$

$$a = T, b = T, c = F, d = F$$

$$\neg a \wedge (a \vee \neg b) \wedge (a \vee b)$$

$$(a \vee \neg b) \wedge (a \vee c \vee \neg d)$$

1	-2			0
1		3	-4	0

SAT Solvers

-1	-2		0
	2	-4	0
1	2	4	0
-1		-3	0

Learn extra clauses to simplify search---
prune search space

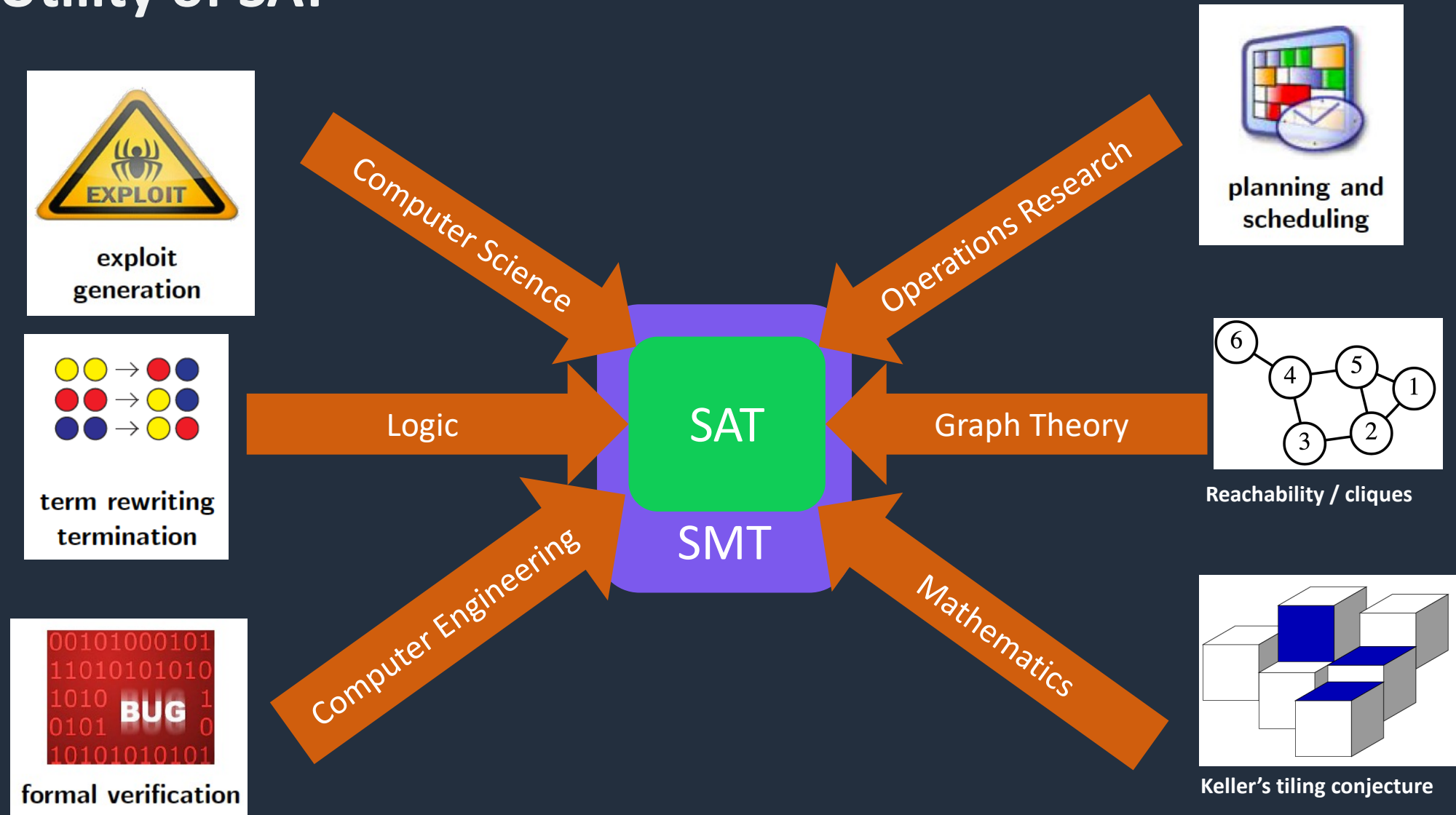
-1	-4	0
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Problem is unsatisfiable if the empty
clause is learned

0



The Utility of SAT



Scaling Proof Engines



Overnight



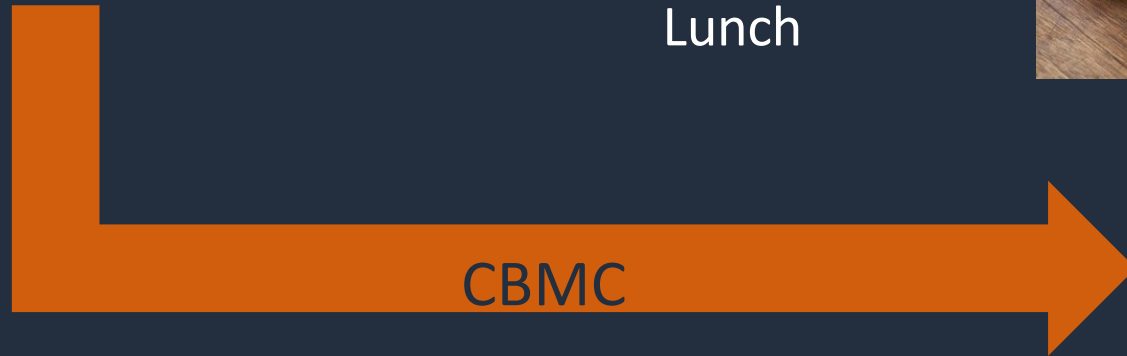
Lunch



Coffee



Squiggle



Scaling Proof Engines



Overnight



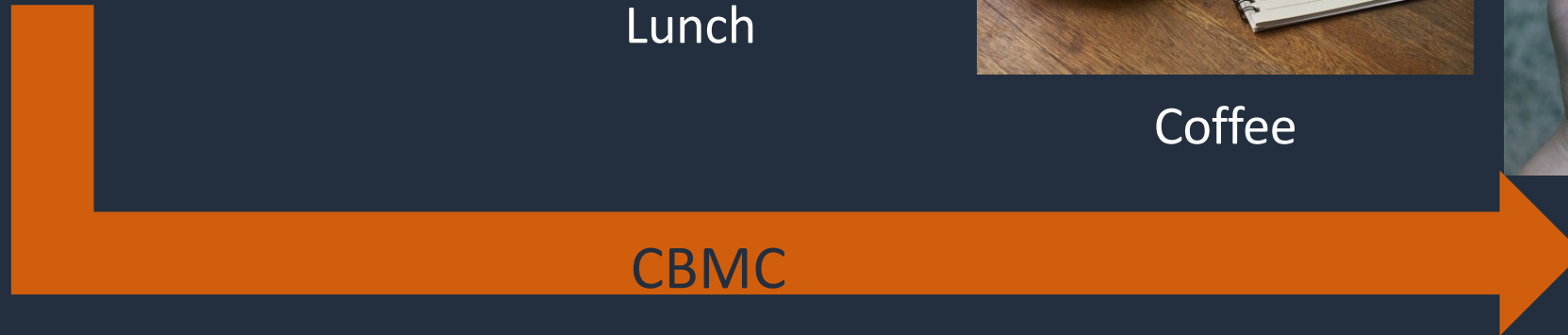
Lunch



Coffee



Squiggle



Scaling Proof Engines

ABSTRACTIONS BLOG

Computer Scientists Attempt to Corner the Collatz Conjecture

13 |

A powerful technique called SAT solving could work on the notorious Collatz conjecture. But it's a long shot.



SAT Scalability

Mid 90s: **thousands** of clauses & variables

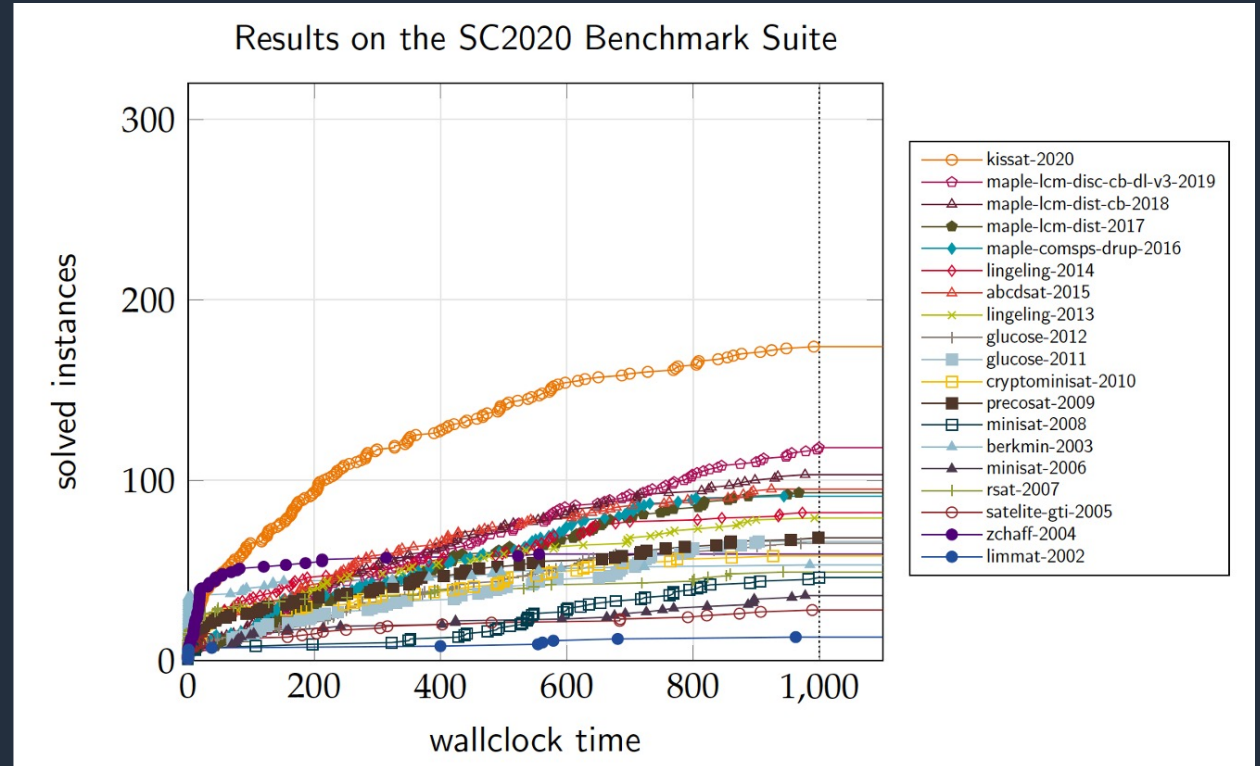
In 2020: **millions** of clauses & variables

SAT Scalability

Techniques for searching for answers improve over time

This graph describes the winning solver for each of the last 20 years, running on the same problems on the same hardware.

Benchmark suite contains 400 problems.

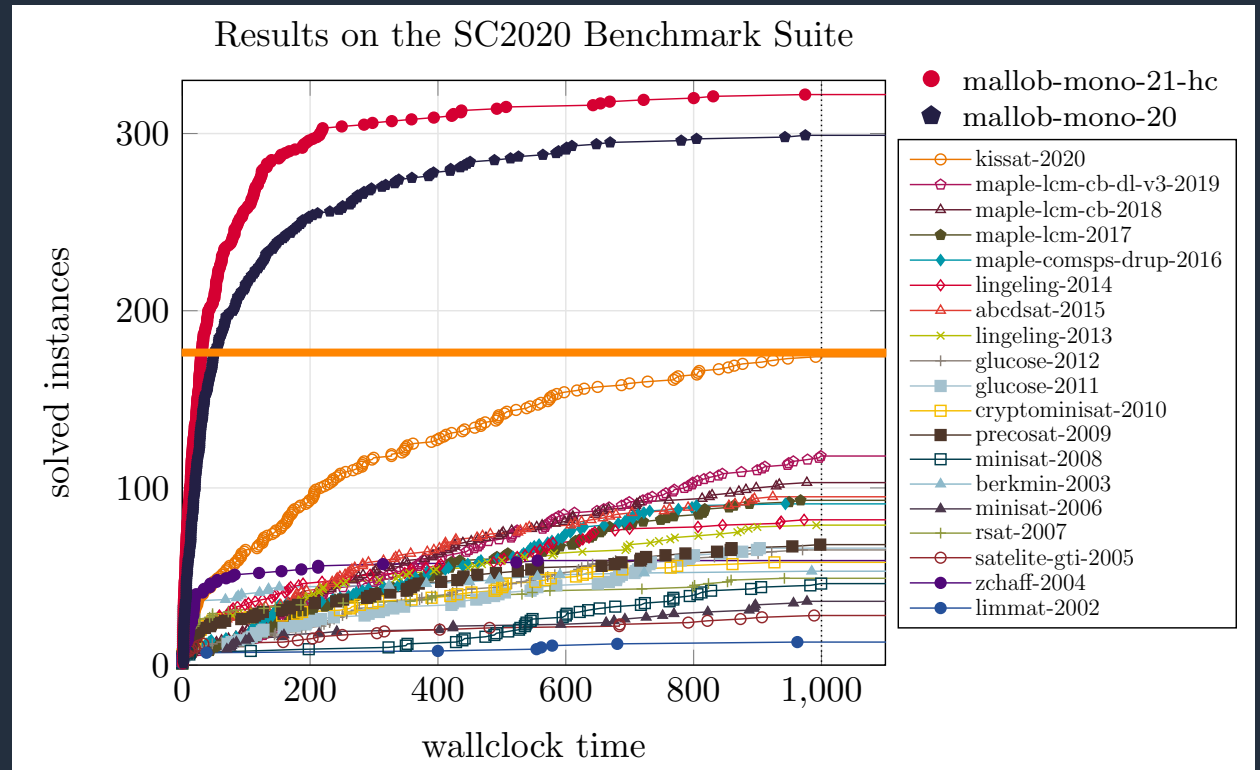


Distributed SAT Solvers

Distributed SAT solvers run multiple instances that work together

Still a new and improving technology

~30x faster than best sequential solver



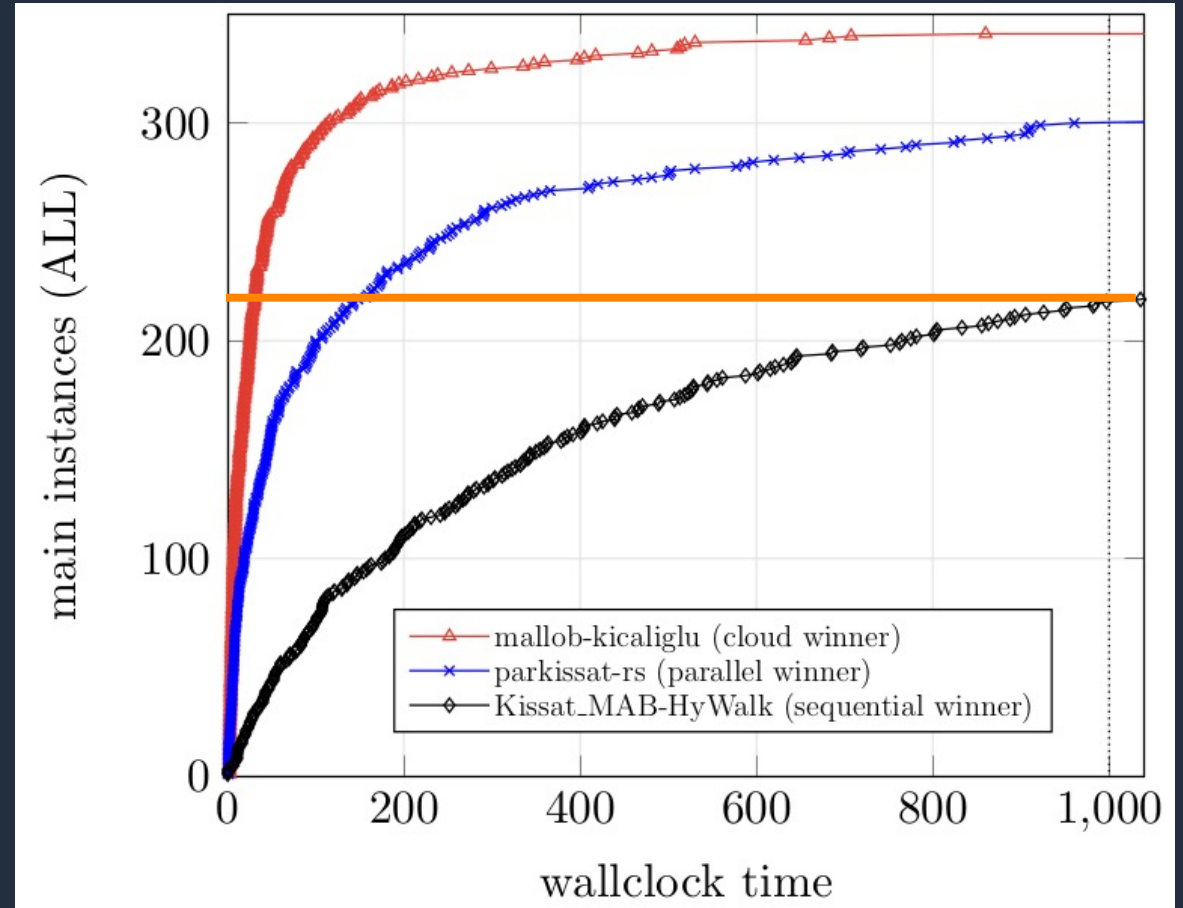
Distributed SAT Solvers

Distributed and parallel solvers improving more quickly than sequential solvers

2020-22: Distributed solver improvement: **56%**

Sequential solver improvement: **18%**

2022 SAT Competition



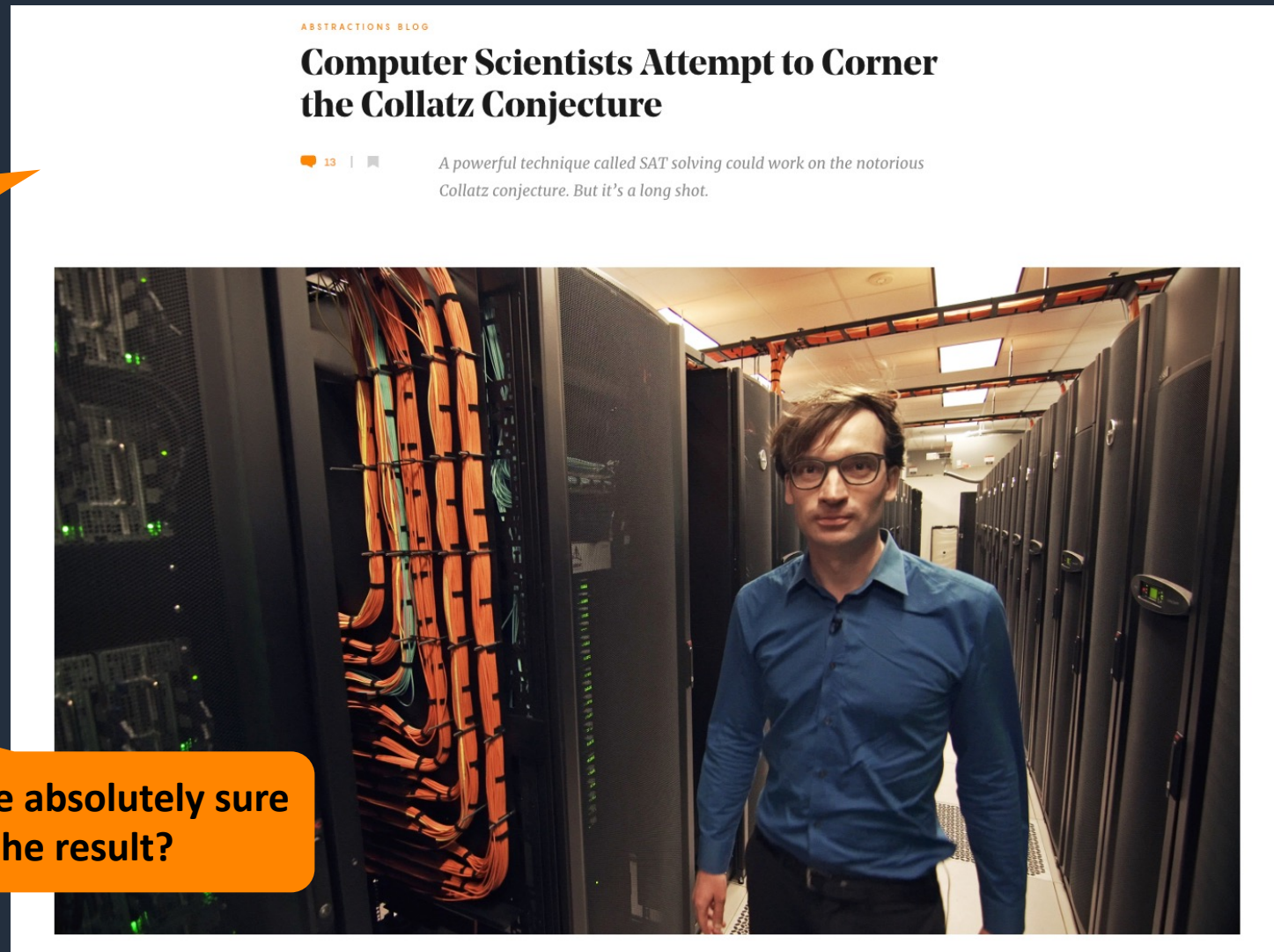
Using Distributed Solvers

Imagine a distributed SAT solver proves the Collatz Conjecture

This means the solver derives UNSAT for the negation of the conjecture

Fame! Glory!!
Accolades!!!

But are we absolutely sure of the result?

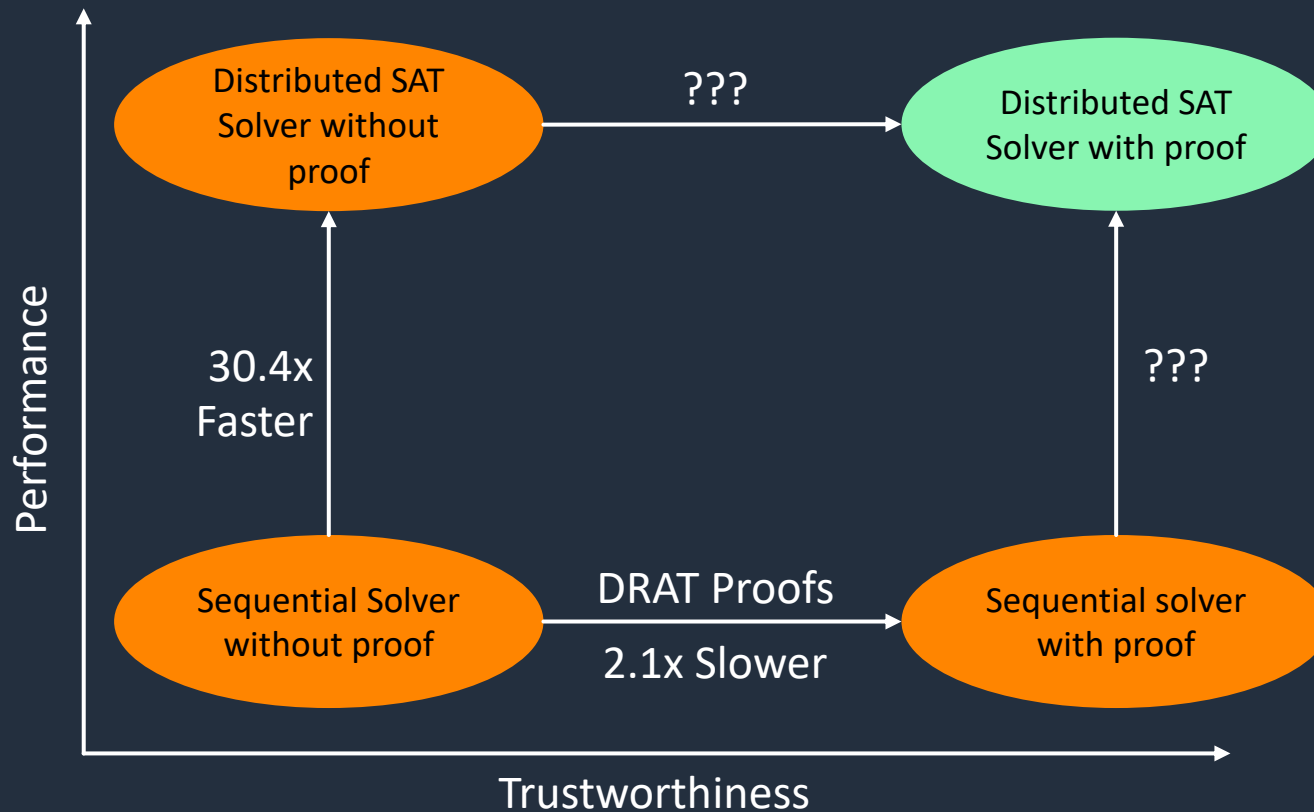


Trusting Results

Although distributed solvers are demonstrably the most powerful tools for solving hard SAT problems, there is an important caveat: unlike sequential solvers, current distributed clause-sharing solvers cannot produce proofs of unsatisfiability. While there has been foundational work in producing proofs for shared-memory clause-sharing SAT solvers [14], existing approaches are neither scalable nor general enough for large-scale distributed solvers. This is not just a theoretical problem—for four problems in the 2020 and 2021 SAT competitions, distributed solvers produced incorrect answers that were not discovered until the 2022 competition because they could not be independently verified.⁶

Unsatisfiability Proofs for Distributed Clause Sharing SAT Solvers, TACAS 2023

Solver Quadrants



Sequential Proof Format

Standard format: DRAT



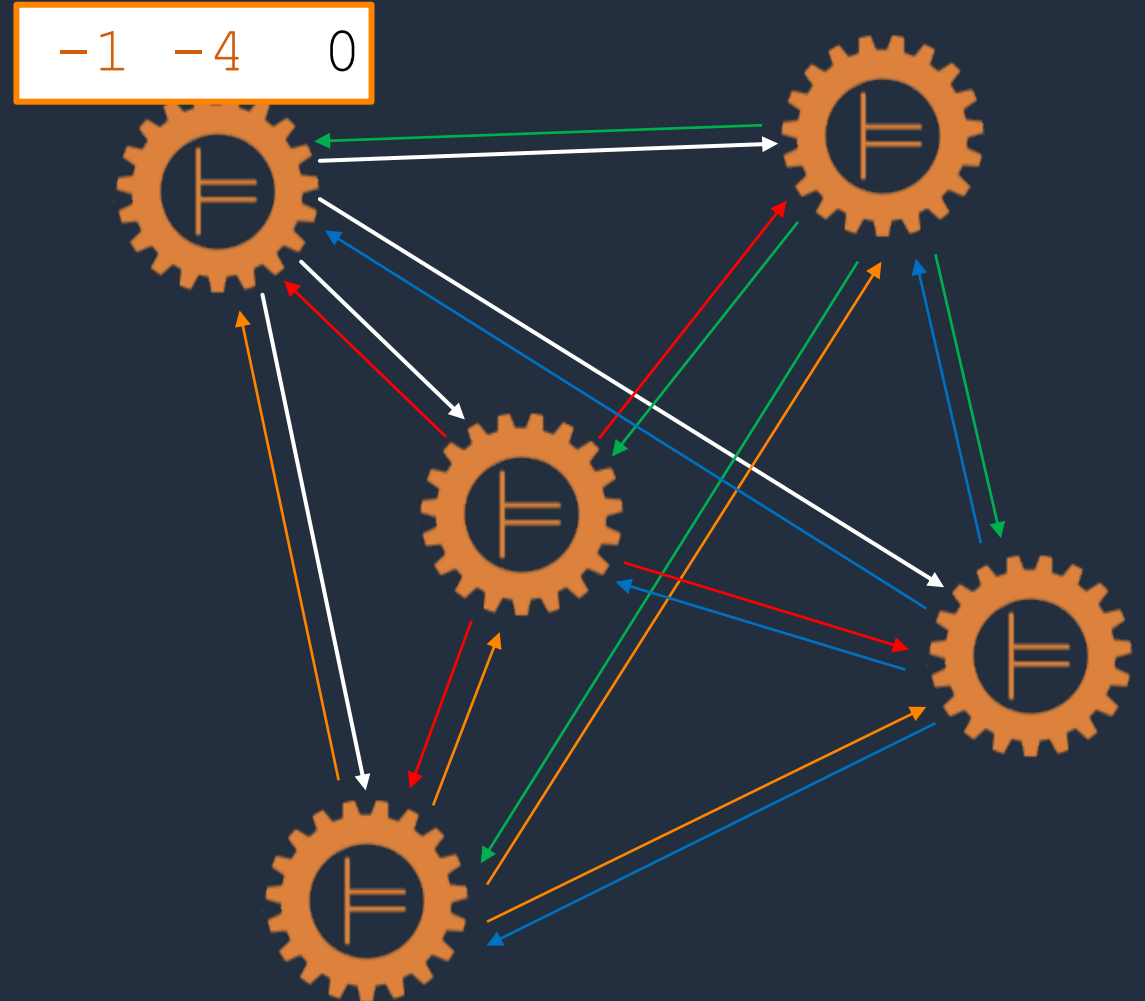
Original Problem

1	-2		0
2	-4		0
1	2	4	0
-1	-3		0
1	-3		0
-1	3		0
1	3	-4	0
1	3	4	0

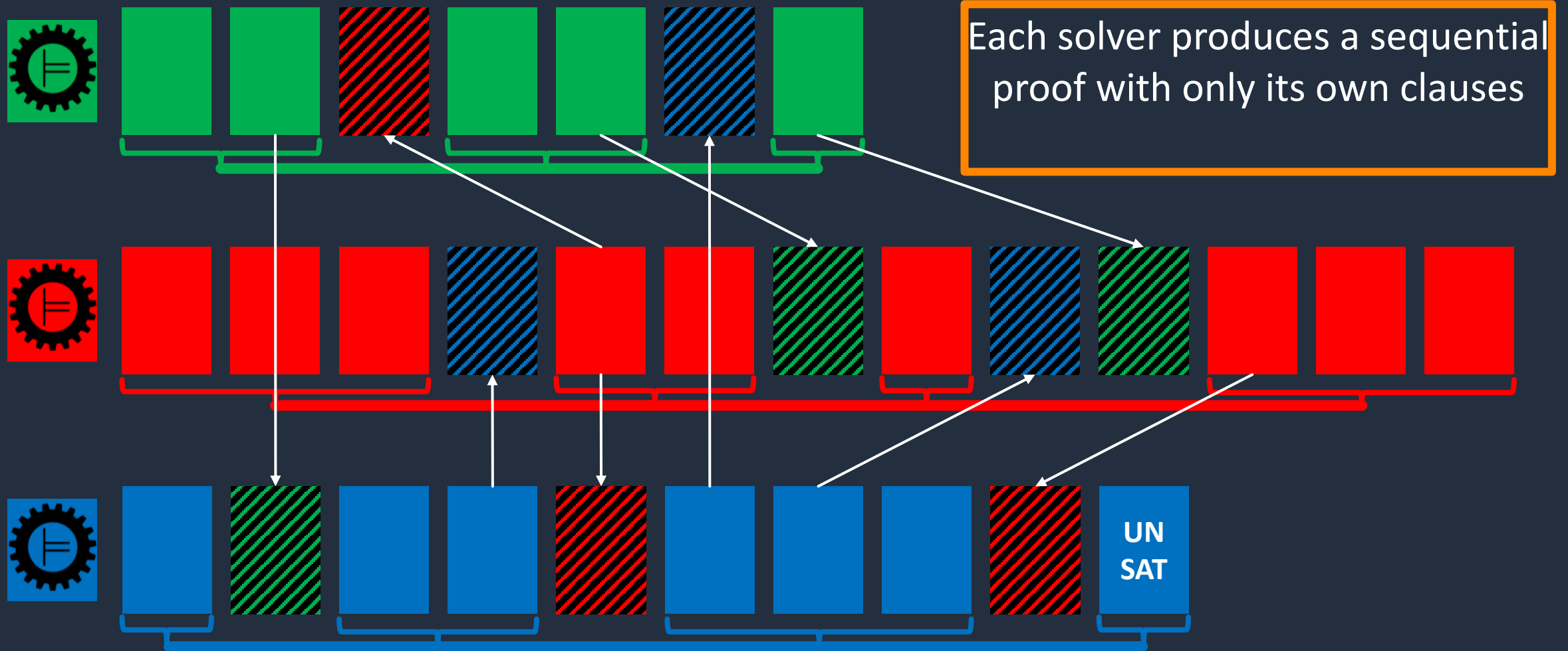
Distributed Solvers

Multiple solver instances running in parallel on same problem

Share learned clauses

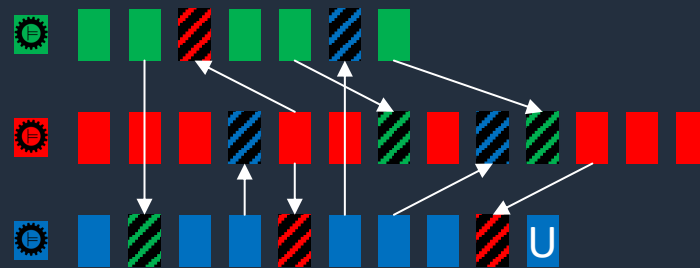


Distributed Solver Proof Challenges



Combining Proofs

- Need dependency order of clauses between solvers
- Need to order clauses from different proofs relative to each other



Combining Proofs

- Need dependency order of clauses between solvers
- Need to order clauses from different proofs relative to each other



Technical Challenges

- Provide metadata to identify which solver produced each learned clause
- Efficiently sort learned clauses in dependency order across all solvers
- Reduce proof size by removing unnecessary clauses, so that proof can be efficiently checked.

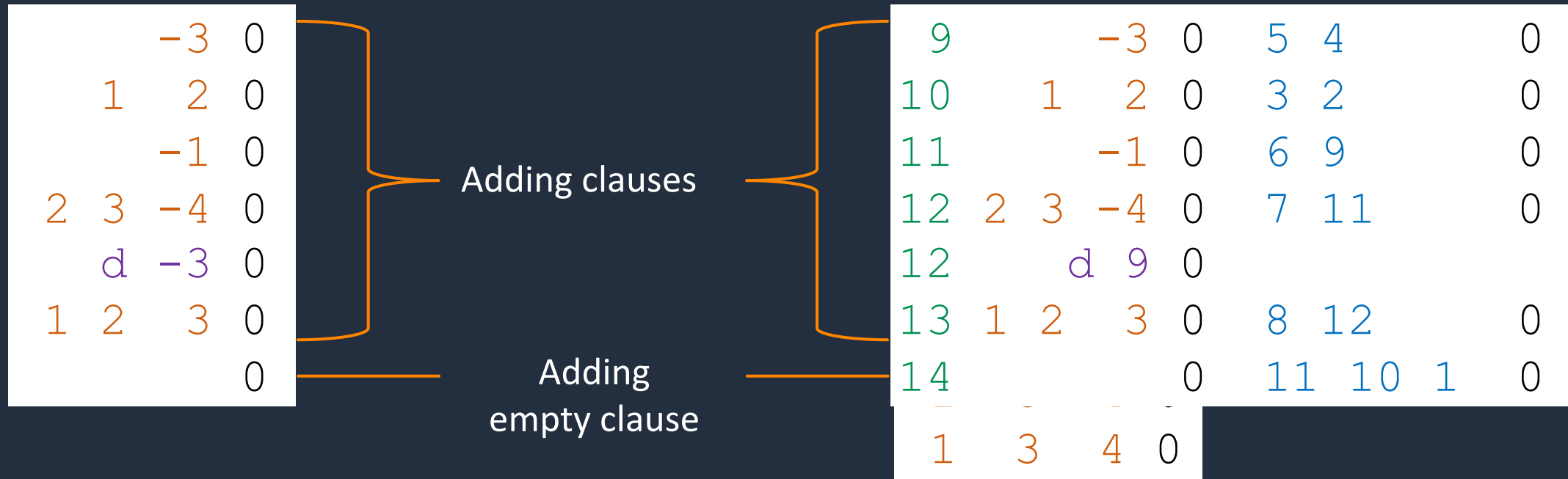
Sequential Proof Format

1
2
3
4

5
6
7
8

Standard format: DRAT

Original Problem LRAT



Distributed Clause ID

- Each clause needs unique ID across all instances
- Needs to be unique without communication
- Formula:

$$o + i + t * k$$

where

- o : original clauses
- i : current instance num
- t : total instances
- k : number of clauses learned in current instance

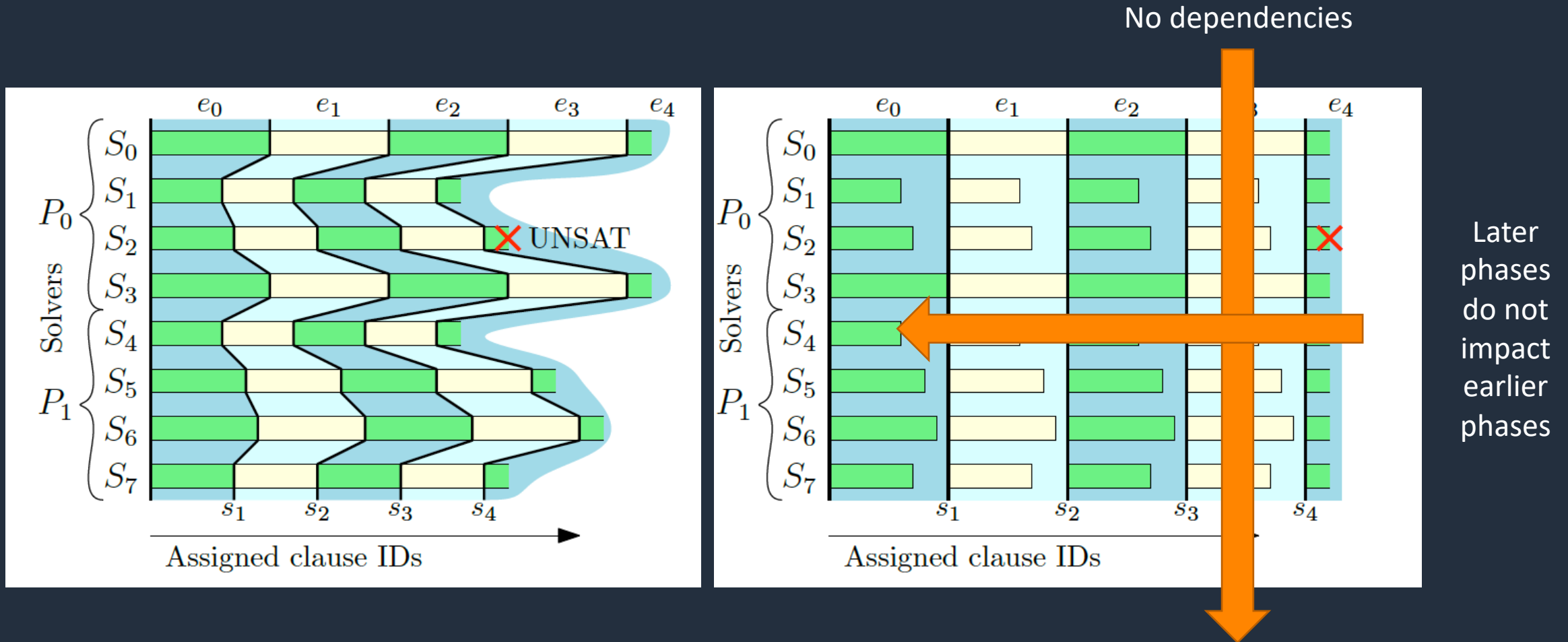
3 Instances
19 Original Clauses

<u>1</u>	<u>2</u>	<u>3</u>
20	21	22
23	24	25
26	27	28

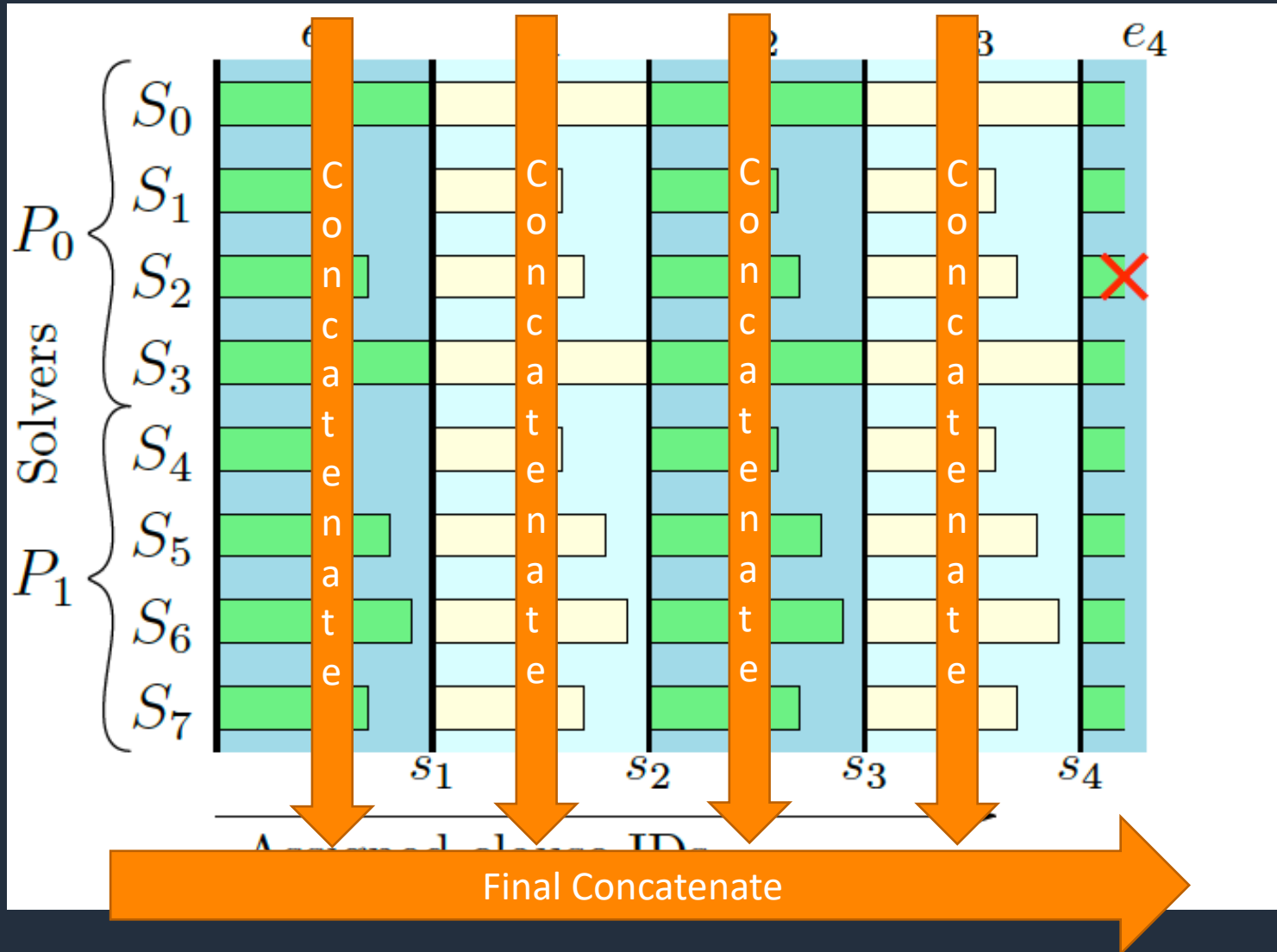
Technical Challenges

- Provide metadata to identify which solver produced each learned clause
 - Cross-Solver Unique Clause IDs used by LRAT format.
- Efficiently sort learned clauses in dependency order across all solvers
- Reduce proof size by removing unnecessary clauses, so that proof can be efficiently checked.

Using Epochs to Structure Proof Reconstruction



Naïve Approach



Because of epochs, could just merge all proofs together.

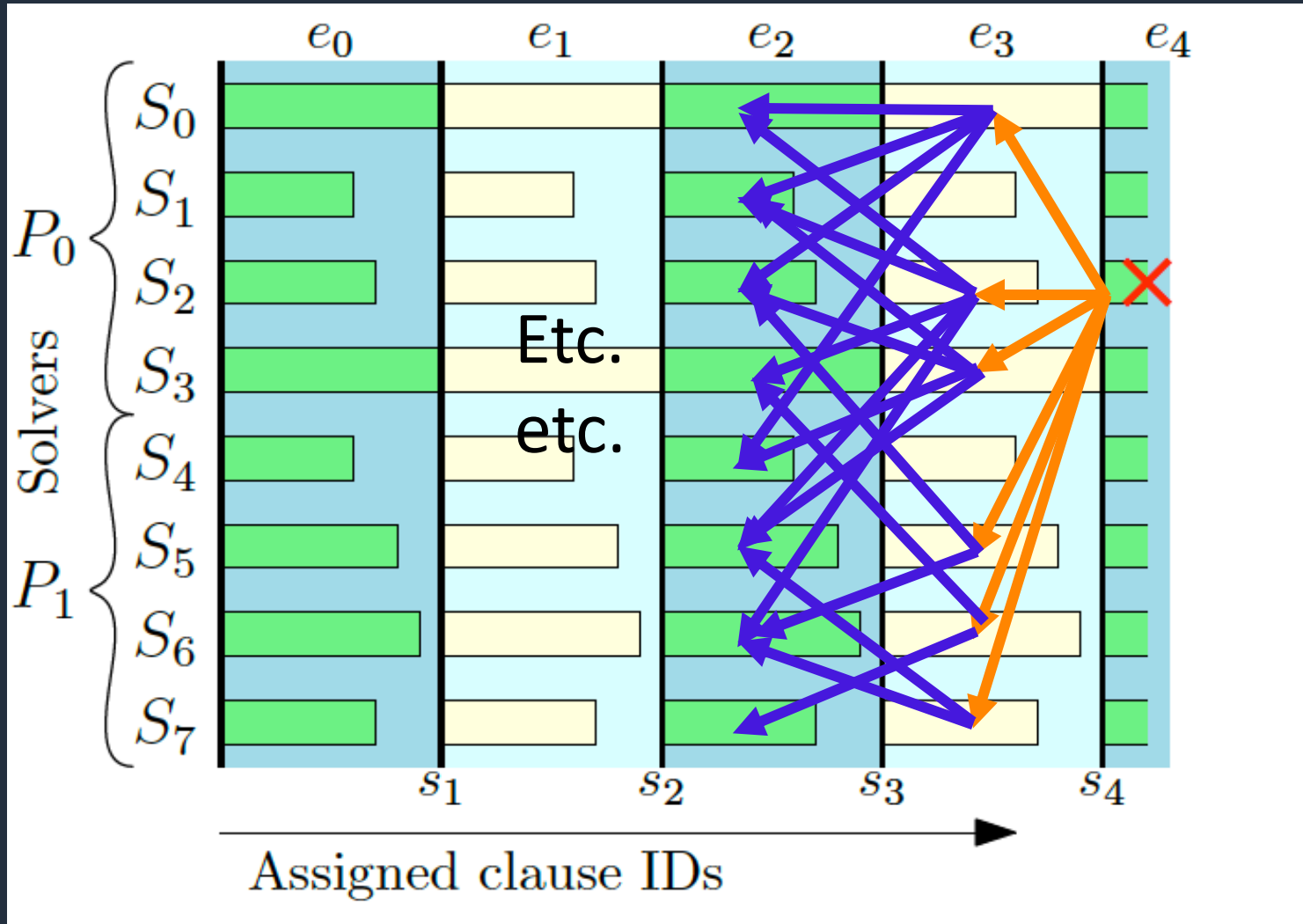
However, proofs are HUGE

Mean proof size: 65 GB

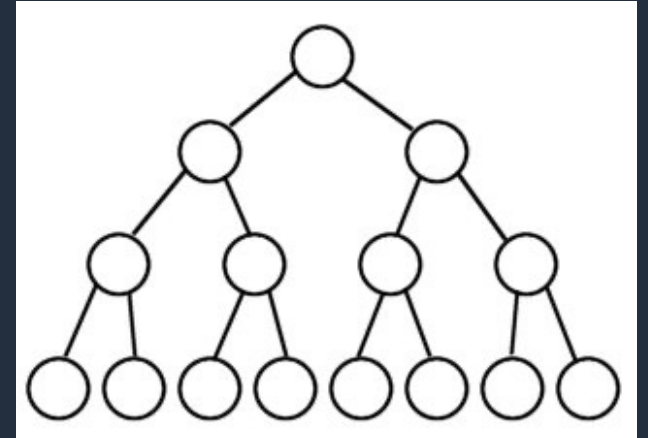
Largest proof size: >1 TB

Unreasonably large to check.

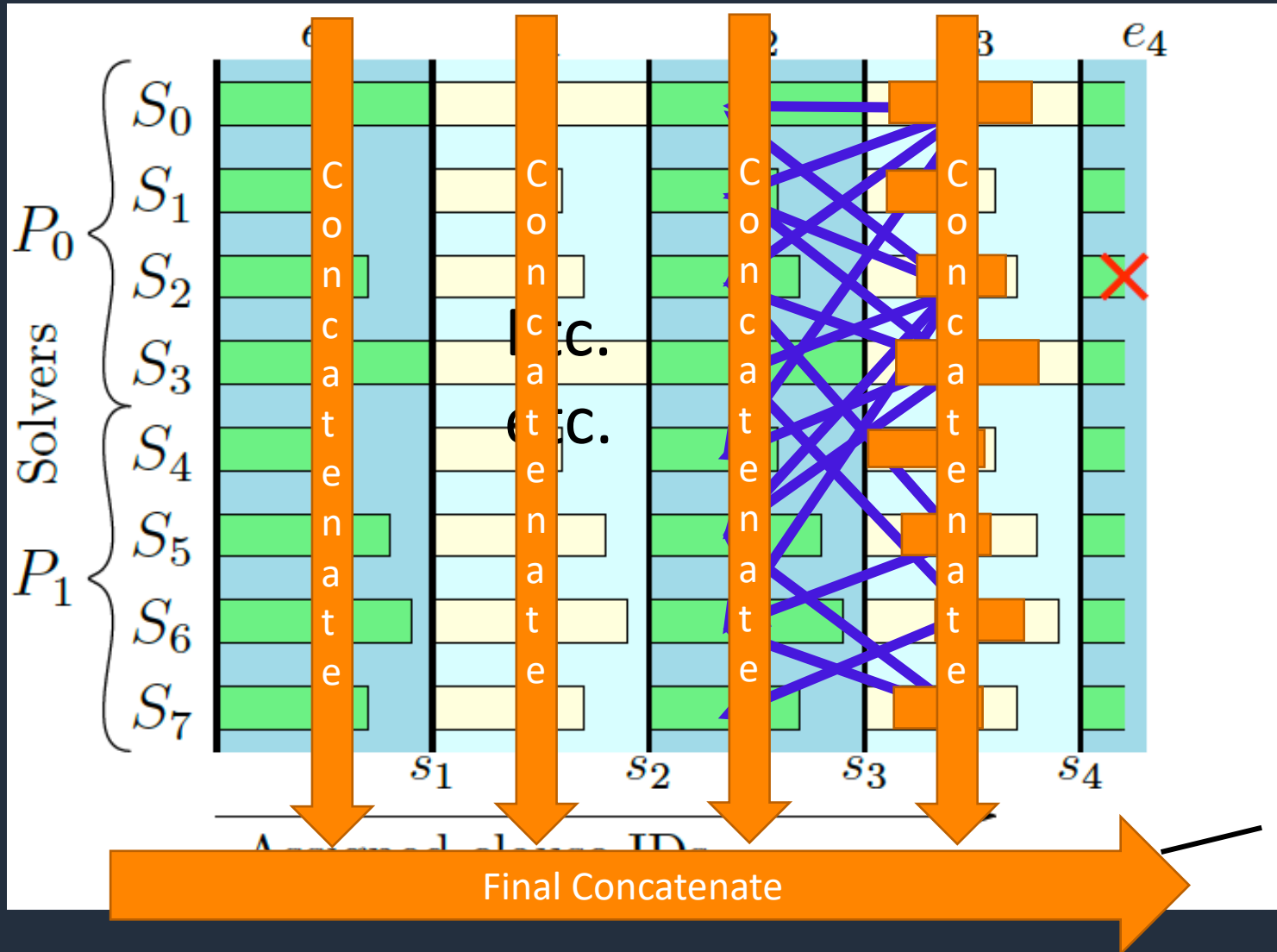
Using Epochs to Structure Proof Reconstruction



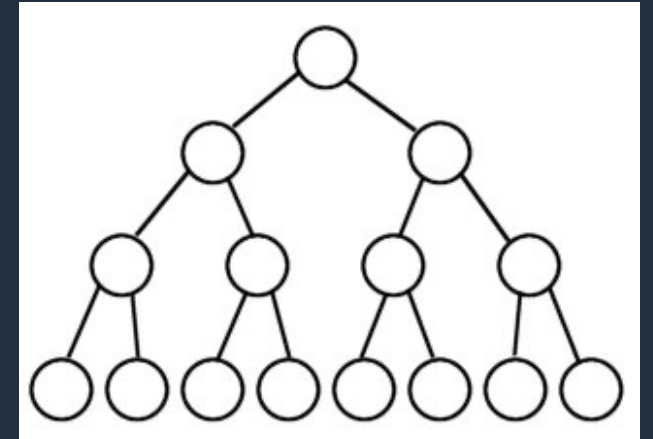
Removing duplicate requests for the same clause from multiple nodes: use same mechanism that Mallob uses for clause sharing.



Using Epochs to Structure Proof Reconstruction



Removing duplicate requests for the same clause from multiple nodes: use same mechanism that Mallob uses for clause sharing.

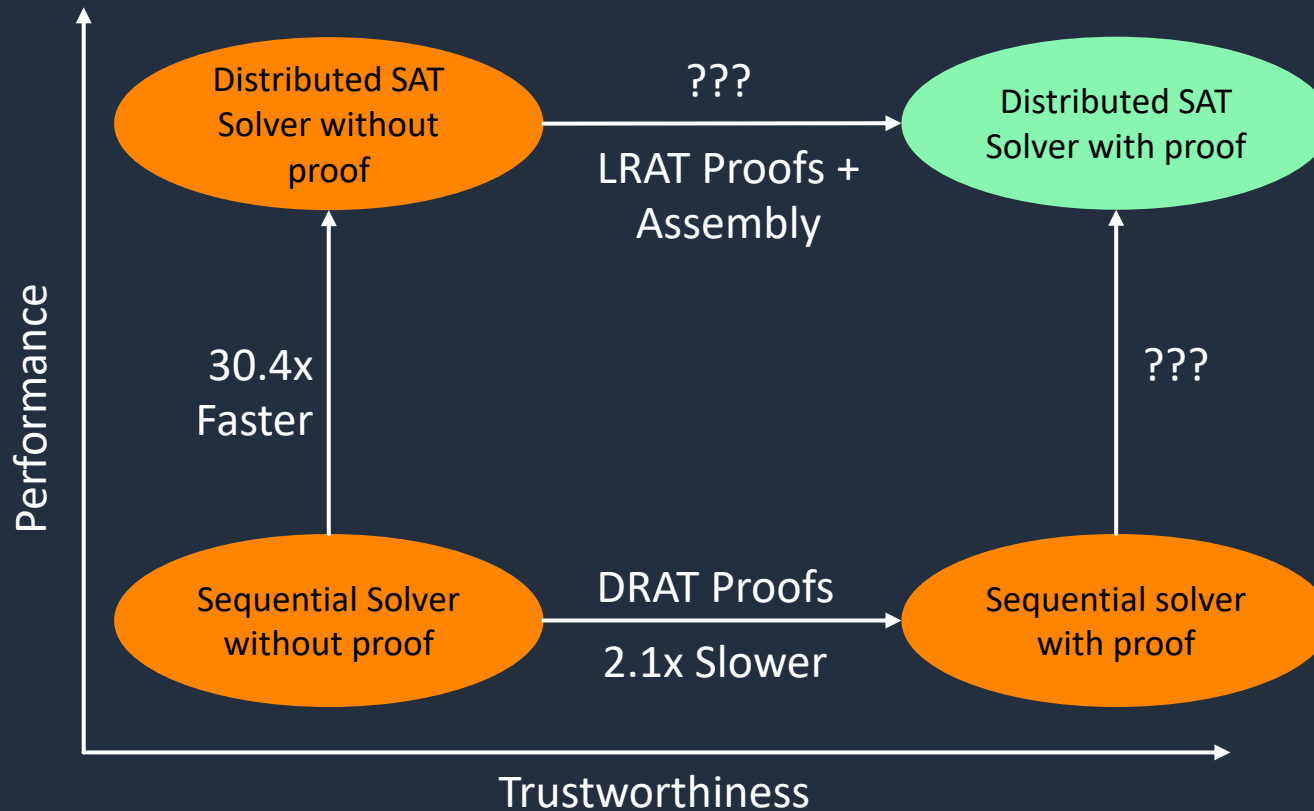


Mean proof size: 5.5 GB.
8% of Naive approach

Technical Challenges

- Provide metadata to identify which solver produced each learned clause
 - Cross-solver unique clause IDs used by LRAT format.
- Efficiently sort learned clauses in dependency order across all solvers
 - Use the same techniques used to share clauses to share proof dependencies
- Reduce proof size by removing unnecessary clauses, so that proof can be efficiently checked.
 - Work backwards from the empty clause to establish results

Solver Quadrants



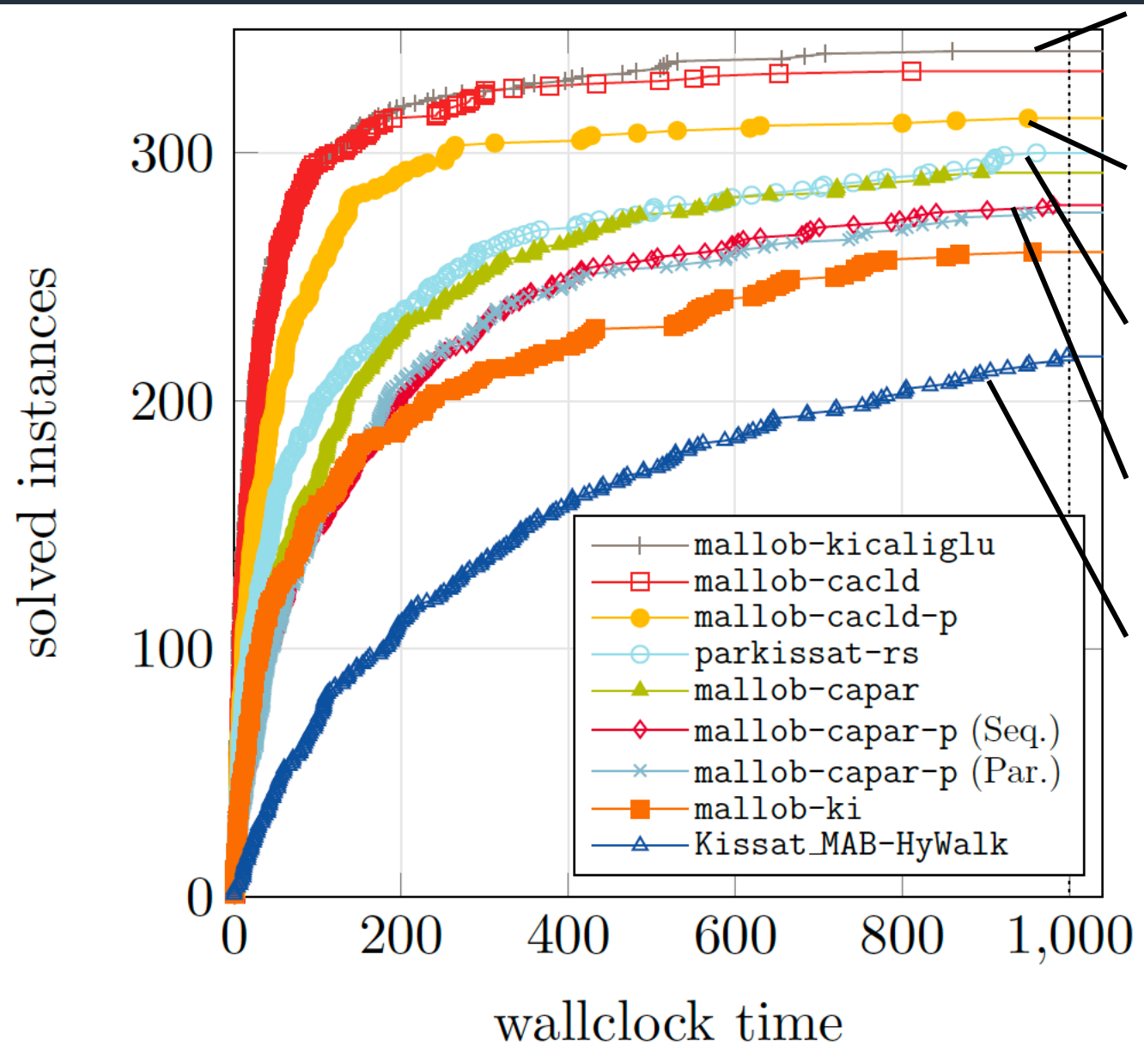
Experiment

- Run parallel and distributed solvers on SAT-Comp 2022 benchmarks (400 problems)
- Compare against the winners of the SAT-Comp 2022 tracks
 - Sequential track – running on m6i.4xlarge AWS instance w/16 cores, 64GB RAM
 - Parallel track - running on m6i.16xlarge AWS instance w/64 cores, 256 GB RAM
 - Distributed track – running on 100 * m6i.4xlarge AWS instances
- Track solving time and proof reconstruction time

Results: Solving

Distributed proof solver is ~1.8x slower (median) than best distributed solver

Distributed proof solver is ~17x faster (median) than best sequential solver



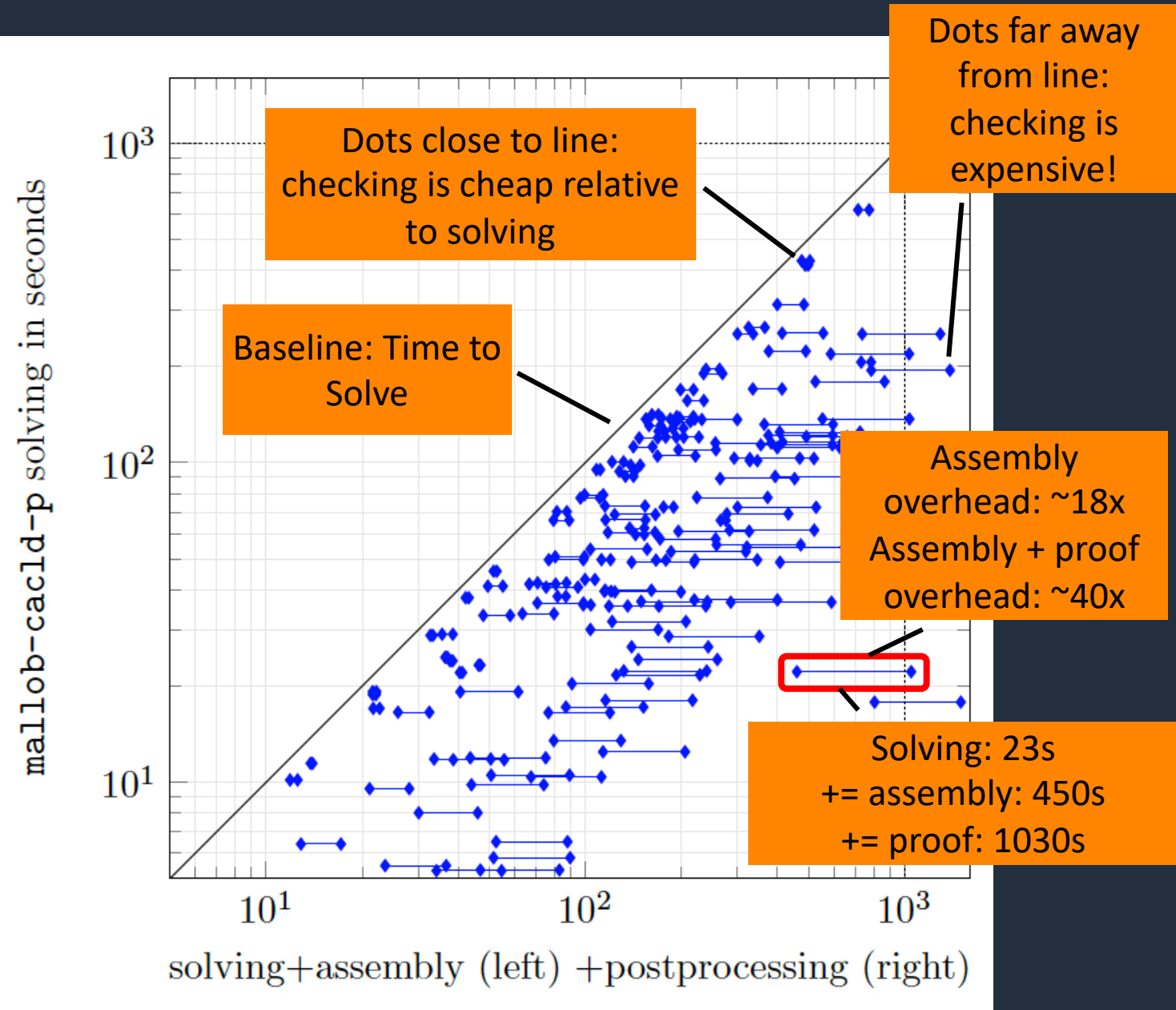
- Fastest Distributed Solver
- Distributed solver with proof
- Fastest Parallel Solver
- Parallel solver with Proof
- Fastest Sequential Solver

Results: Proof

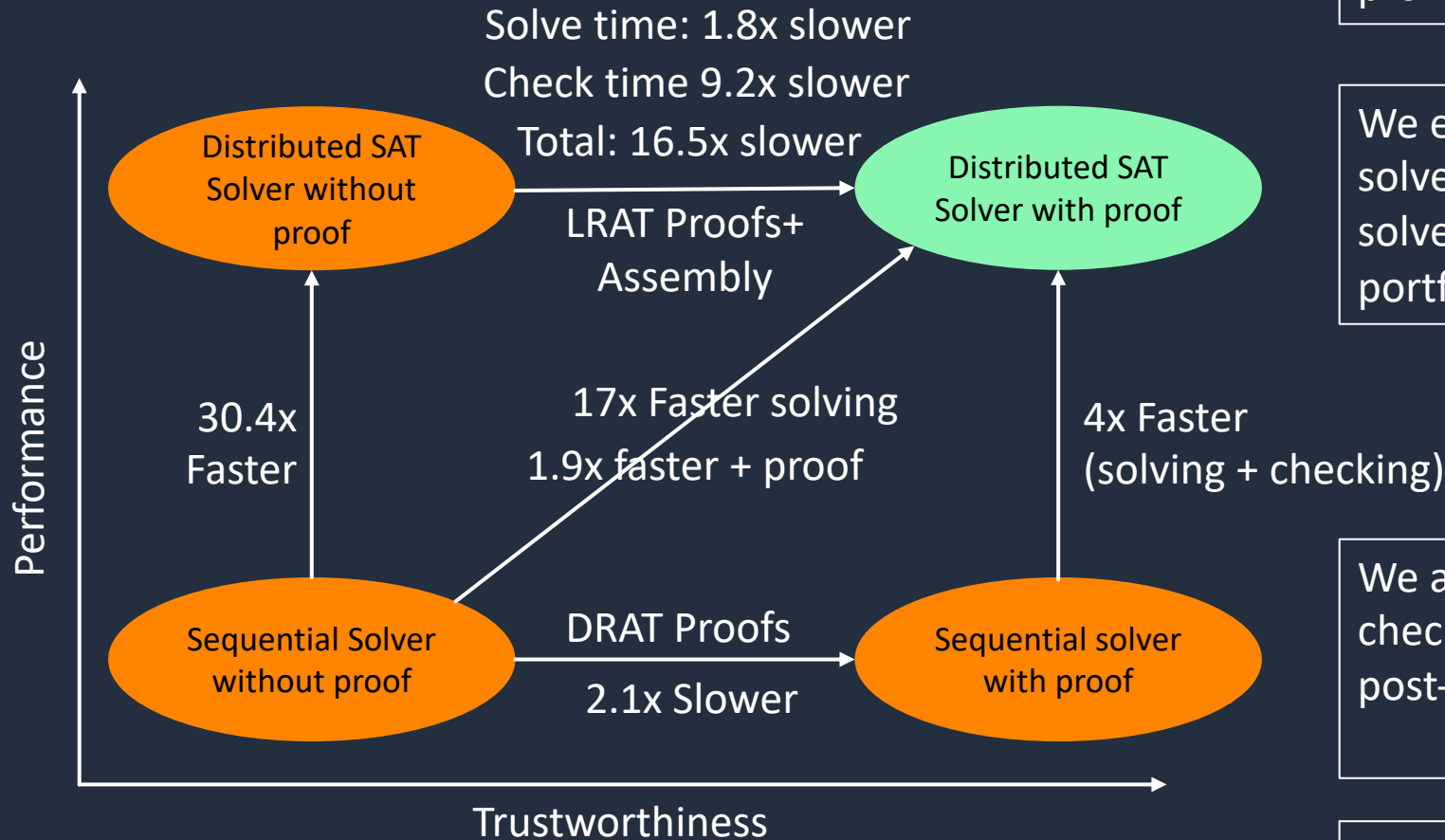
(Not shown) Average overhead for proof checking of sequential solver: 1.1x solve time

Average overhead for proof assembly and postprocessing (proof checking) for distributed solver: 5.1x solve time

Distributed solver is *relatively slower* at performing proofs than sequential solver and has *substantial variance* with solving time.



Solver Quadrants



Even though proof checking adds more overhead than sequential solving proof checking, the total time improvement for the distributed prover is 4x.

We expect to be able to improve solver performance once more solvers support LRAT to diversify our portfolio.

We also expect to be able to improve checker performance using a parallel post-processor and proof checker.

Near term goal:
10x solving + checking

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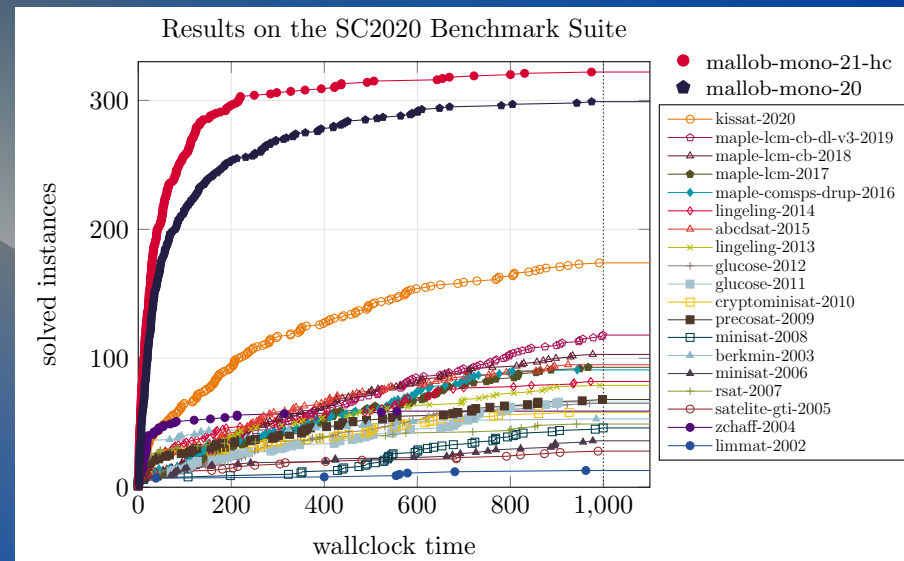


Thank you!

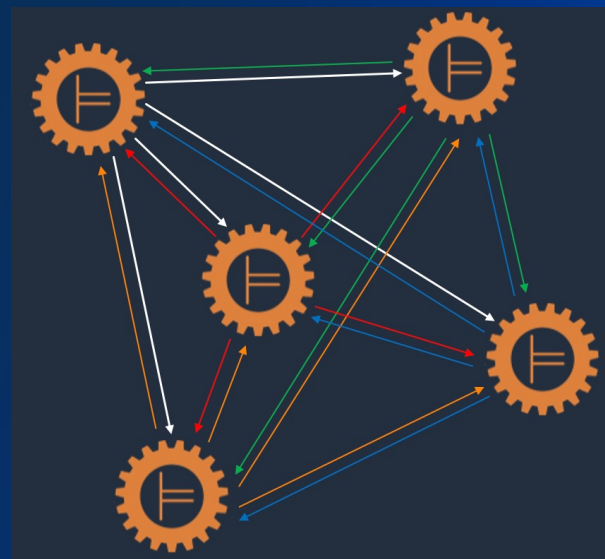
Mike Whalen

mww@amazon.com

Performance on Solving:



Clause Sharing:



Reconstruction Mechanism:

