

Evaluating the usefulness of Global Cardinality constraint propagators in Lazy Clause Generation

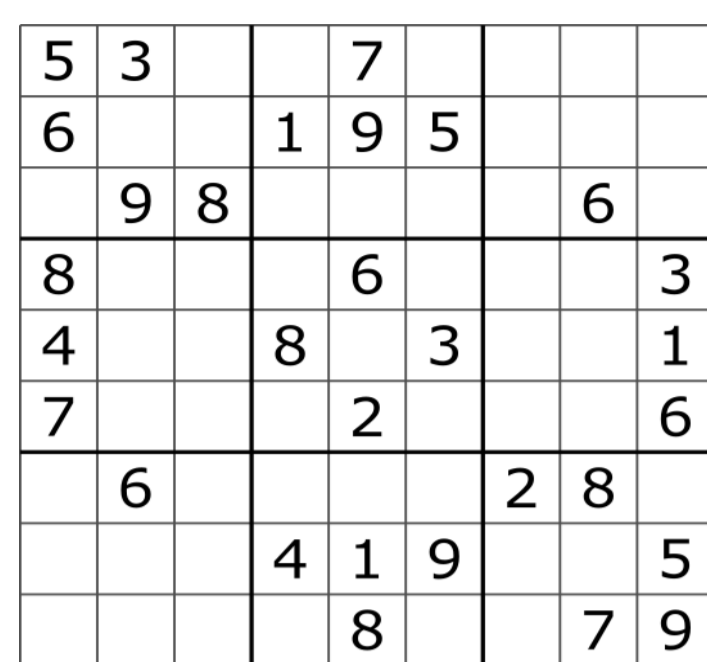


Figure 7: A Sudoku Puzzle, which can be solved using only GCCS
 Tim Steinhilber, CC0, via Wikimedia Commons

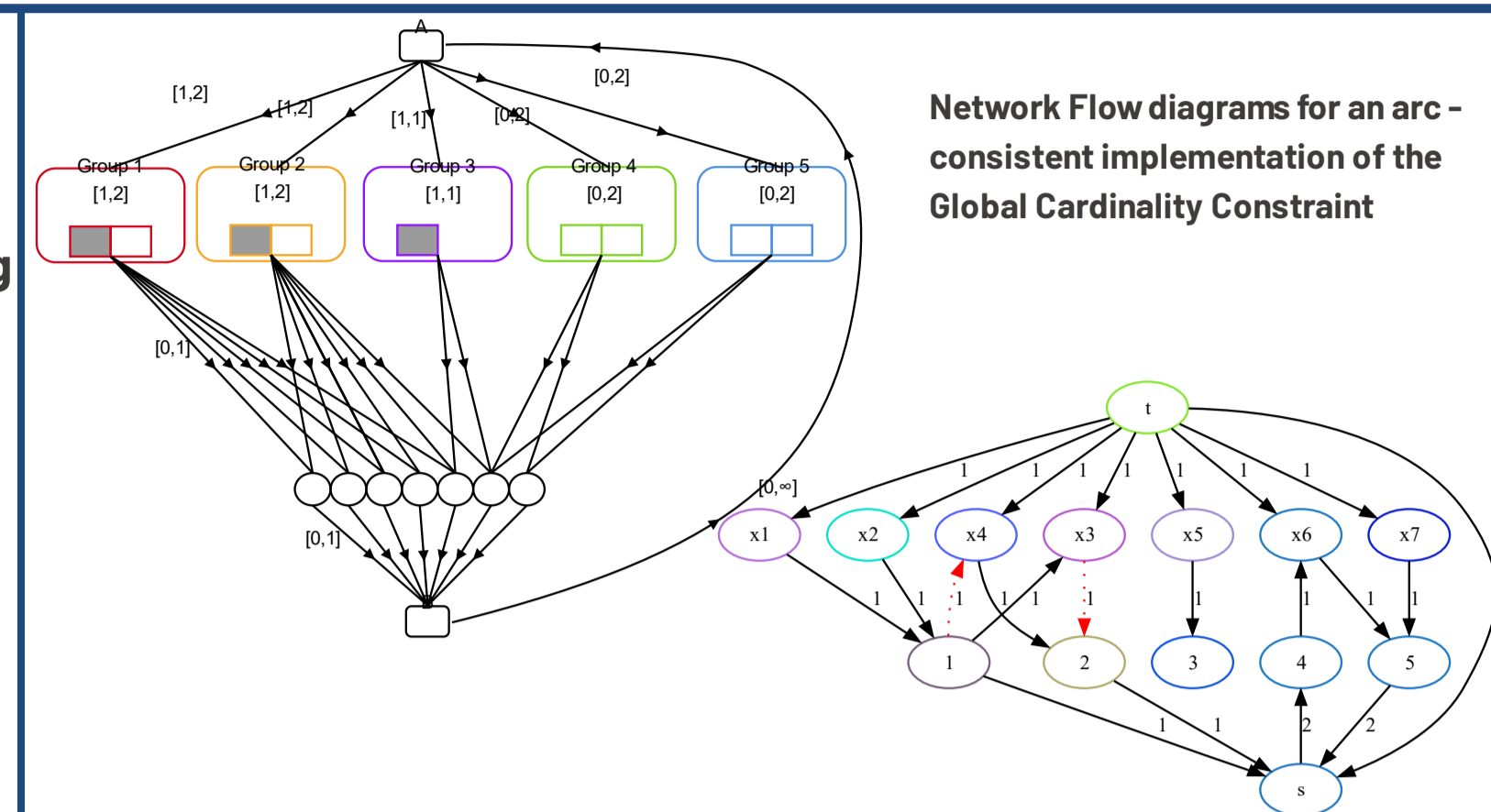
Sudoku can be solved using only the GCC

1 Introduction FDP Vs LCG

Finite Domain Propagation (FDP): Solve problems by modeling restrictions as constraints. Alternate between Search and Propagation.

Trade-off: runtime vs pruning strength [1]

Lazy Clause Generation (LCG): Newer, state-of-the-art. Combines Propagation + SAT Solver



Network Flow diagrams for an arc-consistent implementation of the Global Cardinality Constraint

GCC Vs Decomposition

2 Research question

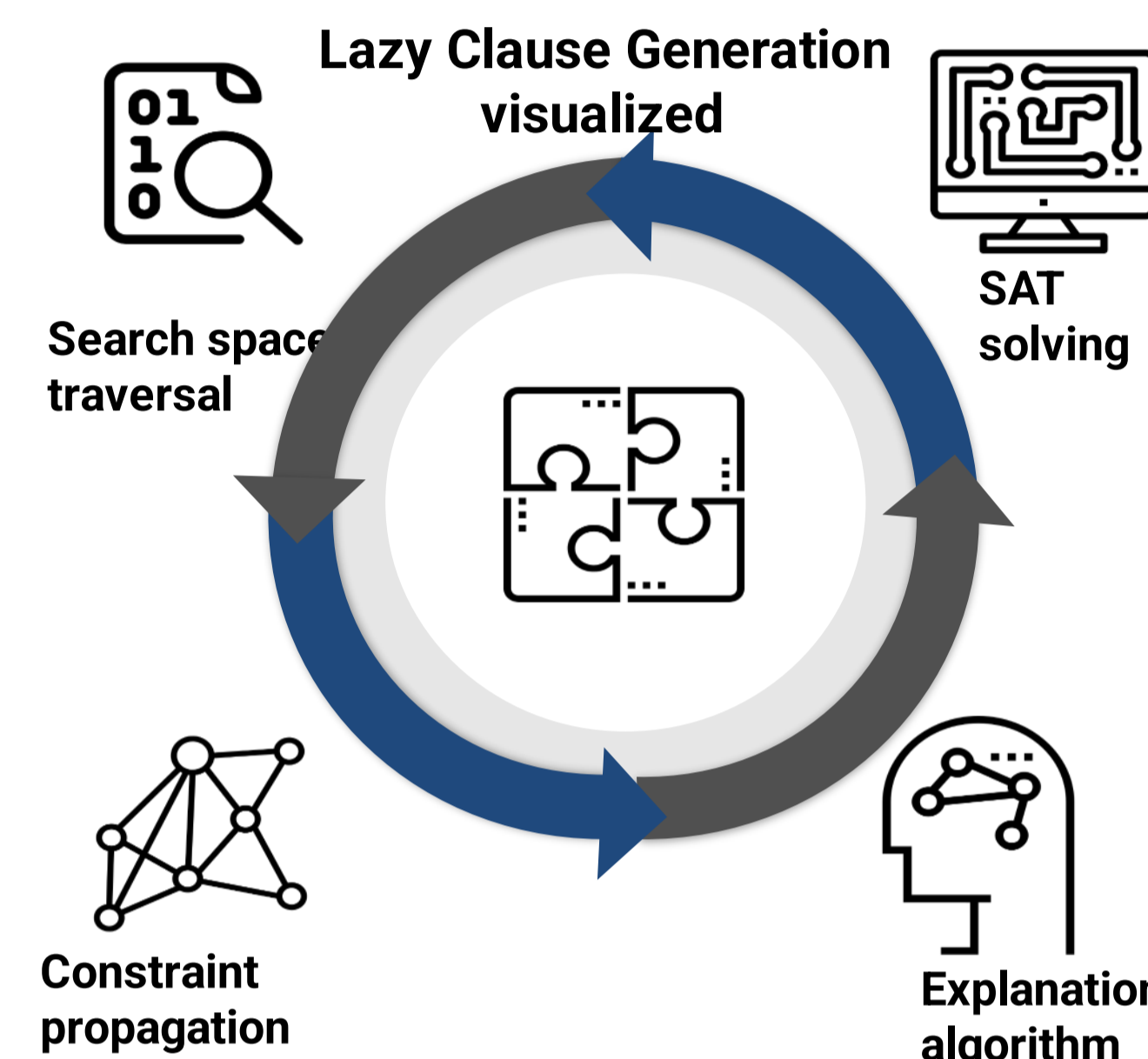
Global Cardinality Constraint (GCC): how many times values can be assigned to variables.

- Arises in real-world scheduling problems [2][3]
- GCC can be decomposed into smaller, simpler constraints.

LCG questions need for constraints like the GCC

- Decompositions can yield state-of-the-art solutions for some constraints. [1][4]
- **But:** Little research about implementing GCC propagators for LCG

Question: Can a GCC propagator in LCG compete against decomposition?



3 Methodology

Two propagators and, two explanations are benchmarked using the Minizinc Challenge [5] and Sudoku against decomposition.

Propagator algorithms: Explanation algorithms:

- Regin GAC [6]
- Basic Filter
- Naive
- Purpose-built

Runtime Vs Pruning strength

4 Results

- Geometric mean increase ratio (compared to decomposition) of Regin GAC and Basic filter

Sudoku:

(Regin, Basic Filter)	Speedup	LBD	LCL	ACS
General expl	(3.42, 1.52)	(0.90, 1.05)	(0.91, 0.95)	(0.92, 0.95)
Purpose built expl	(3.19, 2.28)	(0.85, 1.00)	(0.91, 1.03)	(0.93, 0.84)

Minizinc challenge:

(Regin, Basic Filter)	Speedup	LBD	LCL	ACS
General expl	(1.34, 0.91)	(0.90, 1.05)	(0.91, 0.95)	(0.92, 0.95)
Purpose built expl	(1.36, 1.00)	(0.85, 1.0)	(0.91, 1.03)	(0.93, 0.84)

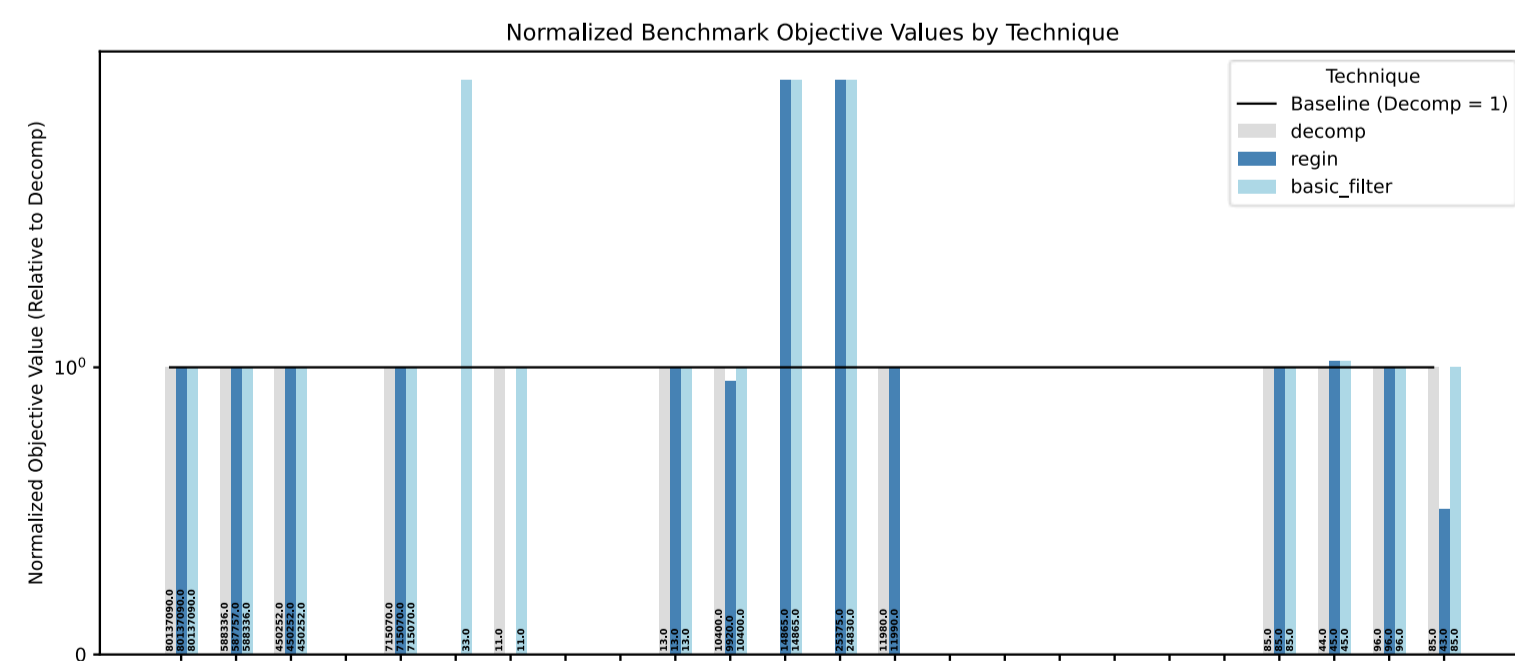
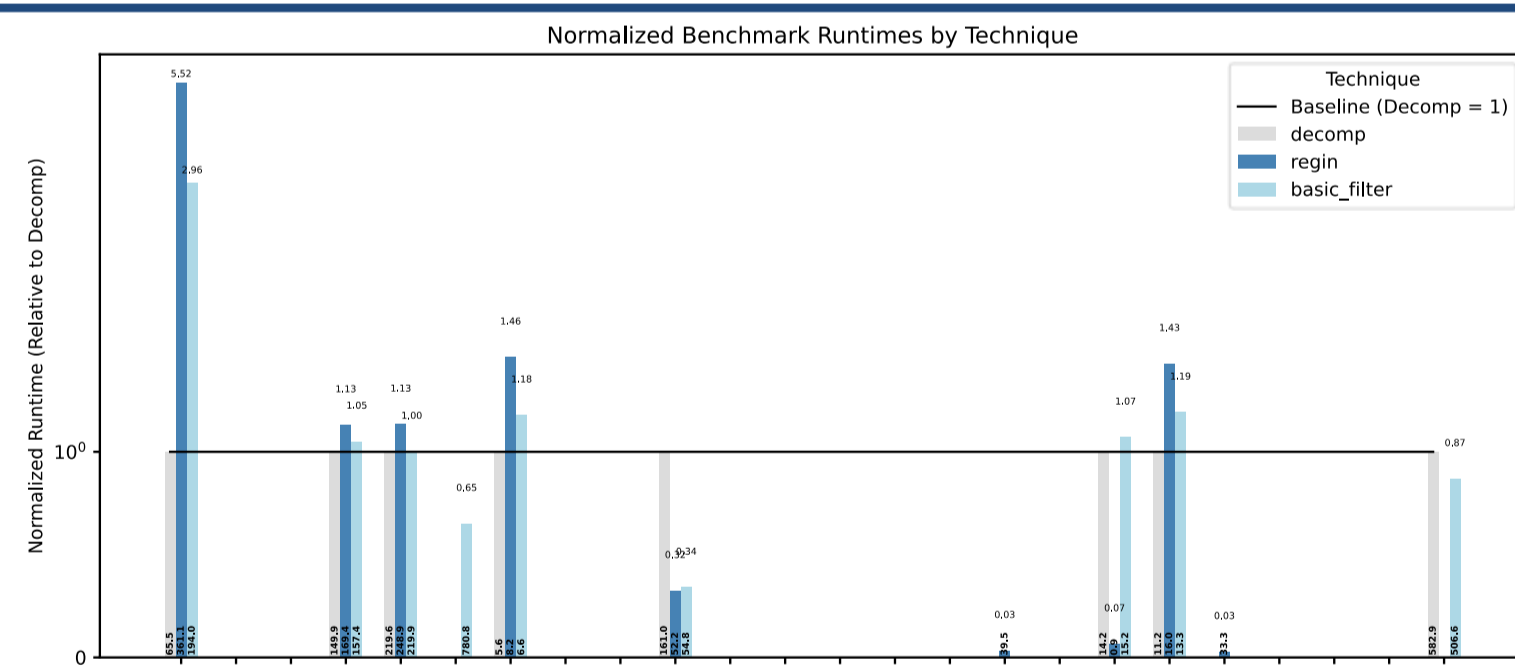
$$E_{naive}(X) = \bigwedge_{v \in X} \left([x \geq lb(x)] \wedge [x \leq ub(x)] \wedge \bigwedge_{h \in H(x)} [x \neq h] \right)$$

$$E_{filt1}(X, v) = \left(\bigwedge_{x \in X} [x = v] \text{ if } x = v \right)$$

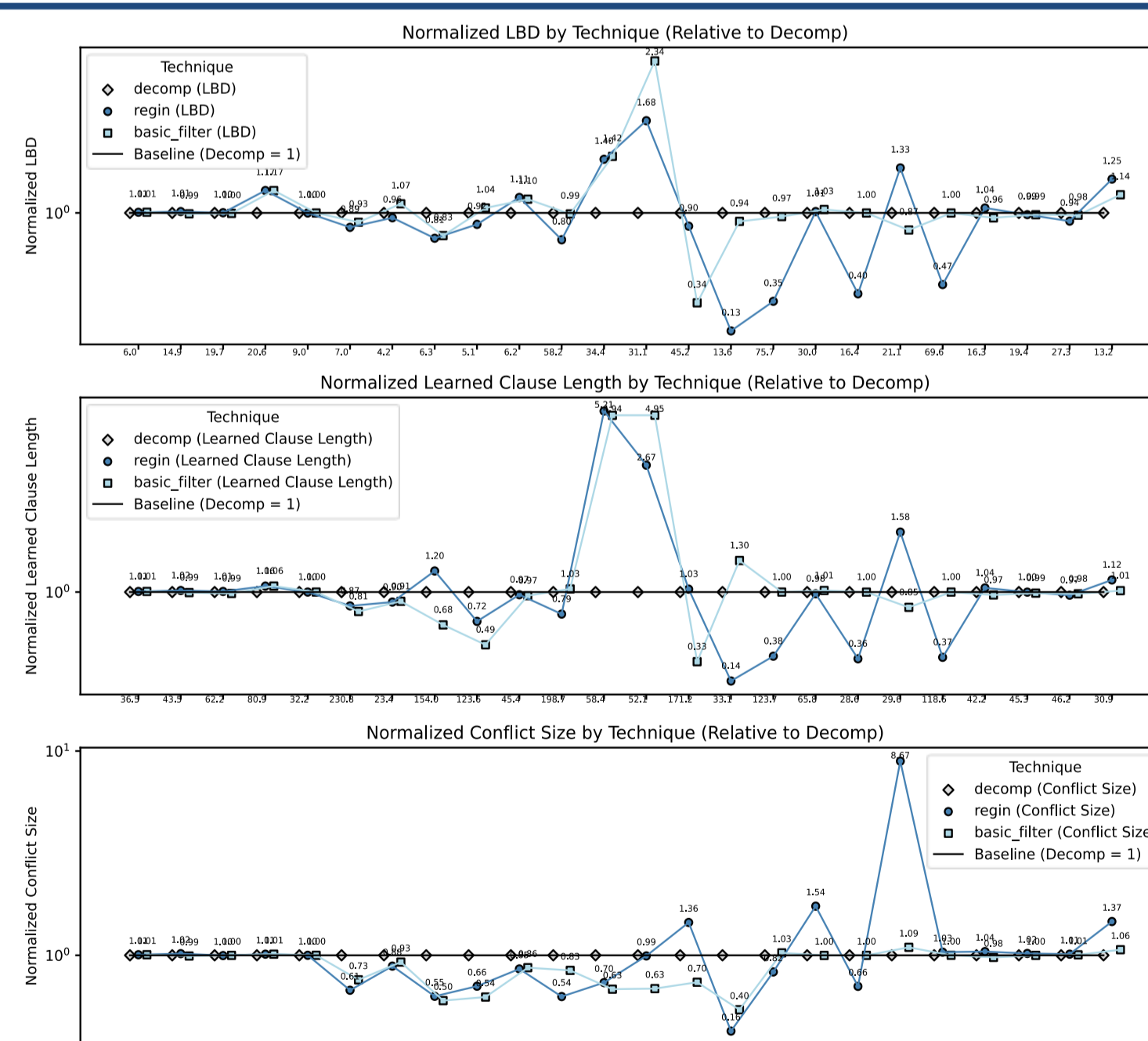
$$E_{filt2}(X, v) = \left(\bigwedge_{x \in X} [x \neq v] \text{ if } v \notin D(x) \right)$$

$$E_{R_{\text{regin}}}(x, v, X, V, G_0, G) = \bigwedge_{(v', x') \in G_0} [x' \neq v'] \text{ if } (v', x') \notin G \wedge J(x', v', R(G), x, v)$$

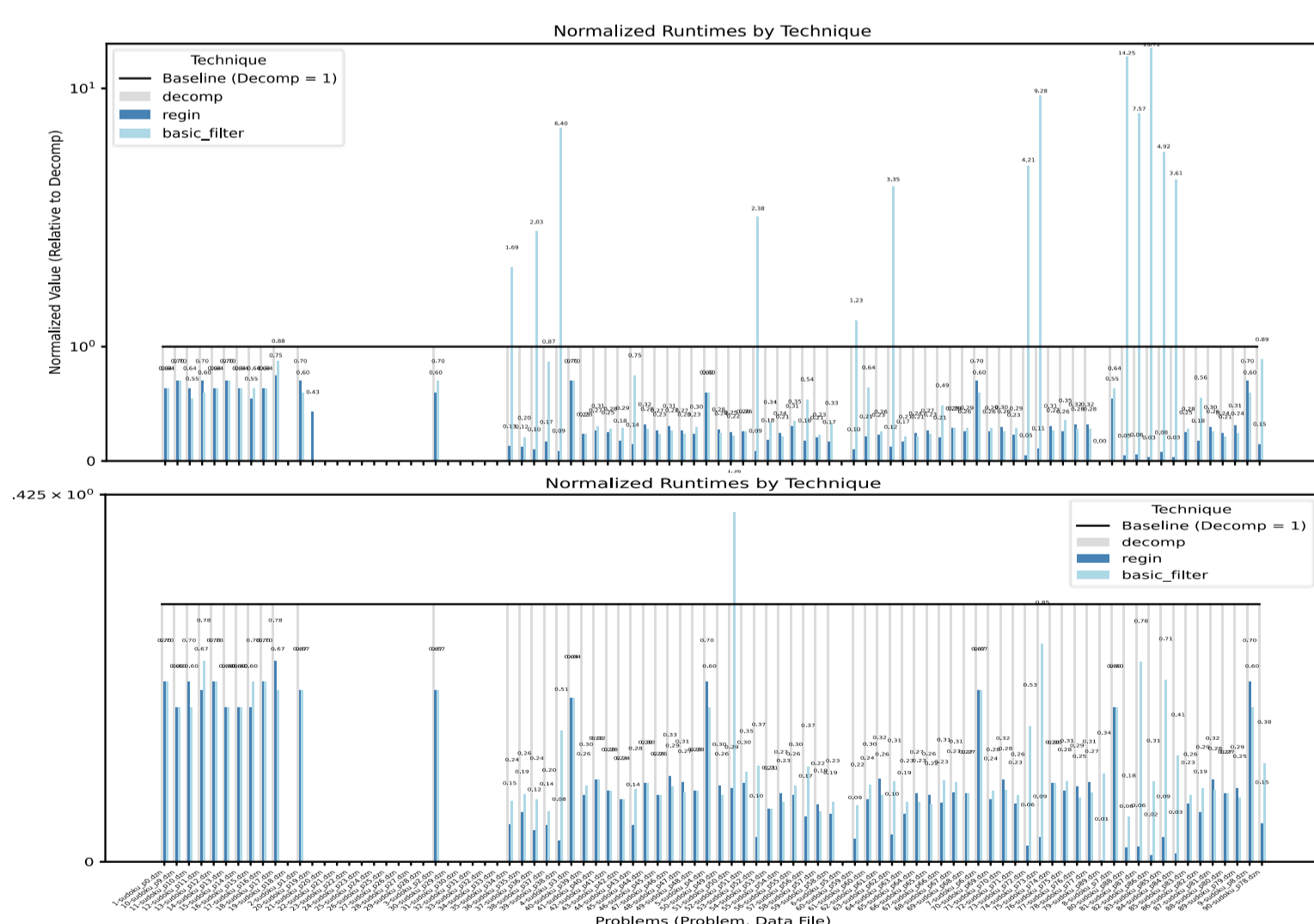
The explanation algorithms: General explanation, Basic Filter explanation and Regin explanation



Minizinc benchmark, relative runtime (decomp = 1) of Regin GAC and Basic Filter using new explanation algorithms



Minizinc challenge relative (decomp = 1) LBD, LCL and ACS of Regin GAC and Basic filter, using new explanation algorithms



Sudoku benchmark relative runtime (decomposition = 1). Top: Naïve explanation. Bottom: Specialized explanation

5 Conclusion

- Regin GAC and Basic Filter superior against decomposition on Sudoku
- Regin GAC wins in runtime against decomposition on Minizinc challenge. Basic Filter roughly equal.
- New explanation greatly helps Basic Filter. Does not help Regin GAC as much because of high runtime cost and the clause length being larger than it should be
- New explanations have a 5% LBD reduction compared to the naïve one in both Sudoku and Minizinc challenge. How much does GCC contribute to LBD?
- Regin GAC achieved a lower (better) LBD compared to decomposition (85%) using new explanation.

References

[1] Stuckey, P. J. (2010). Lazy clause generation: Combining the power of SAT and CP (and MIP?) solving. *Lecture Notes in Computer Science (Including Subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*, 6140 LNCS. https://doi.org/10.1007/978-3-642-13520-0_3

[2] Gebinger, T., Kietzlander, L., Krainz, M., Mischek, F., Muslu, N., & Winter, F. (2021). Physician Scheduling During a Pandemic. *Lecture Notes in Computer Science (Including Subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*, 12735 LNCS. https://doi.org/10.1007/978-3-030-73292-6_29

[3] Maslu, N., Schutt, A., & Stuckey, P. J. (2018). Solver independent rotating work force scheduling. *Lecture Notes in Computer Science (Including Subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*, 10848 LNCS. https://doi.org/10.1007/978-3-319-93031-2_31

[4] Schutt, A., Feydy, T., Stuckey, P. J., & Wallace, M. G. (2009). Why cumulative decomposition is not as bad as it sounds. *Lecture Notes in Computer Science (Including Subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*, 5732 LNCS. https://doi.org/10.1007/978-3-642-04244-7_58

[5] Stuckey, P. J., Feydy, T., Schutt, A., Tack, G., & Fischer, J. (2014). The Minizinc challenge 2008/2013. In *AI Magazine* (Vol. 35, Issue 2). <https://doi.org/10.1609/aimag.v35i2.2539>

[6] Regin, J. C. (1996). Generalized arc consistency for global cardinality constraint. *Proceedings of the National Conference on Artificial Intelligence*, 1.

Image References:

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