



MaxSAT-Based Bi-Objective Boolean Optimization

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Constraint Reasoning and Optimization Research Group



INTRODUCTION

MaxSAT-Based Bi-Objective Boolean Optimization

Paper under review at international scientific conference (SAT'22)

- 1 Motivation
- 2 Contribution (BIOPTSAT)
- 3 Results
- 4 Take away points



MOTIVATION — OPTIMIZATION

Task: Choose the cheapest flat with at least two rooms

What if...

... we want to take commute into consideration as well?

Flat	Rooms	Price
A	2	300 000 €
B	2	240 000 €
C	1	180 000 €
D	2	270 000 €



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MOTIVATION — MULTIPLE OBJECTIVES

Task: Choose a flat with at least two rooms that is as cheap as possible while also having **as short of a commute as possible**.

Optimality for Multiple Objectives

There is no single definition of optimality for multiple objectives!

Pareto Optimality

All solutions for which no other solution is **clearly** better are considered optimal

Flat	Rooms	Price	Commute
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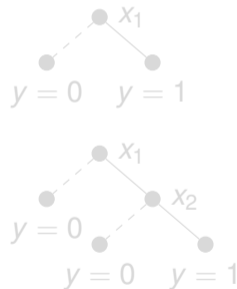
MOTIVATION — IMPLICIT SOLUTION SET

Task: From all valid decision trees, find the smallest ones that also minimize classification error

Hard Problems

Implicit solution definitions are \mathcal{NP} -hard for many real-world problems

x_1	x_2	y
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0	1	0
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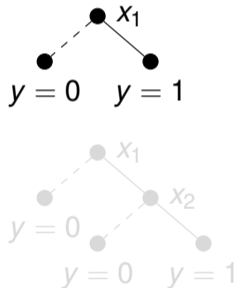
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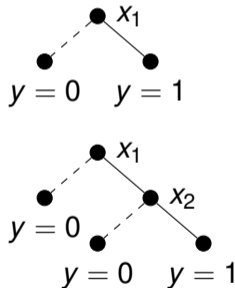
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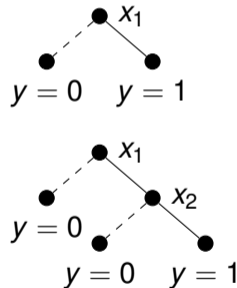
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THESIS CONTRIBUTION

- Development of the BIOPTSAT algorithm
 - Enumeration of exact Pareto-optimal solutions to hard problems
- Implementation of the BIOPTSAT algorithm and its competitors (will be released open source)
- Evaluation of variants of BIOPTSAT and comparison to two competitors
- Evaluation of refinements to BIOPTSAT



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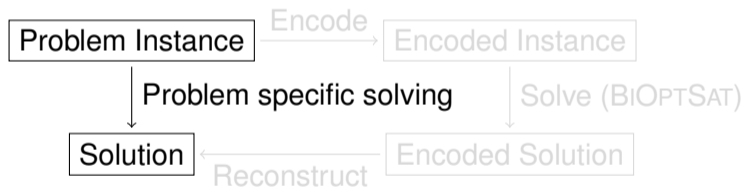
HIGH-LEVEL OVERVIEW OF BIOPTSAT



- Encoding language: Boolean logic
- Lexicographic method
- Based on MaxSAT algorithms
- Making use of so-called SAT solver



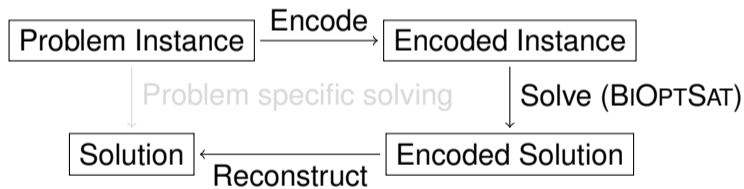
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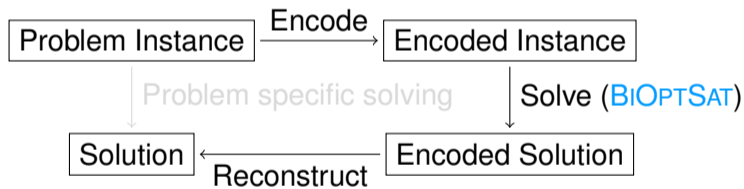
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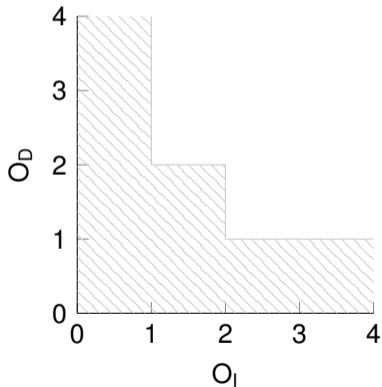
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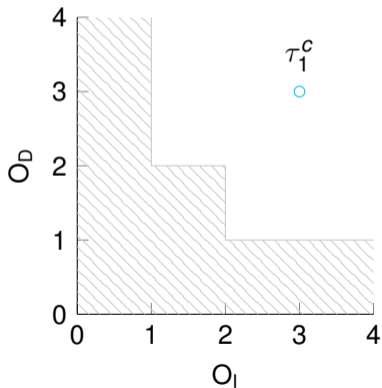
SEARCH PROGRESSION OF BIOPTSAT



- Ordered enumeration from top left to bottom right



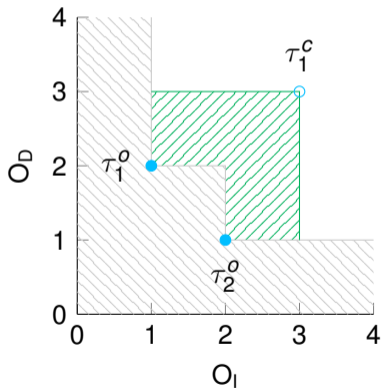
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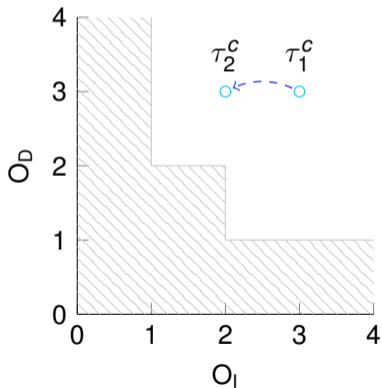
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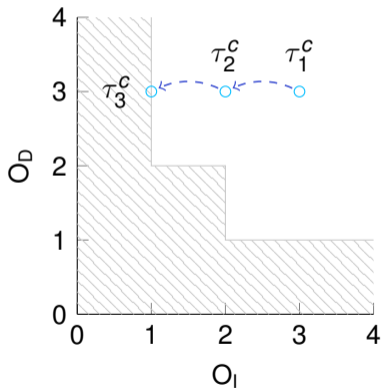
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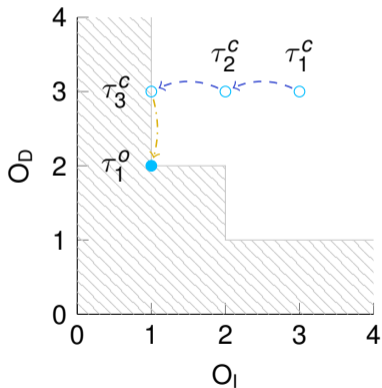
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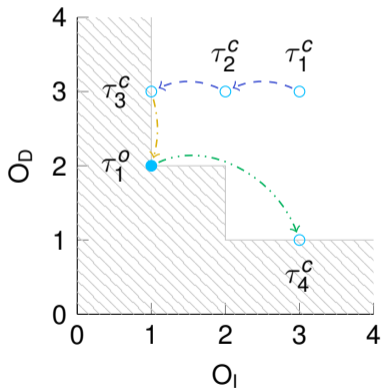
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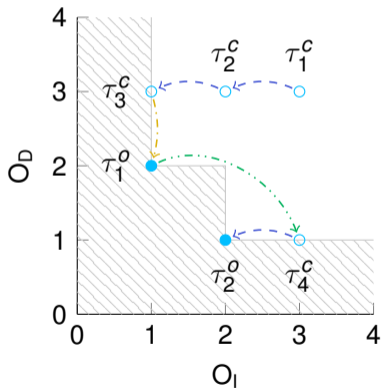
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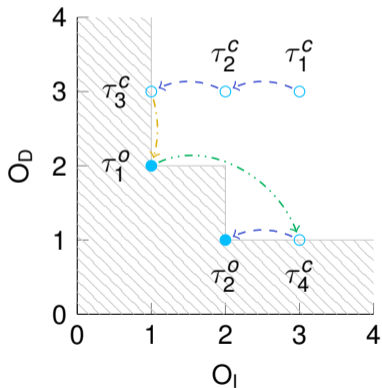
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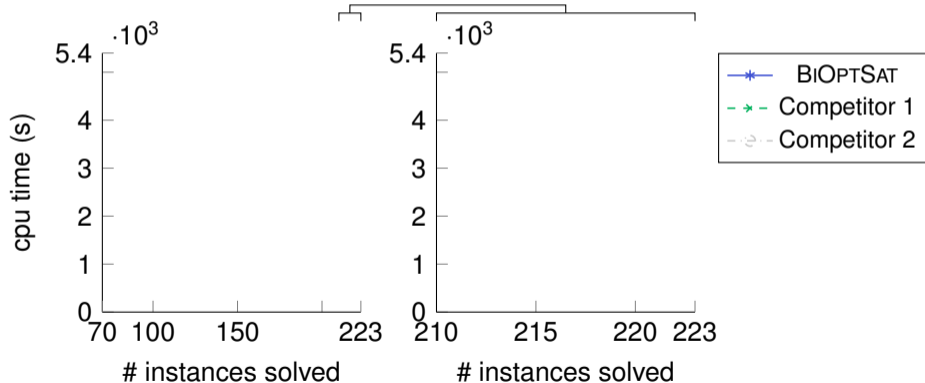


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RESULTS — SOLVED INSTANCES

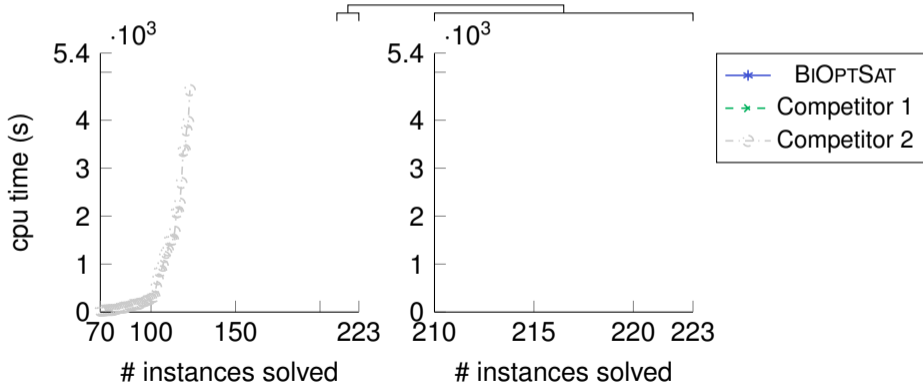
Benchmark: Learning Pareto-optimal decision rules





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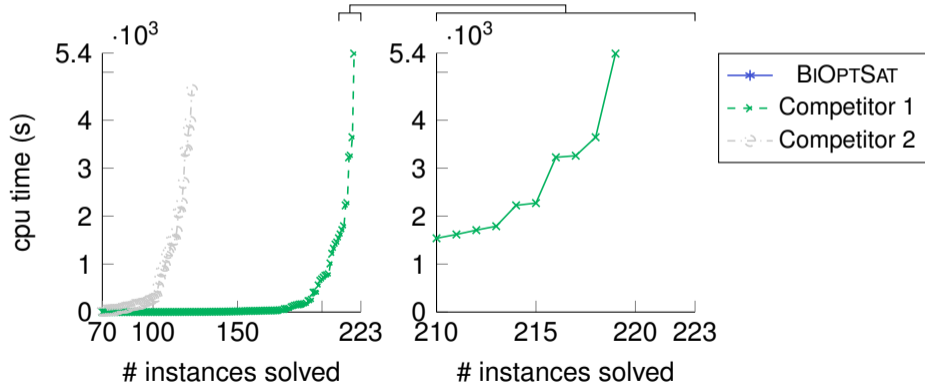
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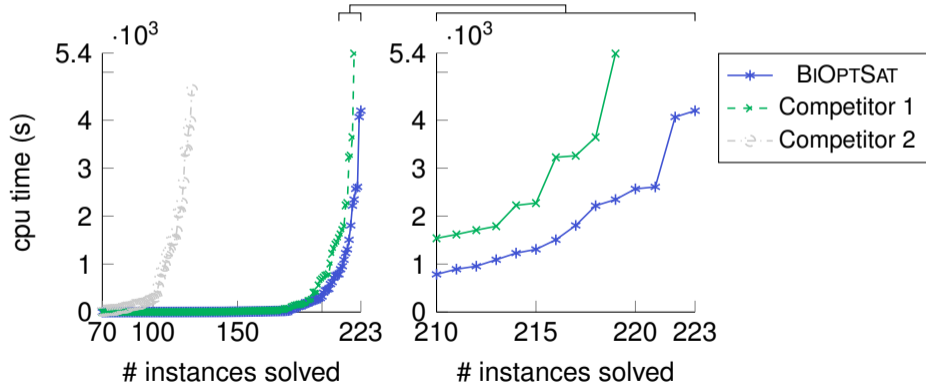
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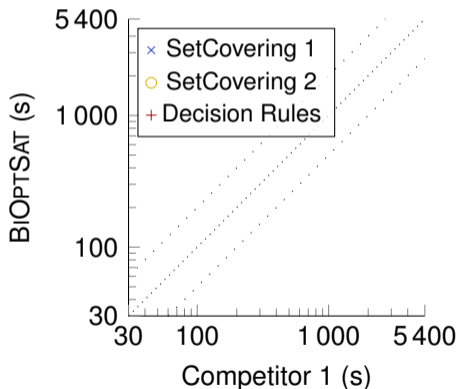
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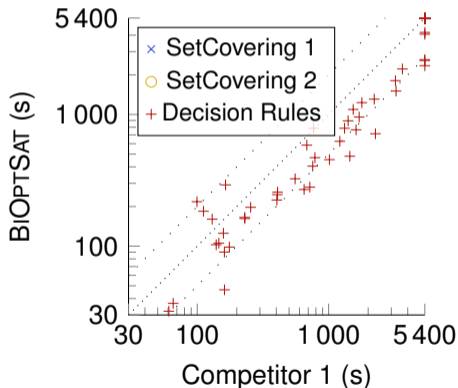
RESULTS — INSTANCE RUNTIMES



- BIOPTSAT outperformed both competitors
- Amount of outperforming competitors depends on benchmark



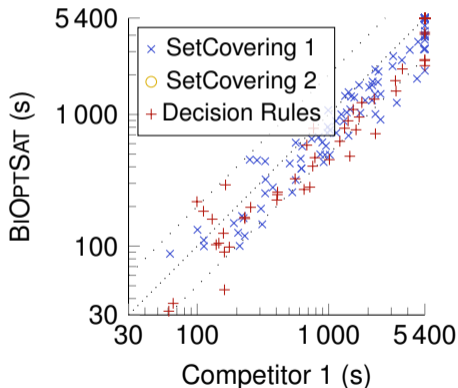
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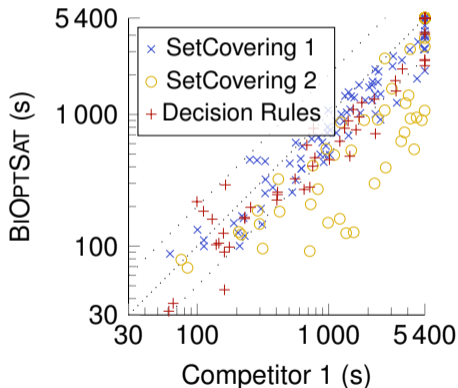
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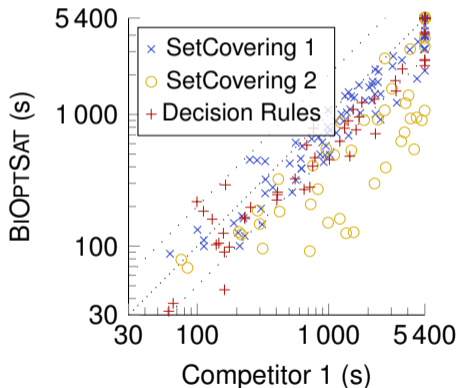
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TAKE AWAY POINTS

Content of Thesis

- Paper in review: significant result for a thesis project
- (Exact) bi-objective optimization is interesting and not that well researched

Thesis Process

- Starting with a summer internship is great
- Working as a research assistant allows for diving deep into a topic



Thank you for your attention!



PARETO OPTIMALITY

Definition (Domination)

Given two objective functions O_1, O_2 and two solutions τ, τ' , τ dominates τ' if (i) $O_i(\tau) \leq O_i(\tau')$ for all $i \in \{1, 2\}$, and (ii) $O_i(\tau) < O_i(\tau')$ for some $i \in \{1, 2\}$. We represent τ dominating τ' as $\tau \prec \tau'$.

Definition (Pareto optimality)

A solution τ is Pareto-optimal iff there is no τ' such that $\tau' \prec \tau$, i.e., τ is not dominated by any other solution.



BIOPTSAT

Algorithm 1 BIOPTSAT: MaxSAT-based bi-objective optimization

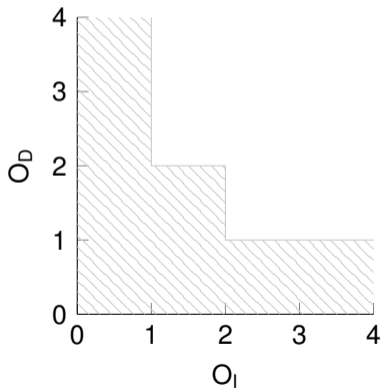
Input: A formula F , two objectives O_I and O_D .

Output: Either one or all Pareto-optimal solution for each Pareto point of F .

- 1: $\tau^r \leftarrow \text{InitSATsolverAndSolve}(F)$ {Invokes the SAT solver on the formula.}
 - 2: $b_D \leftarrow \infty, b_I \leftarrow 0$
 - 3: **while** $\text{res} = \text{SAT}$ **do**
 - 4: $(b_I, \tau^r) \leftarrow \text{Minimize-Inc}(b_D, O_I(\tau^r))$ {Maintains $\text{TOT}(O_I)$ (or similar)}
 - 5: $(b_D, \tau^r) \leftarrow \text{Solution-Improving-Search}(b_I, O_D(\tau^r))$ {Builds $\text{TOT}(O_D)$ }
 - 6: **yield** τ^r {Optionally: **yield** $\text{EnumSols}(b_D, b_I)$ }
 - 7: $(\text{res}, \tau^r) \leftarrow \text{isSAT}(\{\langle O_D < b_D \rangle\})$
-



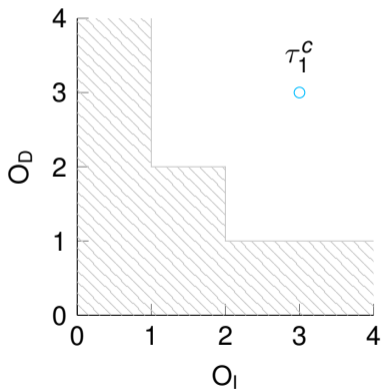
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- Progression from top left to bottom right



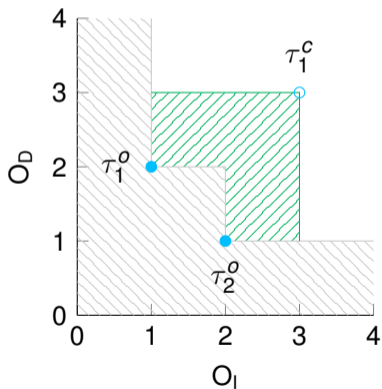
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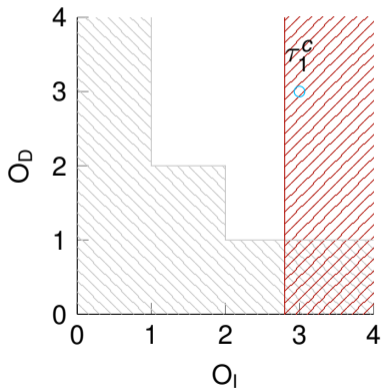
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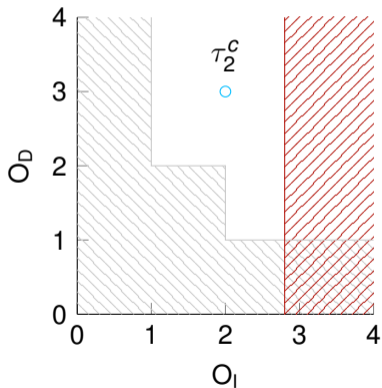
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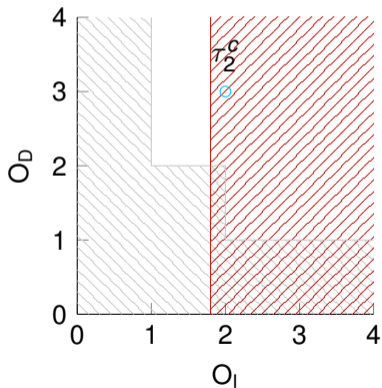
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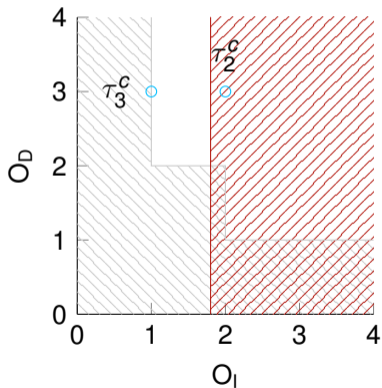
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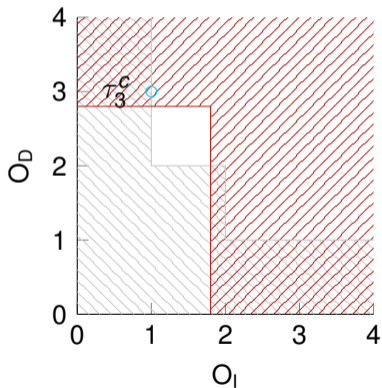
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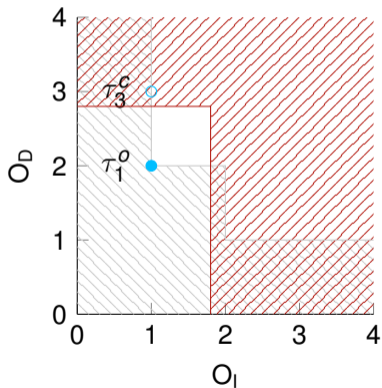
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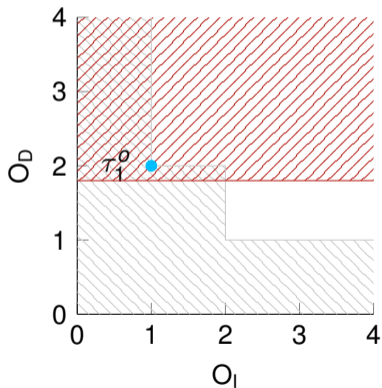
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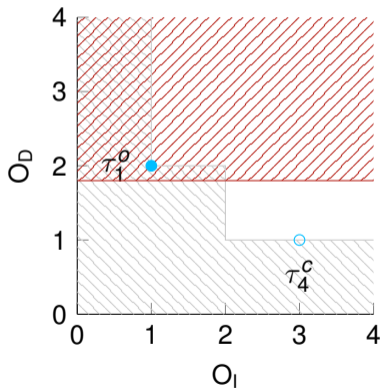
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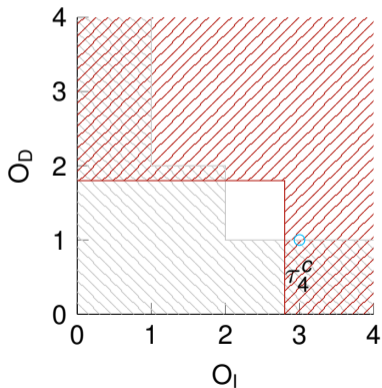
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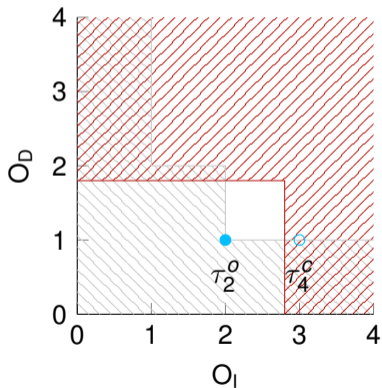
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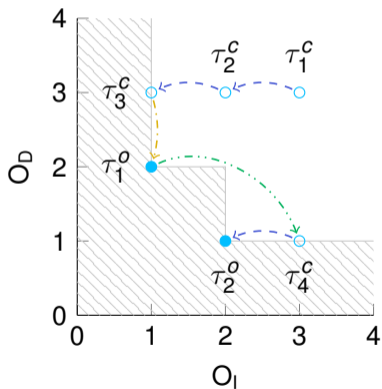
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