

Last-minute Crew Rescheduling: Model and Heuristic Approach

Liyun Yu, Carl Henrik Häll, Anders Peterson, Christiane Schmidt



- Introduction
- Problem Description
- Methodology
- Case Study

Introduction



Infrastructure failure

Bad weather

Rolling stock breakdown

Blocks on rail track

Infeasible vehicle schedule

Employee shortage



Timetable

Infrastructure failure

Bad weather

Rolling stock breakdown

Blocks on rail track

Timetable

Rolling Stock

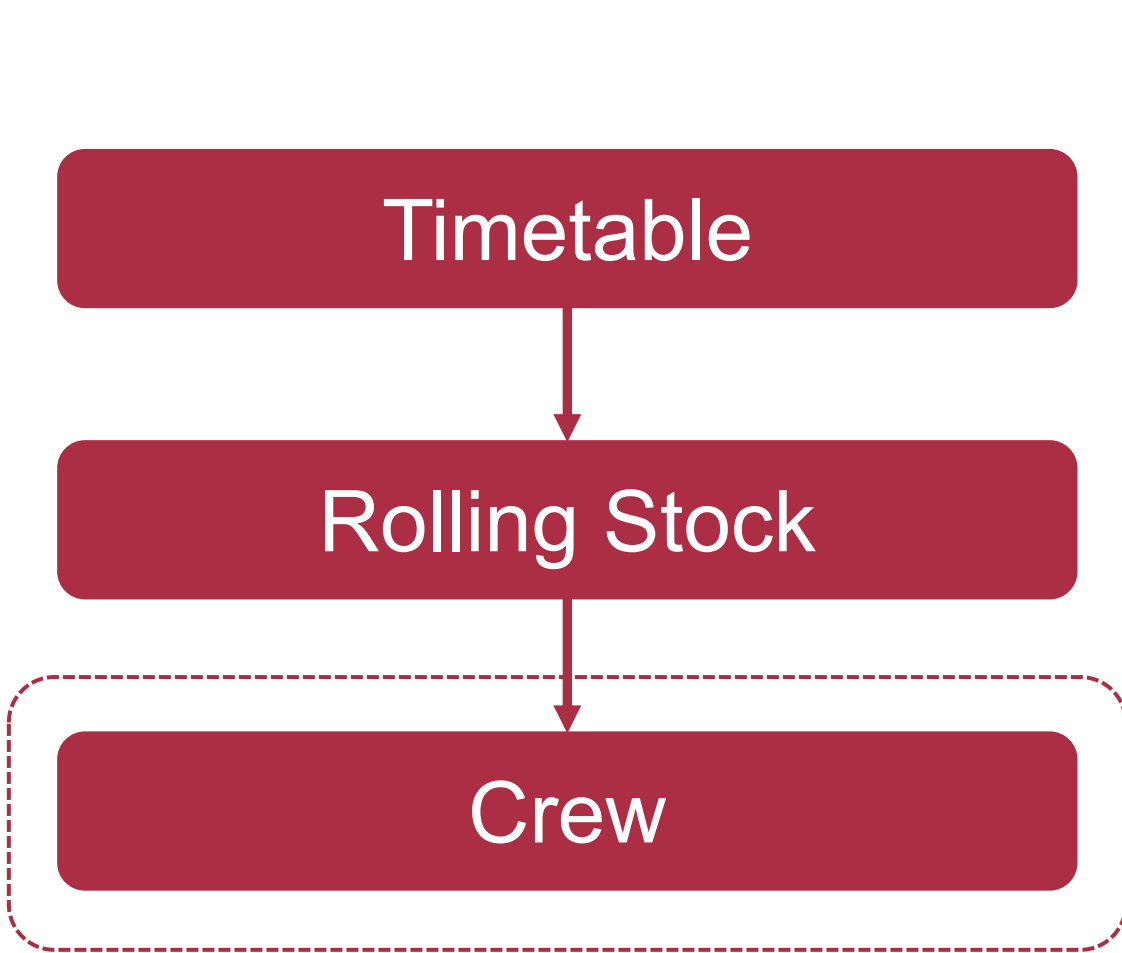
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Blocks on rail track

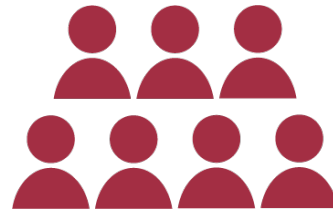
Infeasible vehicle schedule

Employee shortage

Employee shortage

Driver
~~**Employee shortage**~~

Driver shortage



Society

- Prefer as planned
- Less changes on short notice
- Fast-generated schedule after disruption



Rail Undertaking

Cancel fewer trains



High cancellation cost

Short term

- Financial loss
- Employees' working time loss



Rail Undertaking

Cancel fewer trains



High cancellation cost

Long term

- Less trust from passengers
- Poor competitiveness

Problem Description

Problem Description



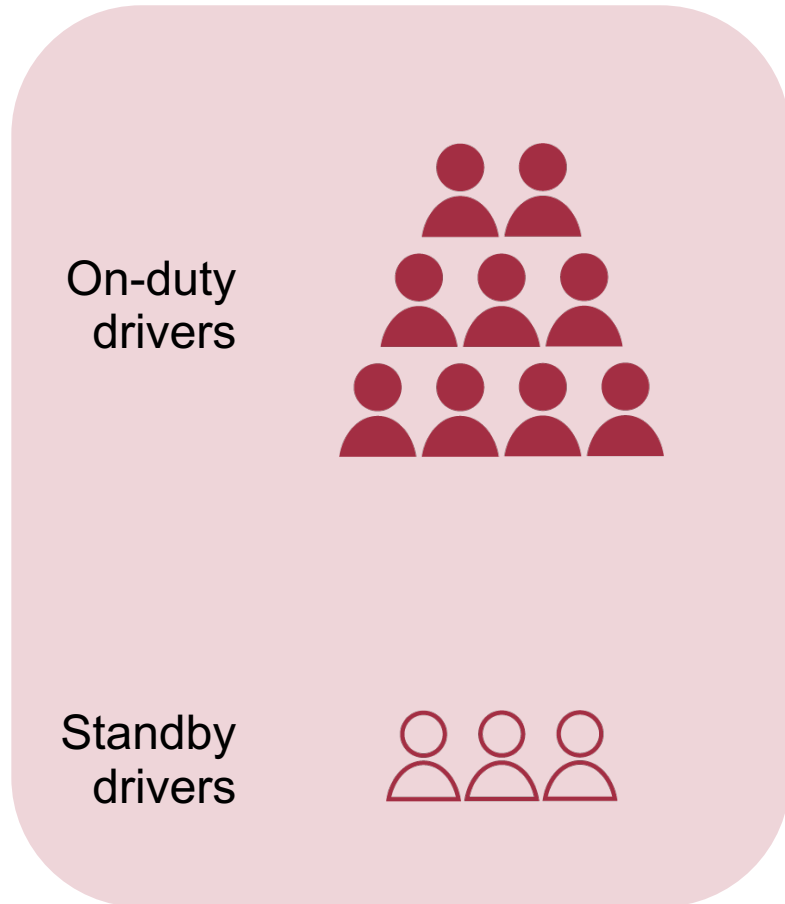
Our scenario

Daily driver shortage

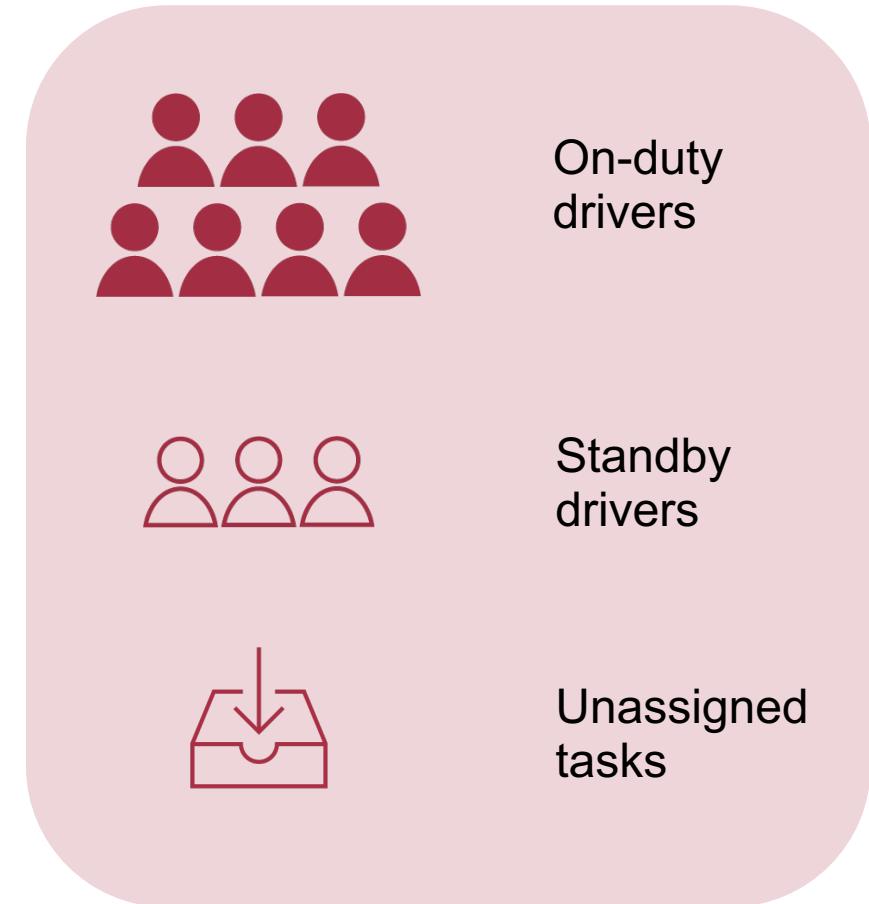
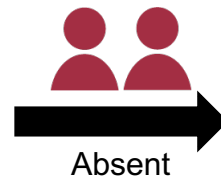
- Take leaves in short notice
- Insufficient standby drivers

Goal

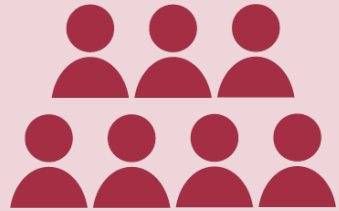
- # Task cancellations
- # Changed tasks



Original



Now



On-duty
drivers

Given (feasible) schedules



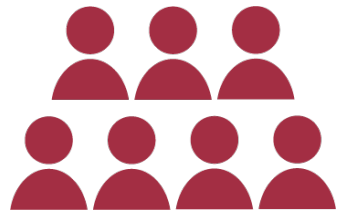
Standby
drivers

Empty schedules with fixed working time and depot



Unassigned
tasks

Set of unassigned tasks



On-duty
drivers

Given (feasible) schedules



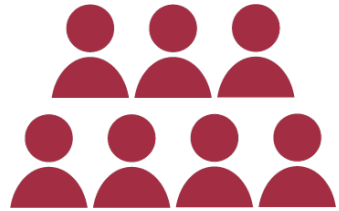
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Methodology

MILP problem

Objective function

$$\text{minimize } f(x_{g,d}, z_{g,d})$$

MILP problem

Objective function

$$\text{minimize } f(x_{g,d}, z_{g,d})$$

Total number of
unassigned tasks

MILP problem

Objective function

minimize $f(x_{g,d}, z_{g,d})$

The diagram illustrates the objective function $f(x_{g,d}, z_{g,d})$. The variable $x_{g,d}$ is enclosed in a solid red box, and the variable $z_{g,d}$ is enclosed in a dashed red box. A solid red line connects the $x_{g,d}$ box to a callout box below it, and a dashed red line connects the $z_{g,d}$ box to another callout box below it.

Total number of unassigned tasks

Total number of changed tasks

MILP problem

Constraints

- Consistent connections of time and geographical location
- Total working time
- Driver's license
- Rest
- Break

MILP problem

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MILP problem

Constraints

- Consistent connections of time and geographical location
- Total working time
- Driver's license
- Rest
- Break 
 - Break time duration
 - Maximum work hour without a break

MILP Model with Commercial Solver

- to get the optimal solution

Approach Based on Tabu Search

- less computational time and space
- good enough result

Tabu Search

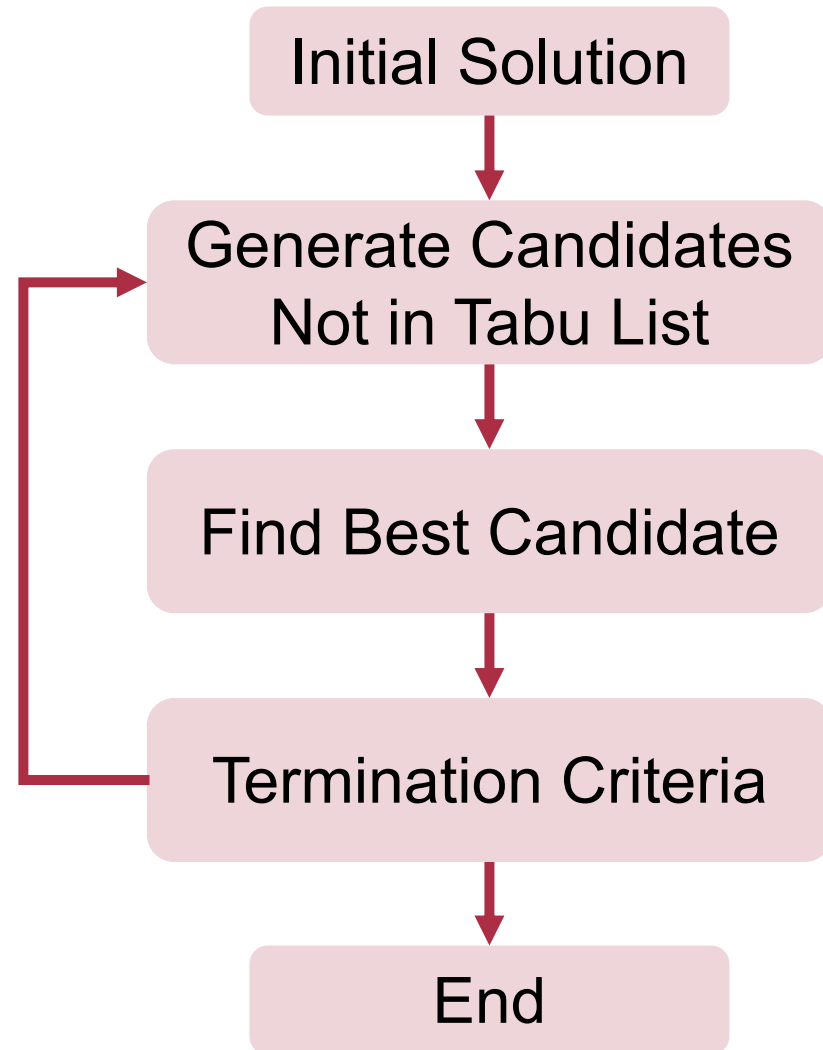
A local search-based heuristic that avoids revisiting solutions by recording the recent history of the search in a short-time memory called Tabu List. [1]

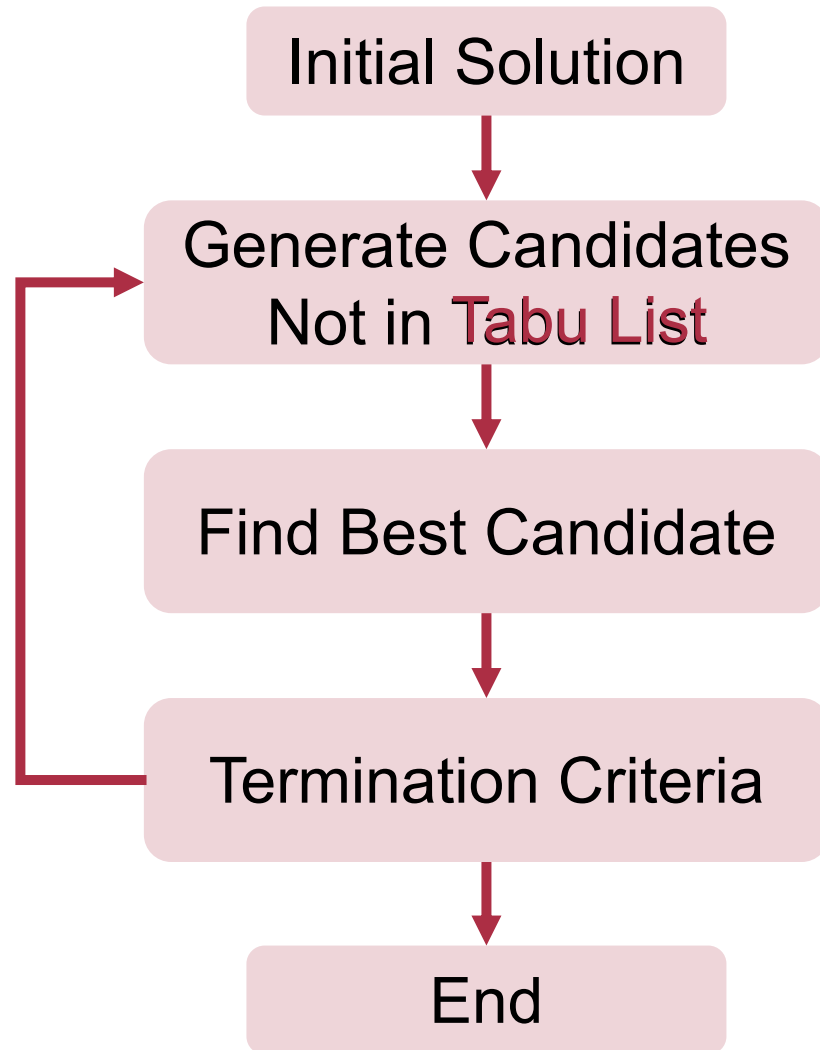
Tabu Search

A local search-based heuristic that avoids revisiting solutions by recording the recent history of the search in a short-time memory called Tabu List. [1]

Tabu Search

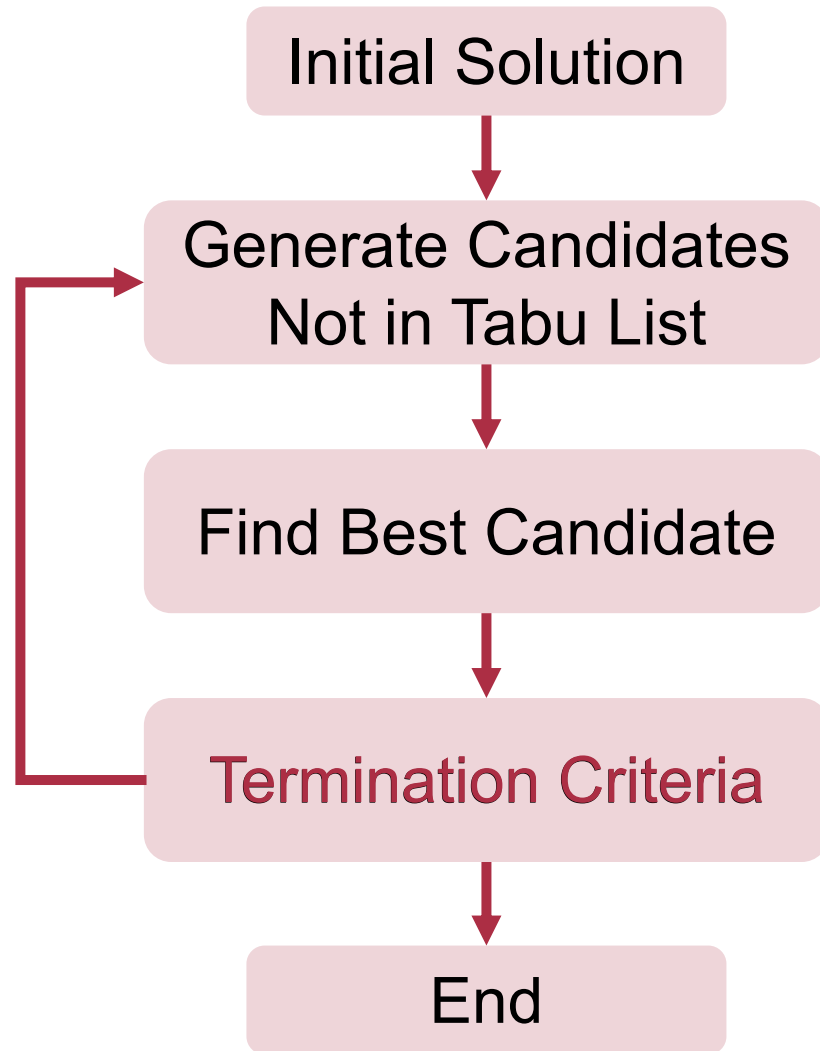
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Tabu List

- Short-time memory
- Avoiding local optimum



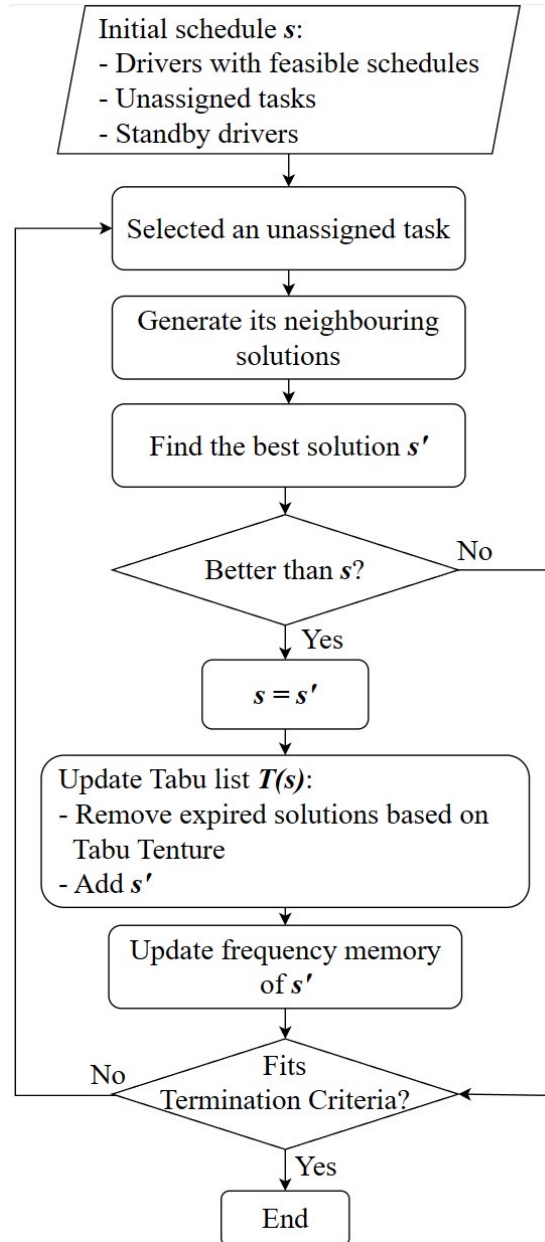
Tabu List

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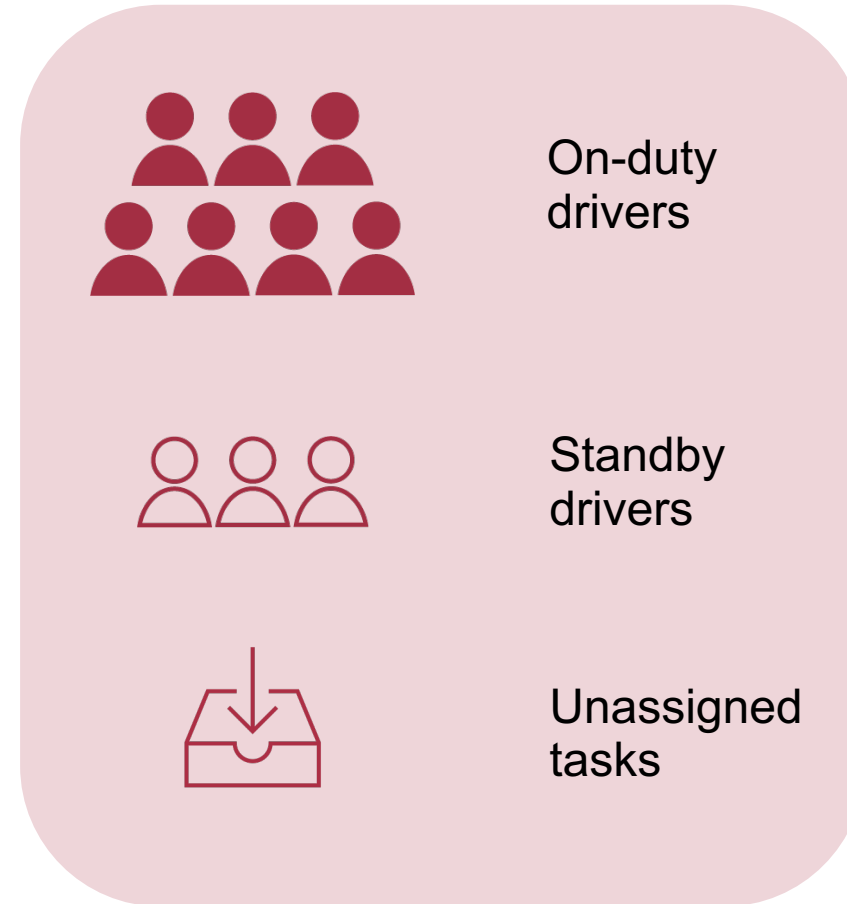
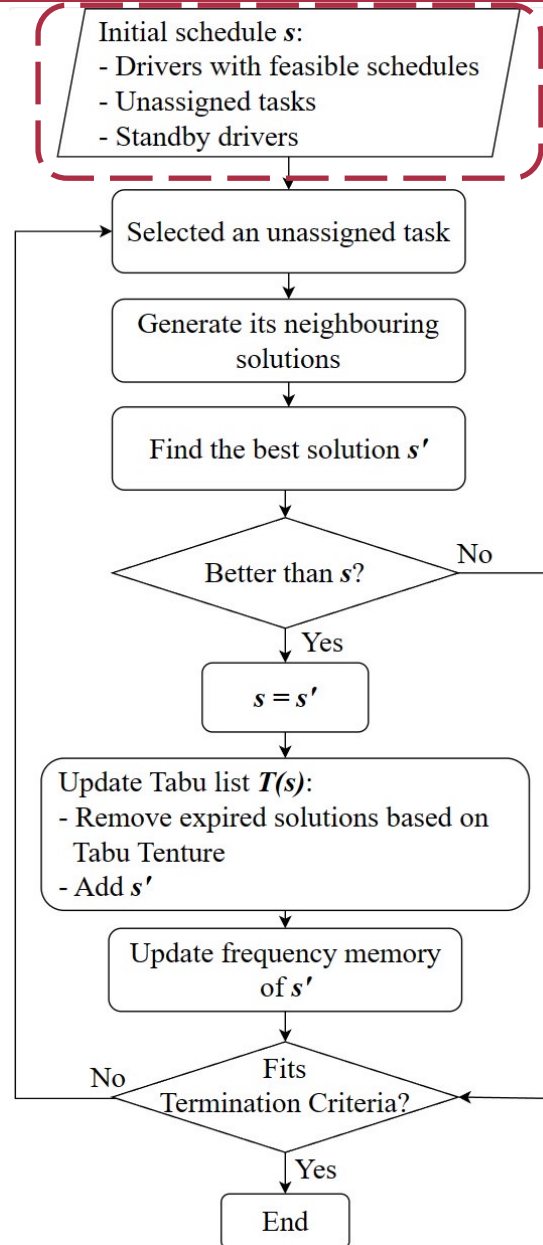
Termination Criteria

- Terminating the approach

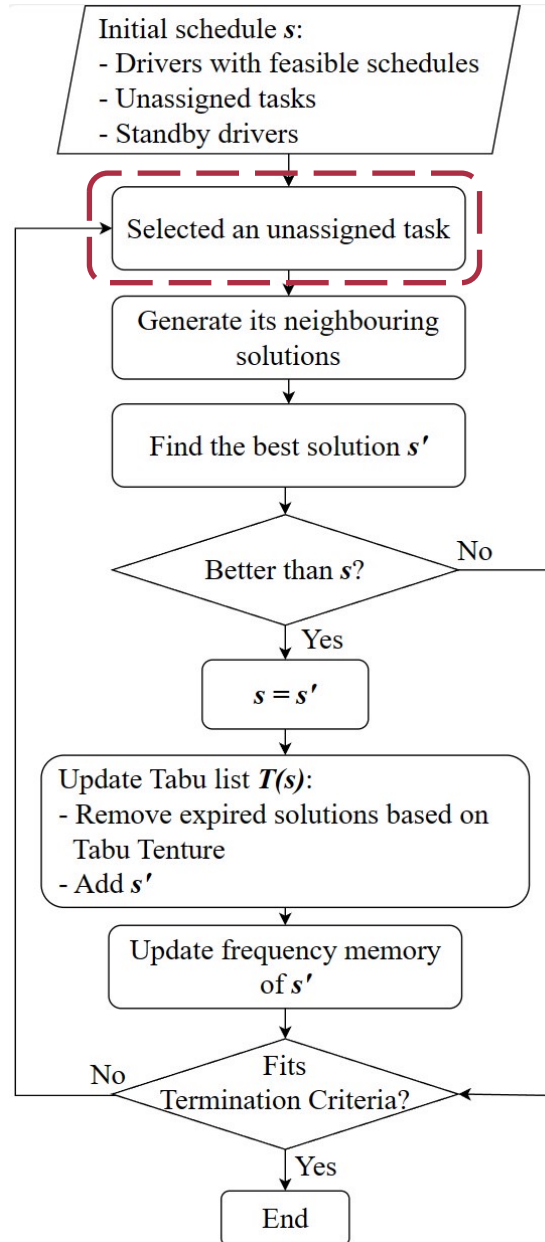
Approach: Tabu Search



Approach: Tabu Search



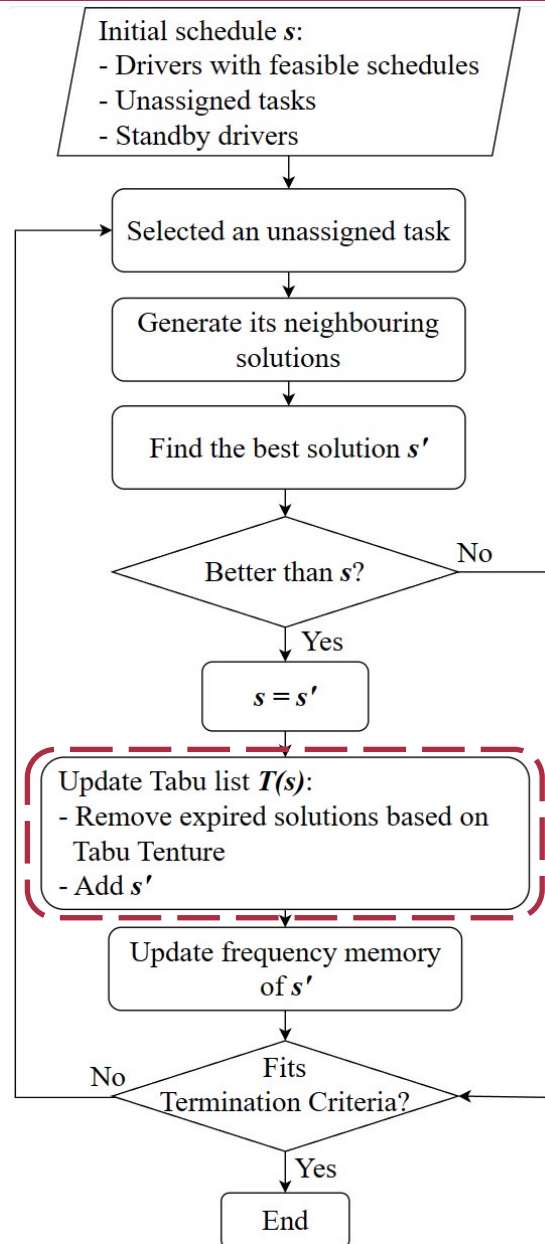
Approach: Tabu Search



Select an unassigned task:

- Randomly

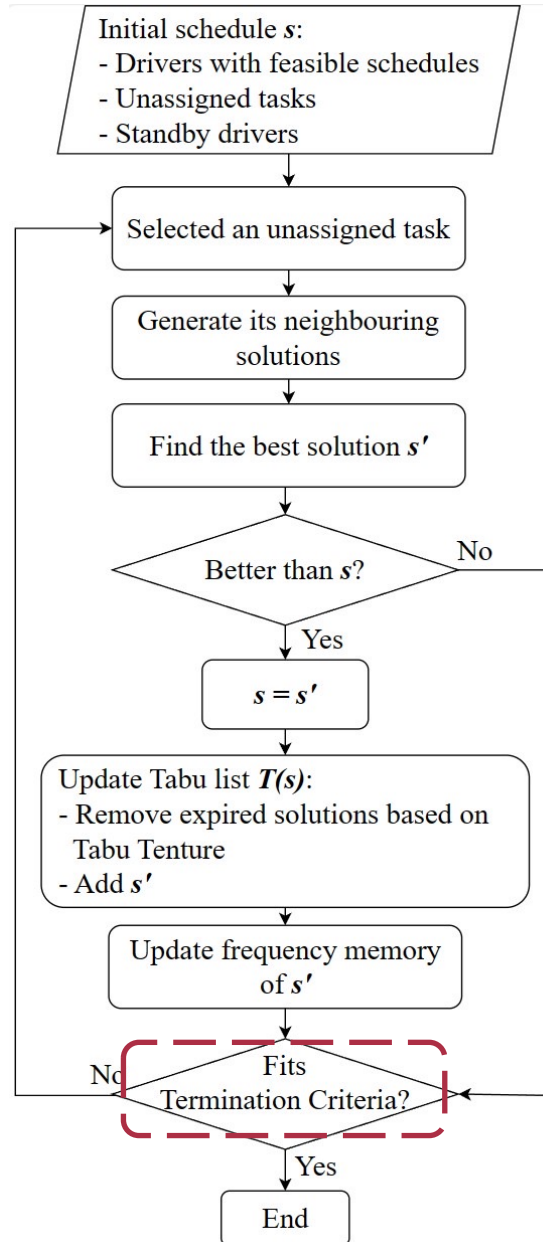
Approach: Tabu Search



Tabu List:

- The schedule of all drivers

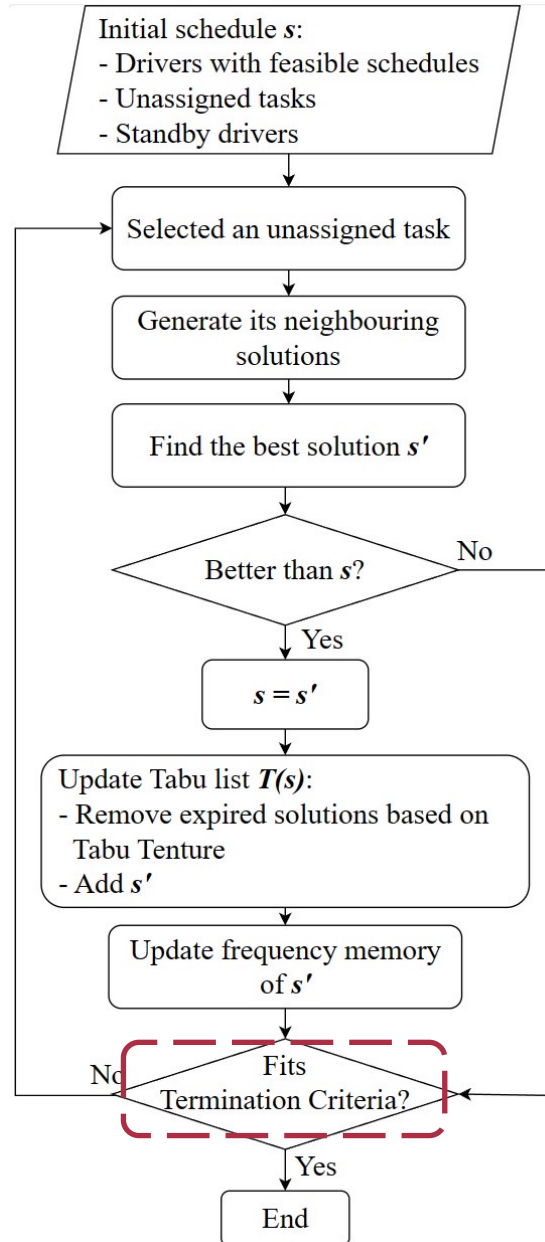
Approach: Tabu Search



Termination Criteria:

- Maximum number of iteration
- The rest of all unassigned tasks cannot be assigned
- The unassigned task pool is empty

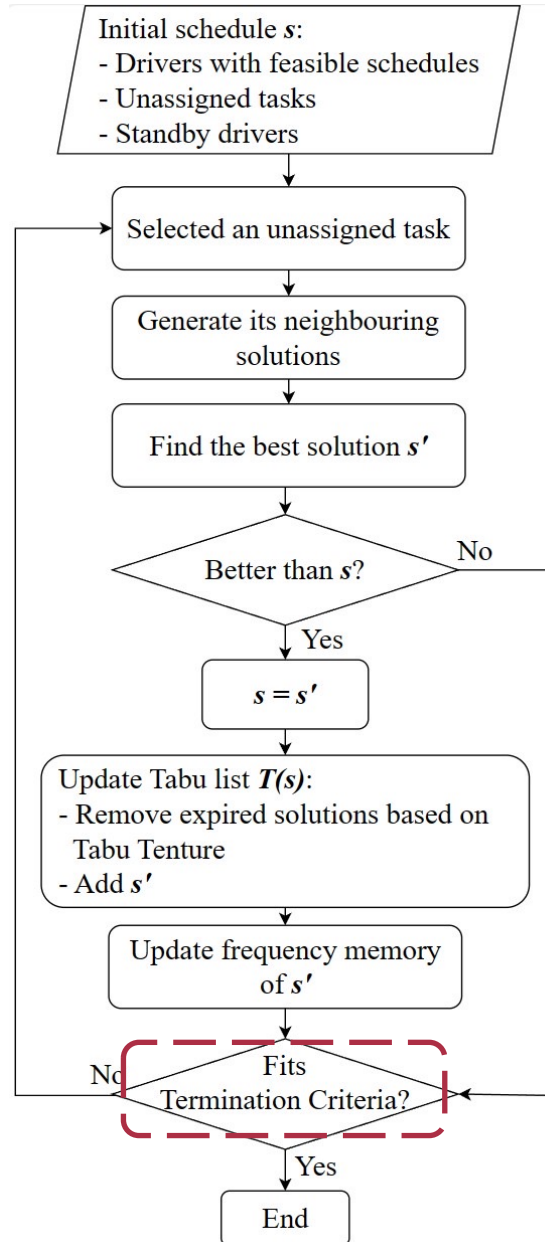
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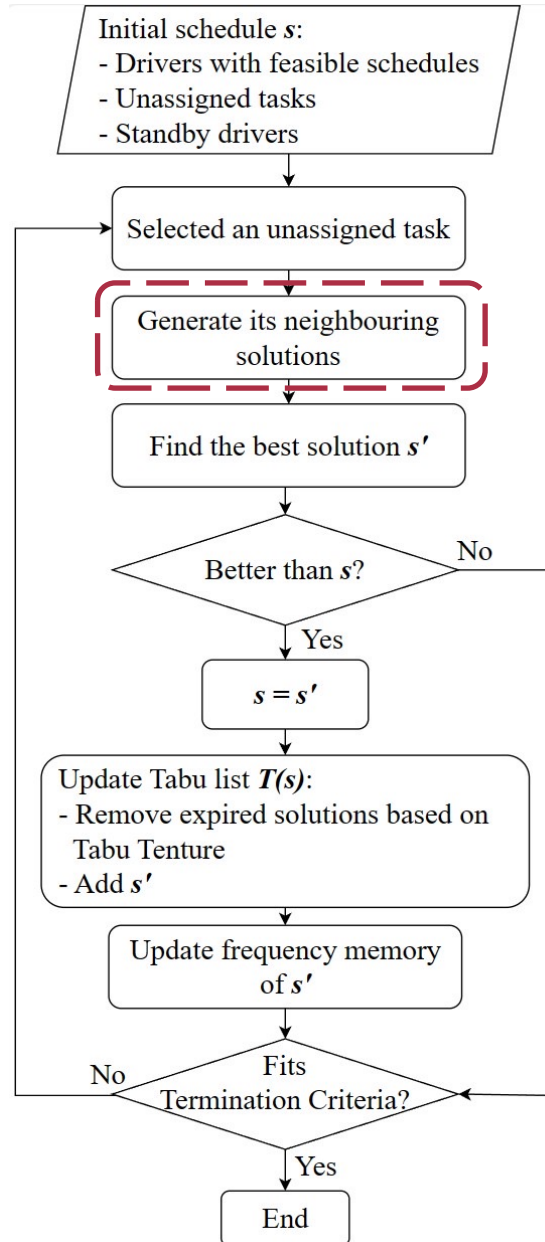
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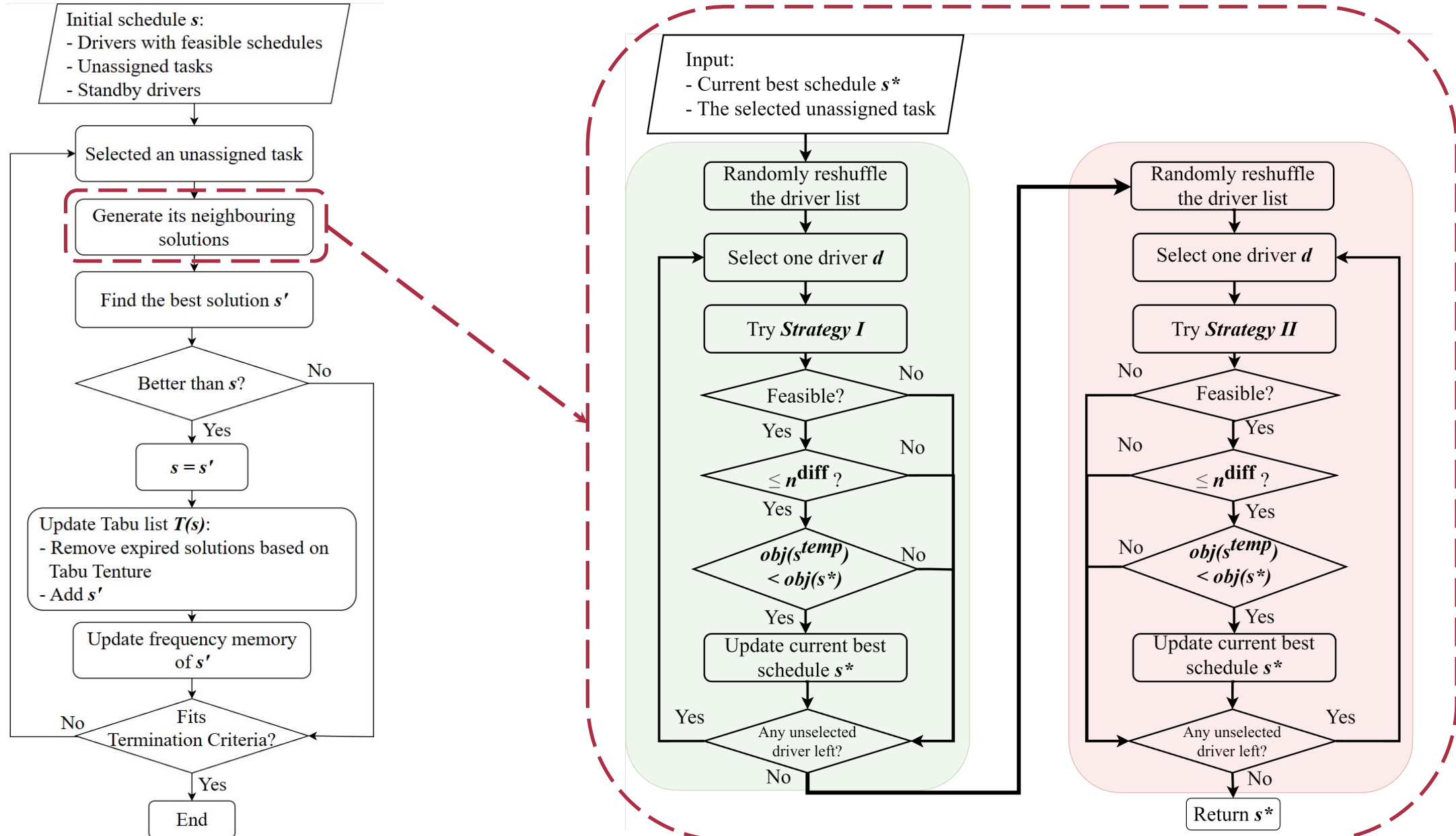
Approach: Tabu Search



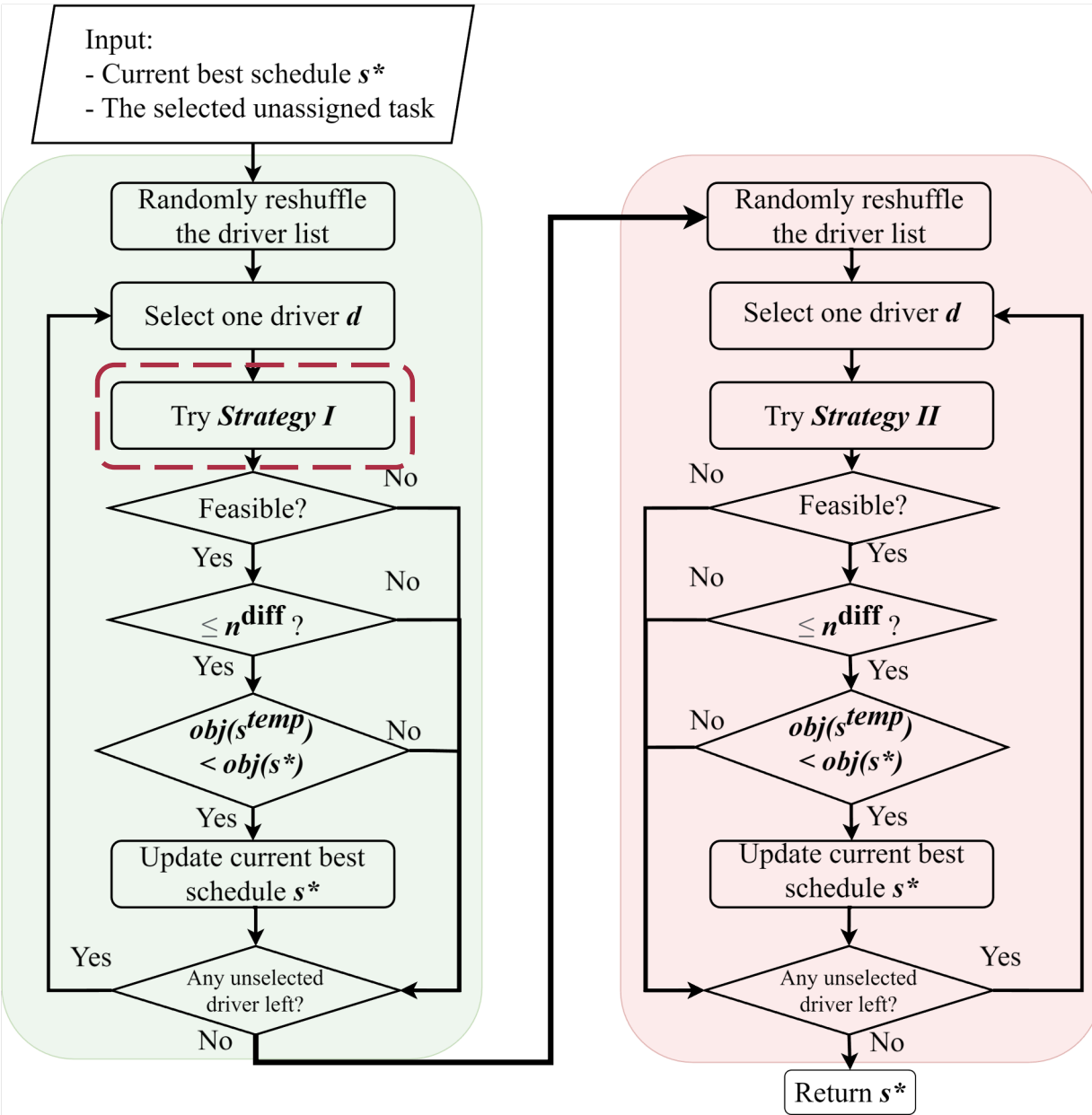
Neighboring Solutions:

- Deadheading
- Extra assign

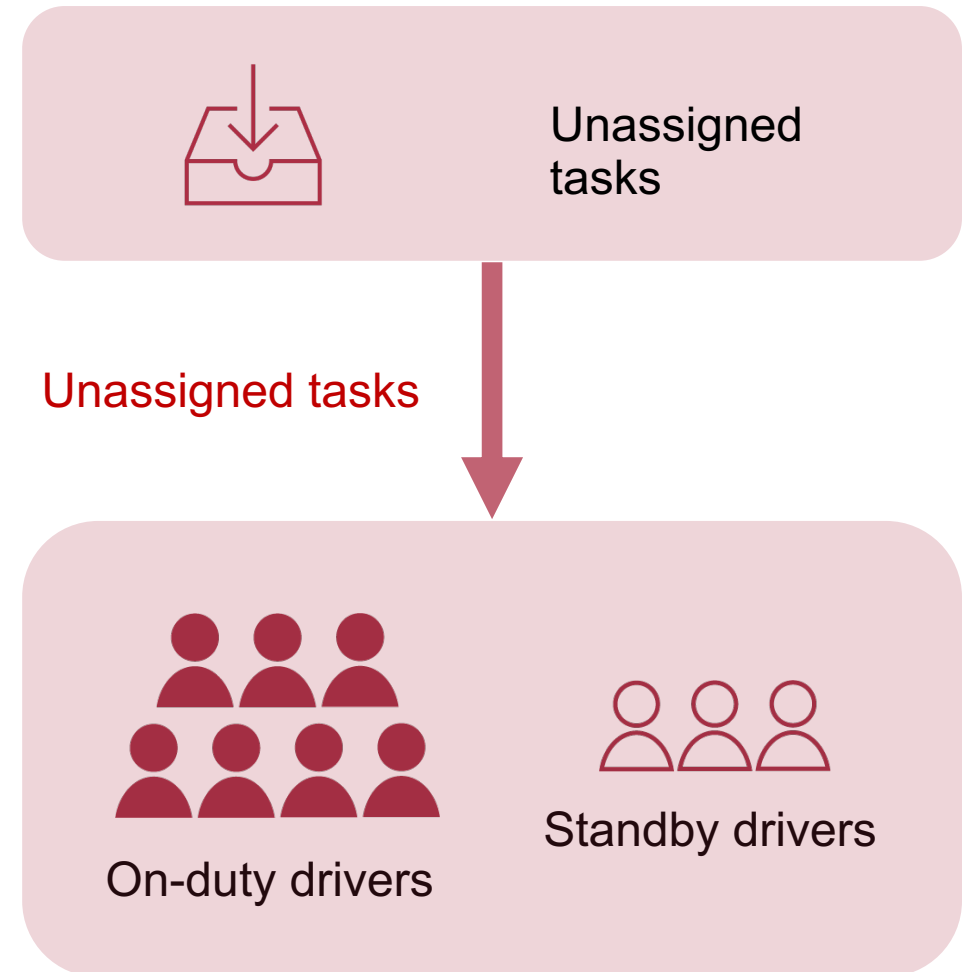
Approach: Tabu Search



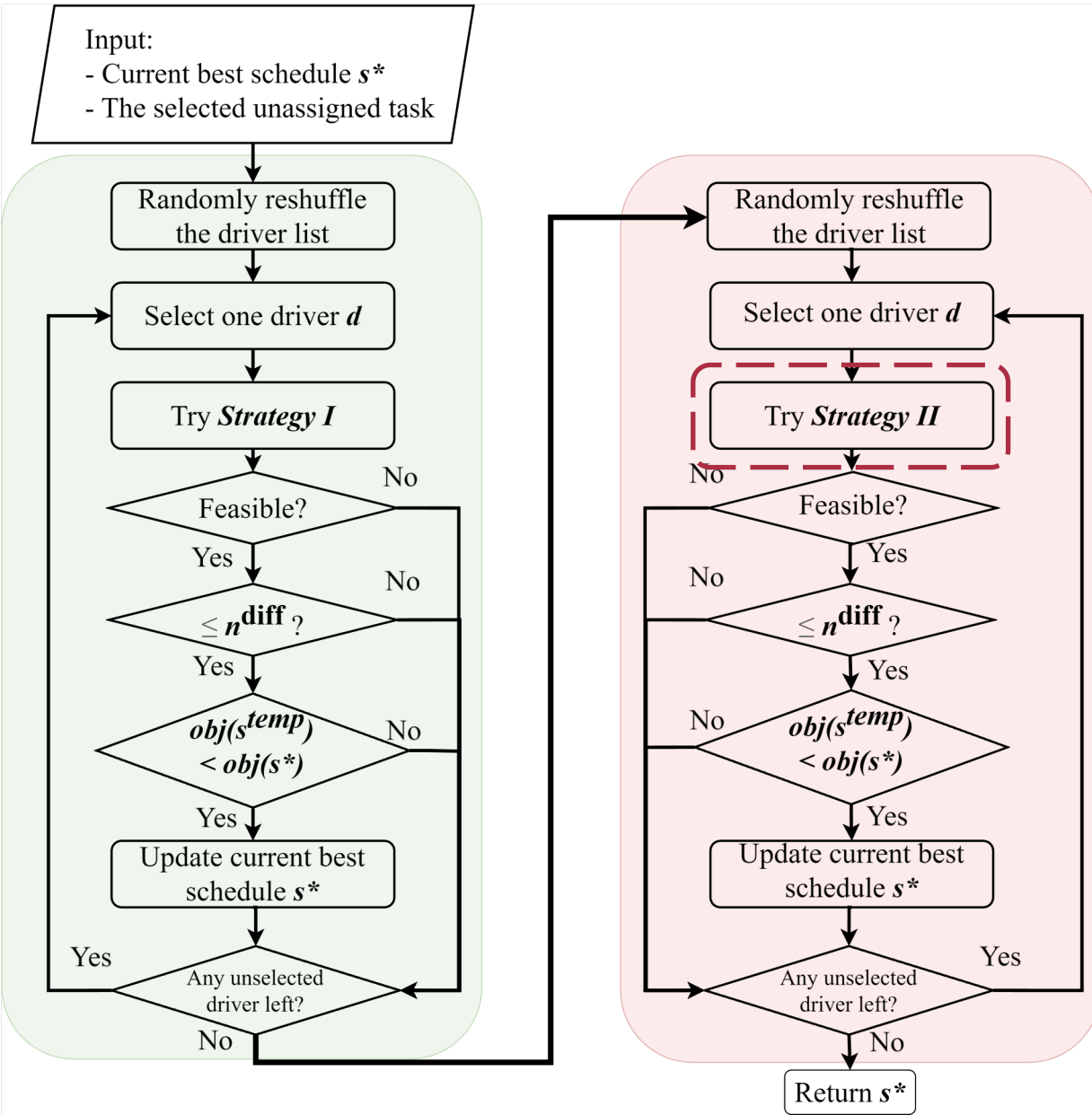
Approach: Neighboring Solutions



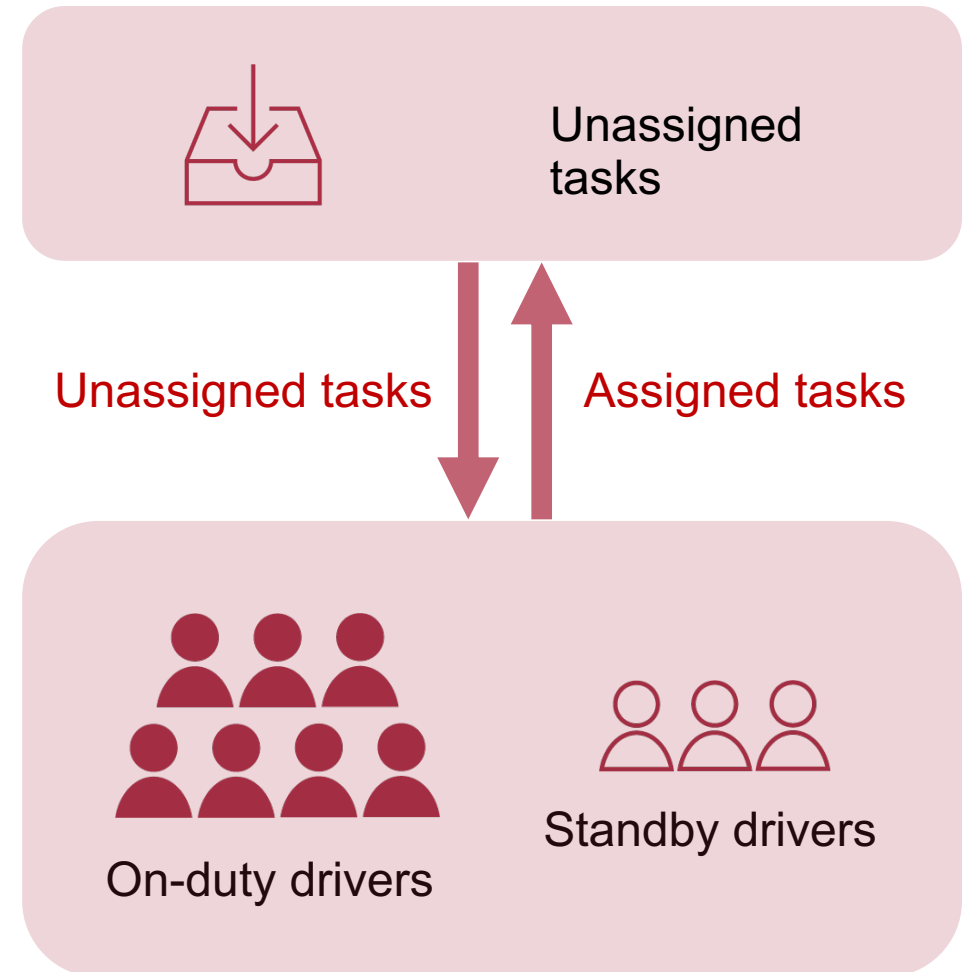
Strategy I: Directly assign to all drivers



Approach: Neighboring Solutions



Strategy II: Swap with assigned tasks



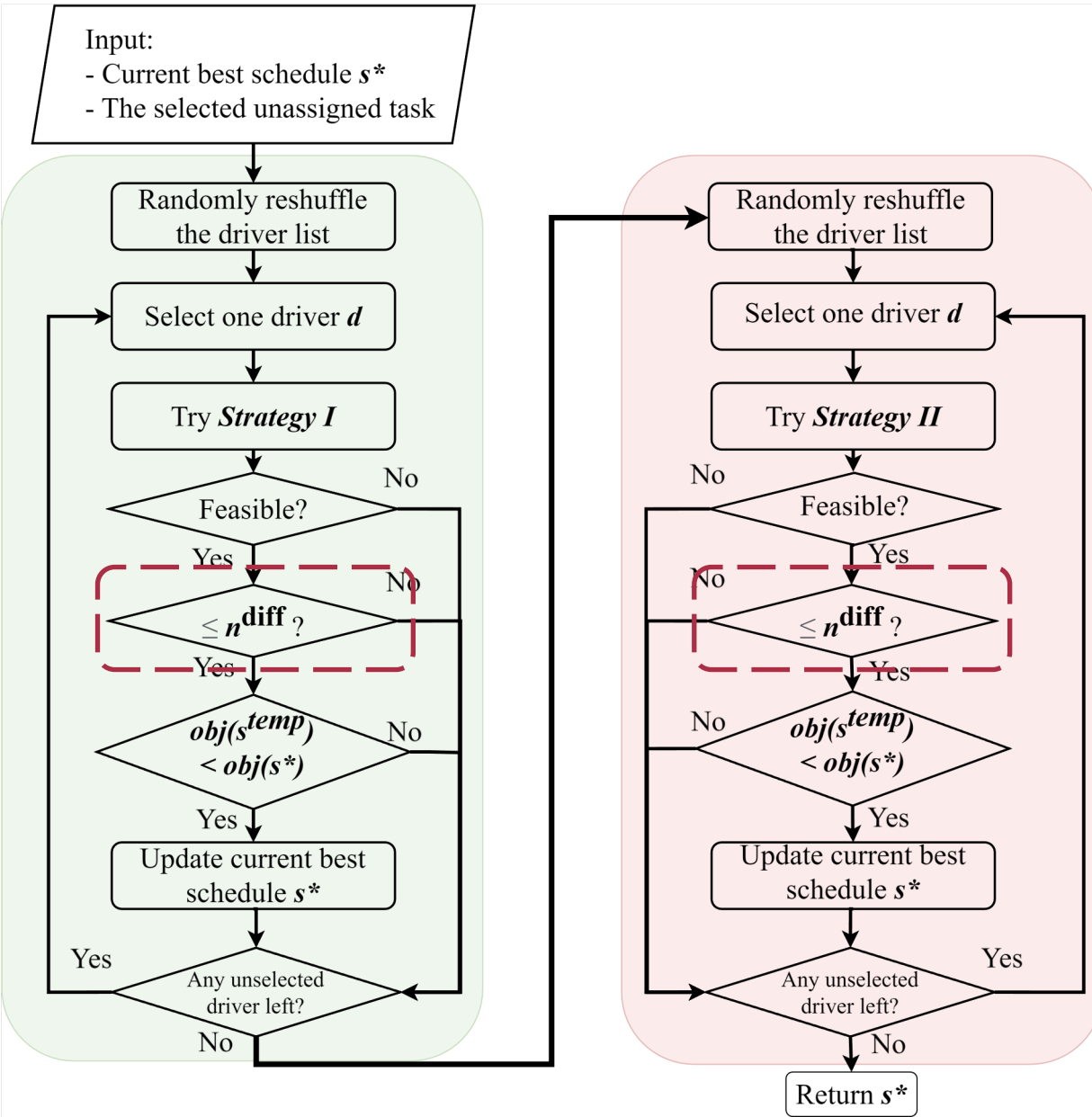
Approach: Neighboring Solutions



Feasibility Check

Equivalent with the constraints in MILP

Approach: Neighboring Solutions



Level of freedom

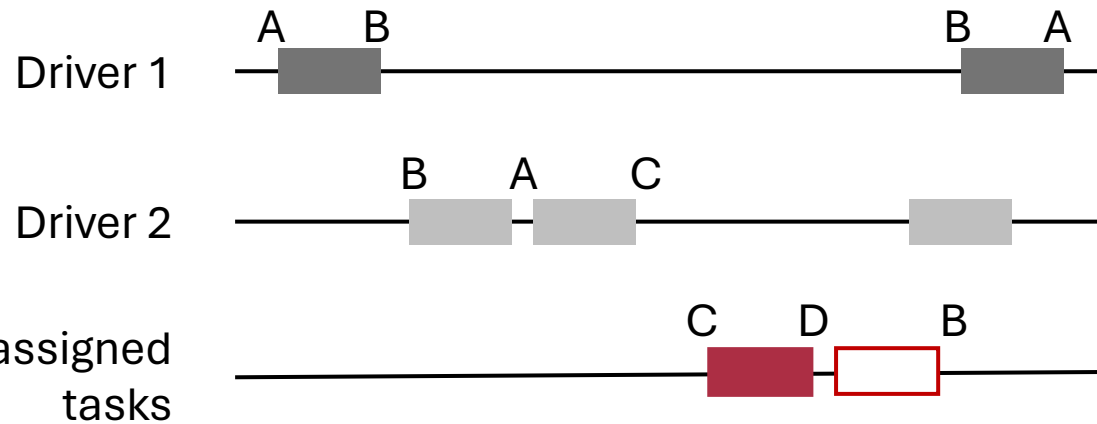
n^{diff} : maximum allowed difference between # tasks unassigned from driver d and # tasks assigned to driver d .

Approach: Neighboring Solutions

Strategy I: Directly assign to all drivers

■ Selected unassigned task

□ Another unassigned task

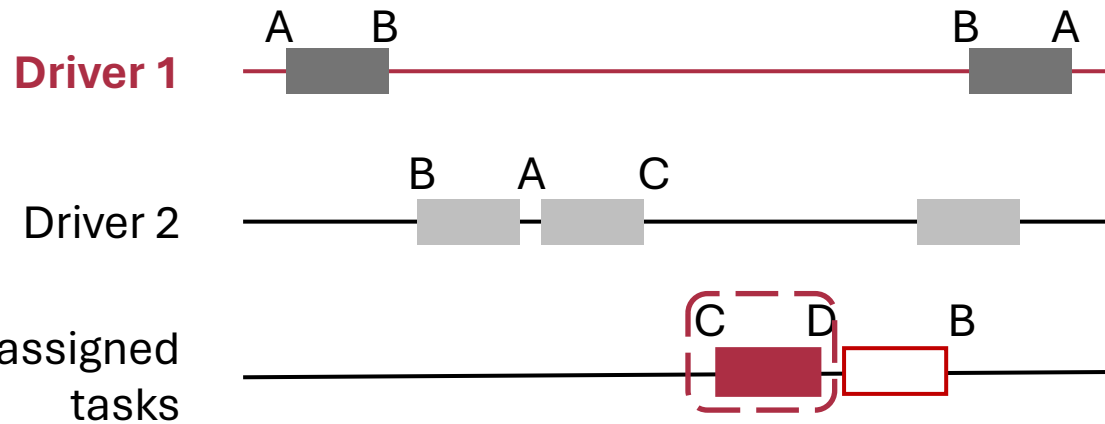


Approach: Neighboring Solutions

Strategy I: Directly assign to all drivers

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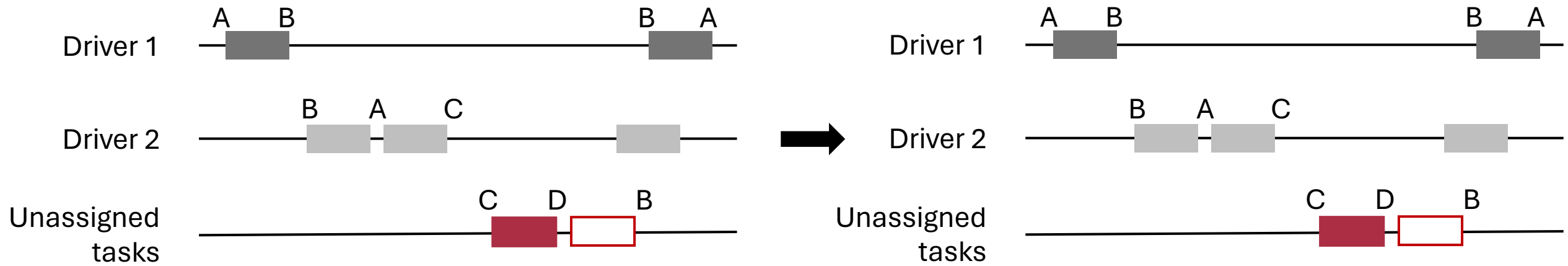
□ Another unassigned task



Approach: Neighboring Solutions

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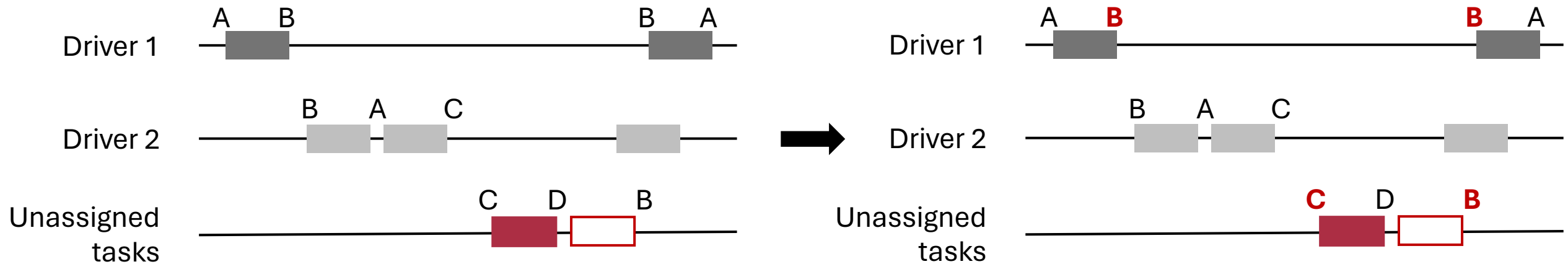
- Selected unassigned task
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Approach: Neighboring Solutions

Strategy I: Directly assign to all drivers

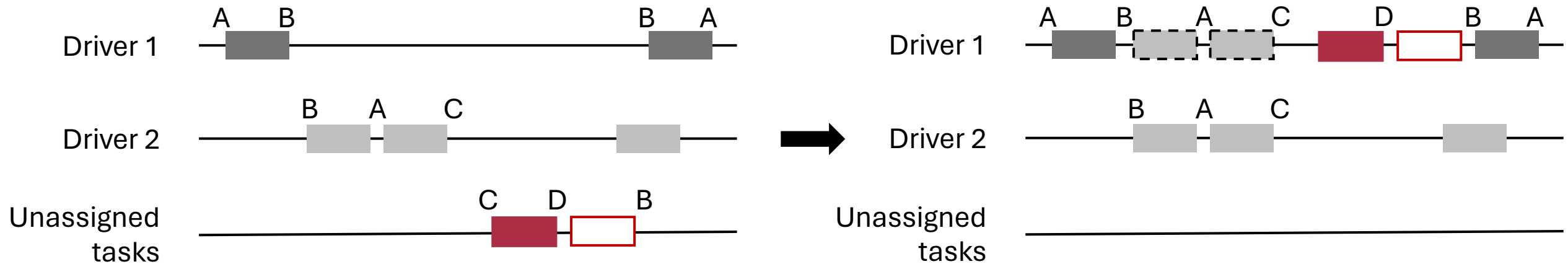
- Selected unassigned task
- Another unassigned task



Approach: Neighboring Solutions

Strategy I: Directly assign to all drivers

- Selected unassigned task
- Another unassigned task
- Deadheading assigned task

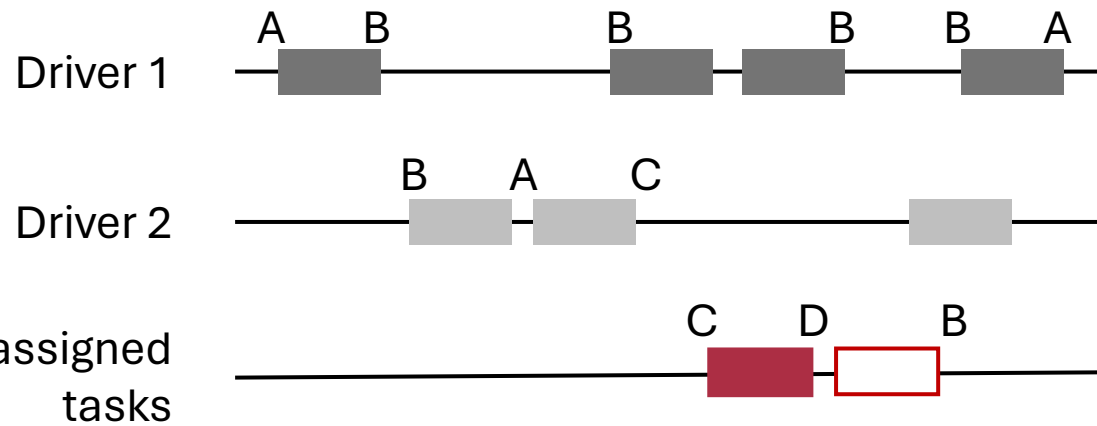


Approach: Neighboring Solutions

Strategy II: Swap with assigned tasks

■ Selected unassigned task

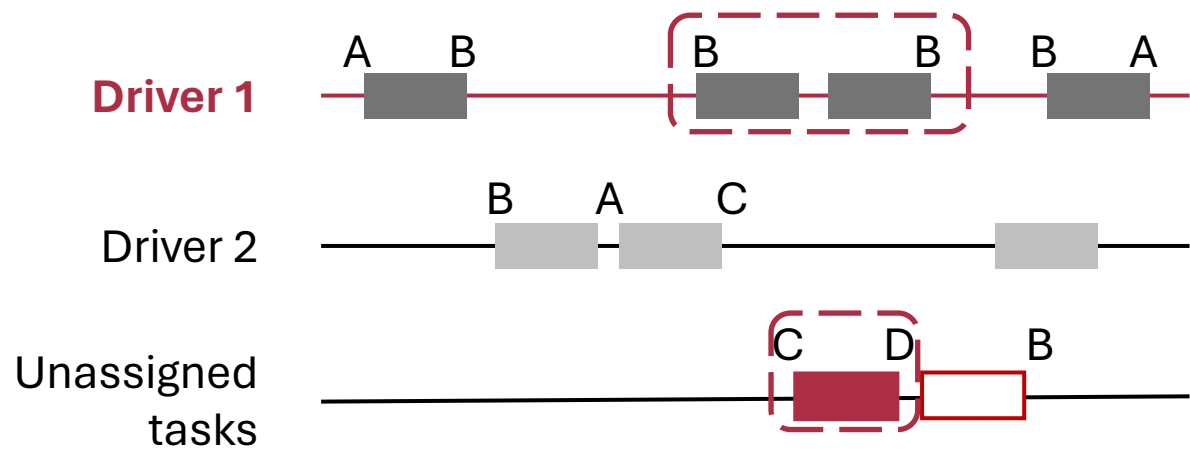
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Approach: Neighboring Solutions

Strategy II: Swap with assigned tasks

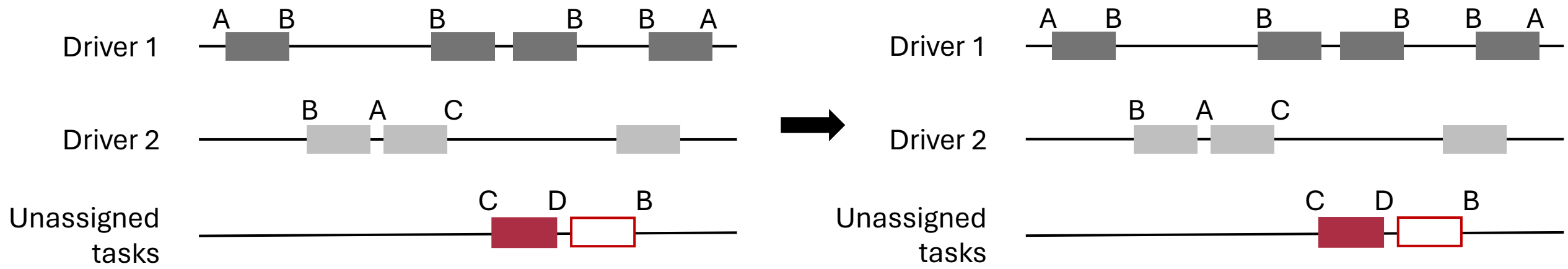
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Approach: Neighboring Solutions

Strategy II: Swap with assigned tasks

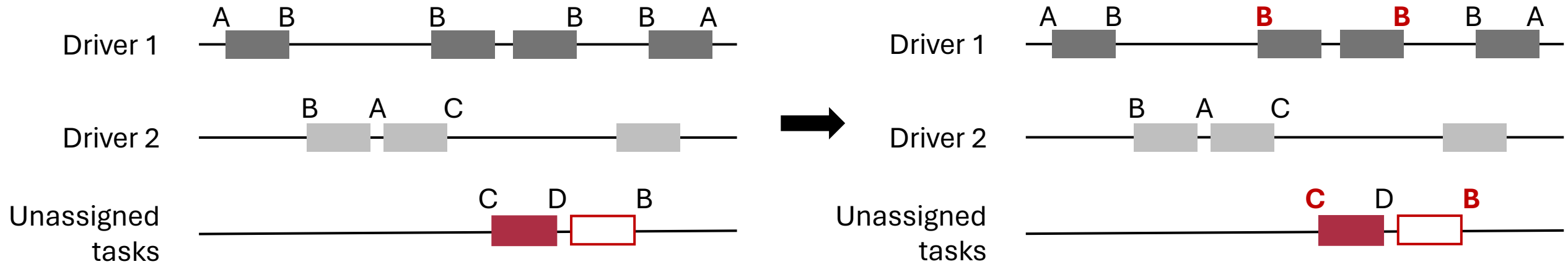
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Approach: Neighboring Solutions

Strategy II: Swap with assigned tasks

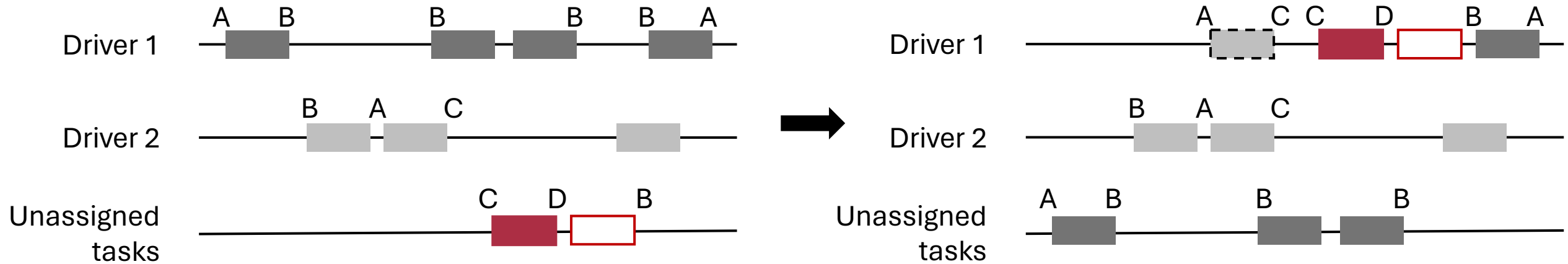
- Selected unassigned task
- Another unassigned task



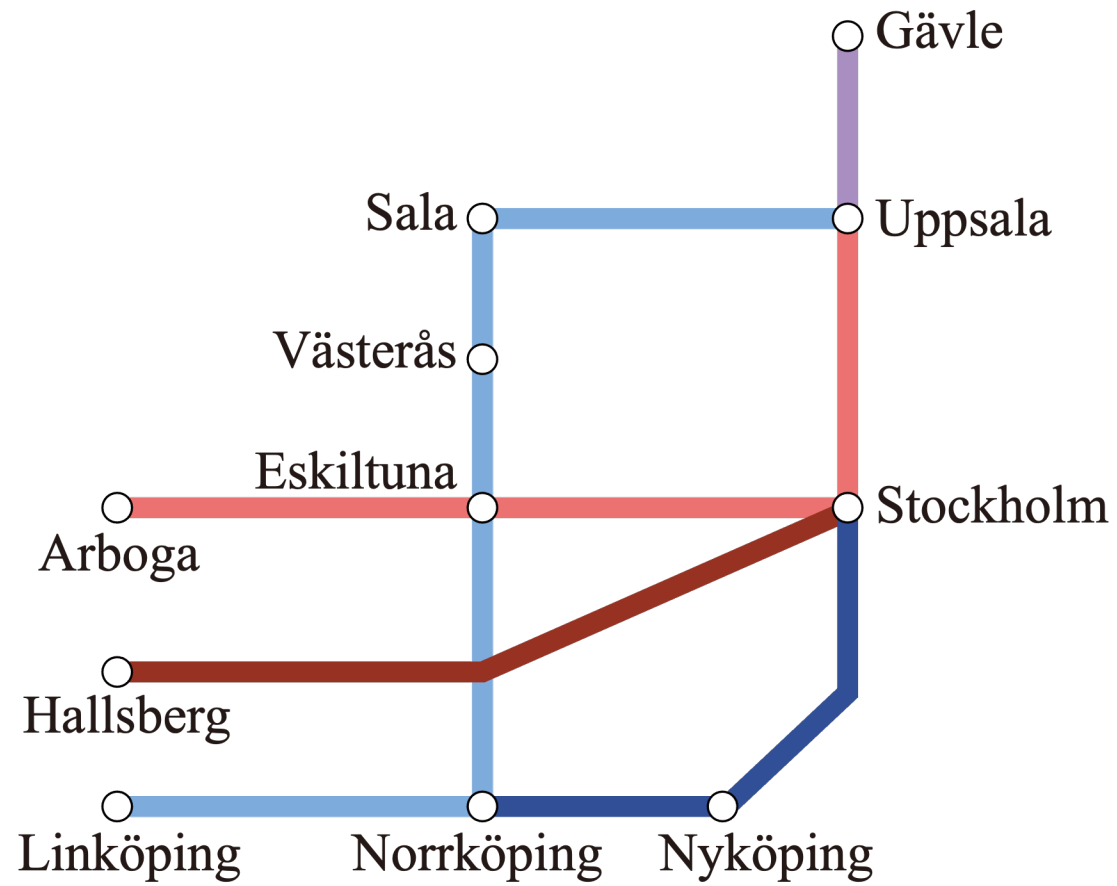
Approach: Neighboring Solutions

Strategy II: Swap with assigned tasks

- Selected unassigned task
- Another unassigned task
- Deadheading assigned task



Case Study



MILP Model vs Approach

	Data size	Method	Time	Space	Successful Assigned Rate
0.50*Large	small	Tabu-Search-based Approach	9.7 s	0.16 GB	8/18
		MILP Model (Gurobi 11.0)	0.7 h	12.24 GB	10/18
0.75*Large	medium	Tabu-Search-based Approach	12.3 s	0.18 GB	12/22
		MILP Model (Gurobi 11.0)	7.5 h	35.00 GB	17/22
One-day Schedule	large	Tabu-Search-based Approach	21.7 s	0.20 GB	29/37
		MILP Model (Gurobi 11.0)	-	out of space	-

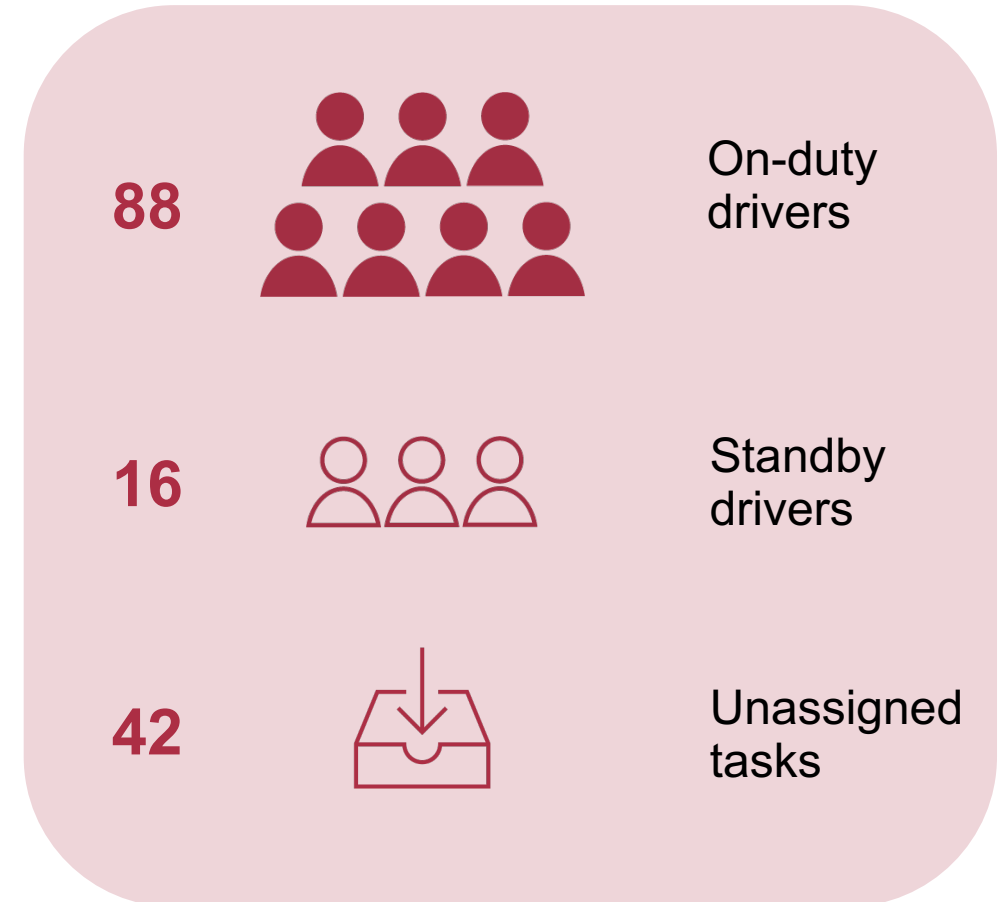
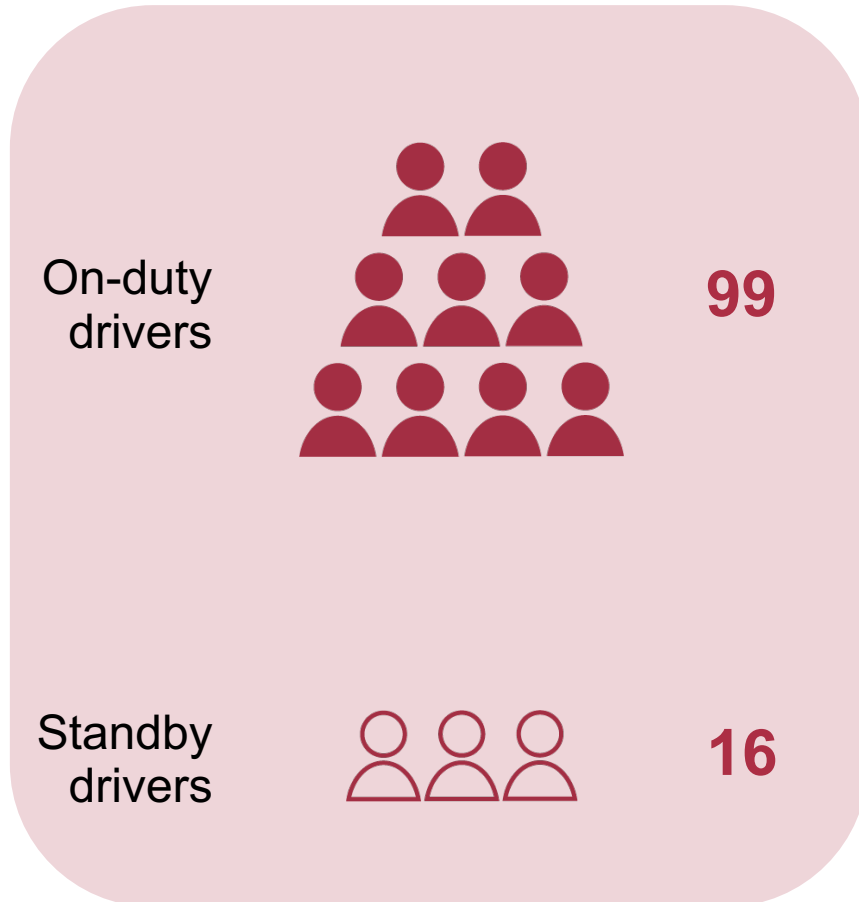
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Approach: one-day schedule



Approach: one-day schedule

Unassigned tasks

Before **42**

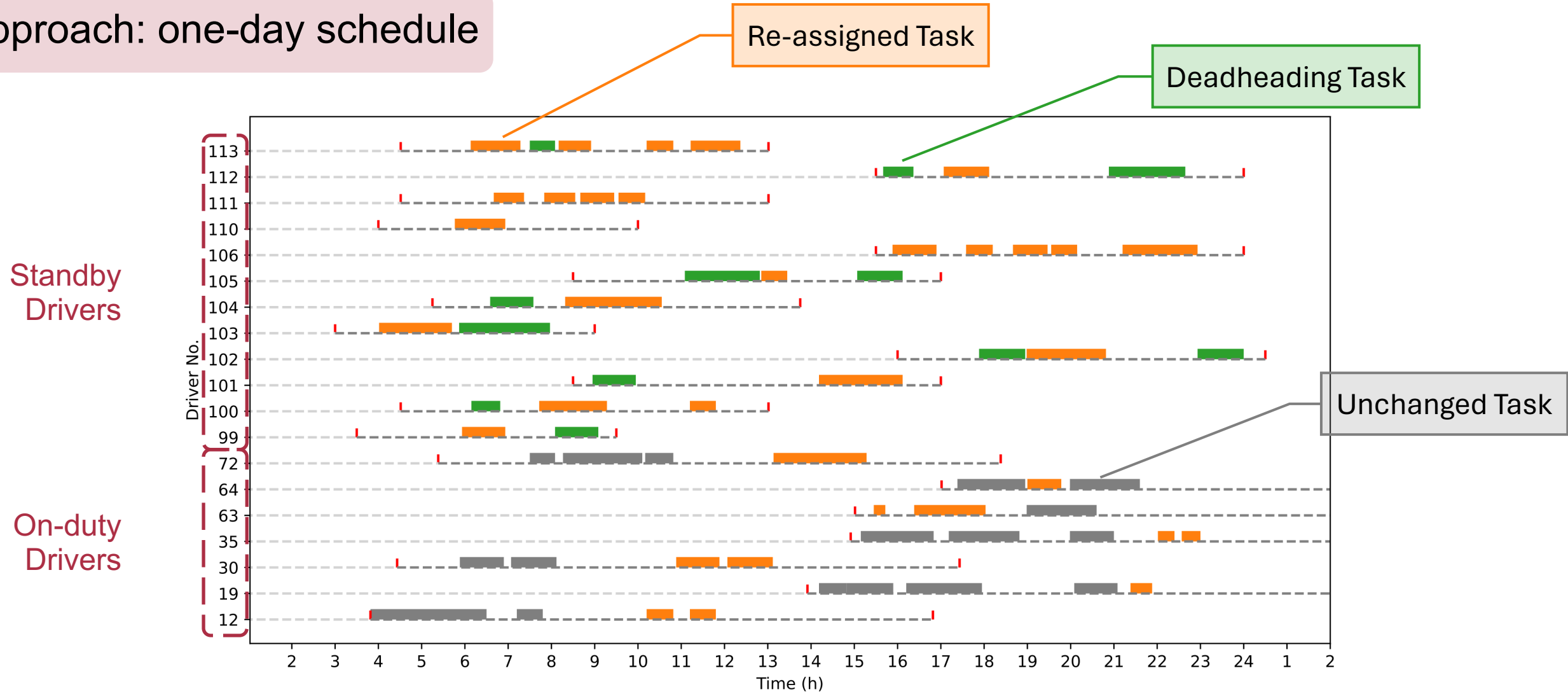


Calculation Time: 19 seconds
Calculation Space: 0.2 GB

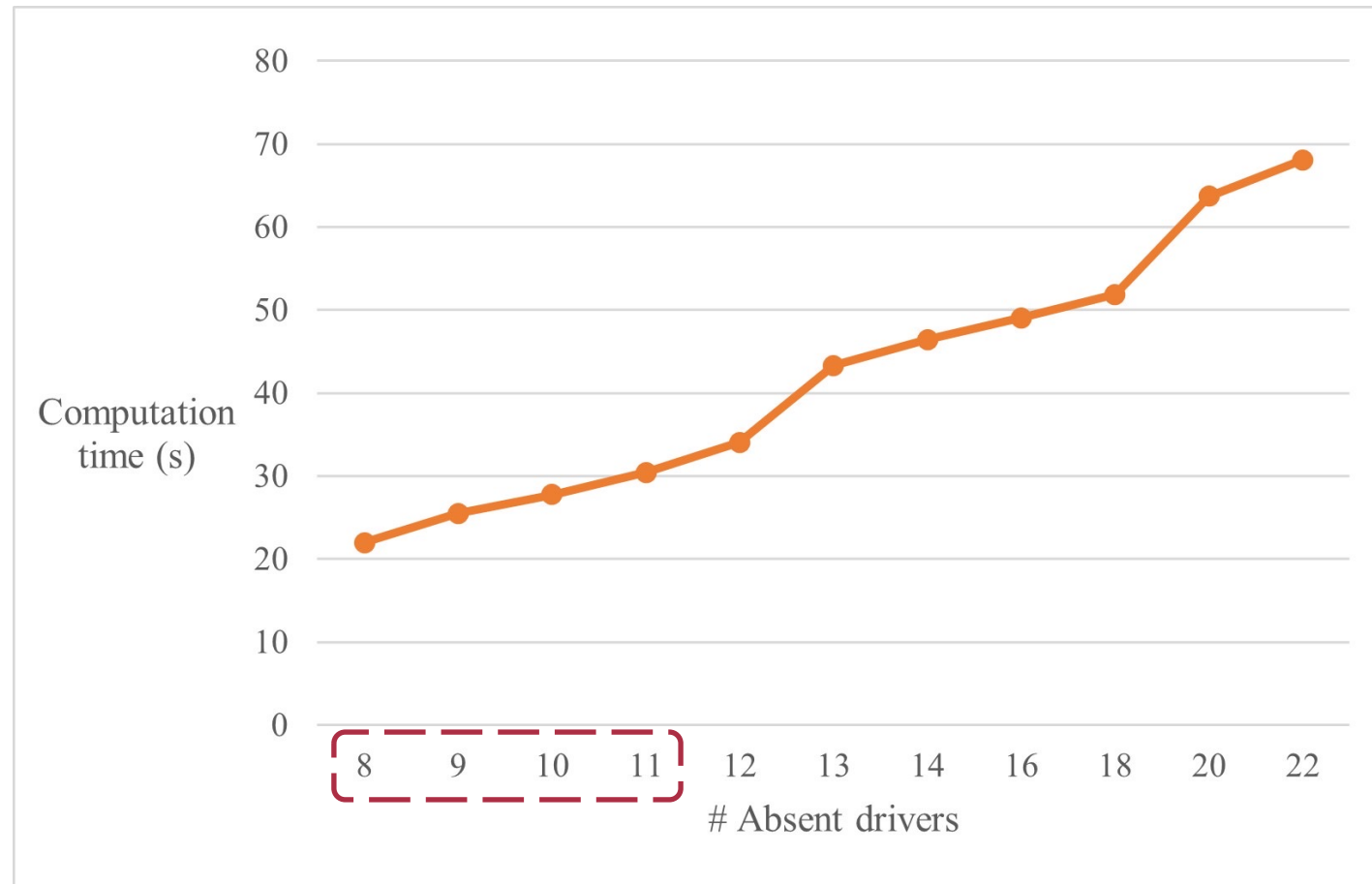
After **10**

Case Study

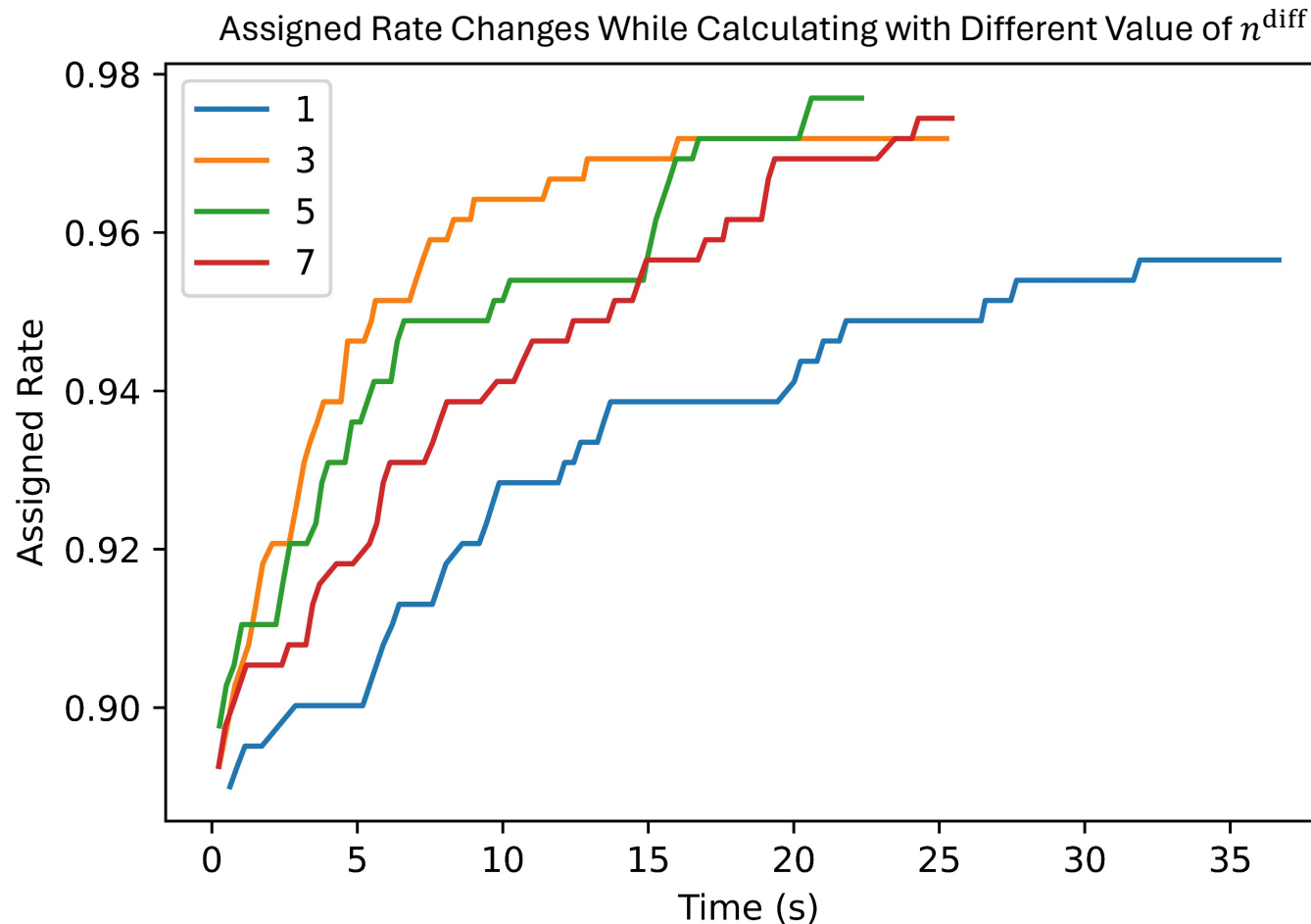
Approach: one-day schedule



Approach: performance



Approach: performance

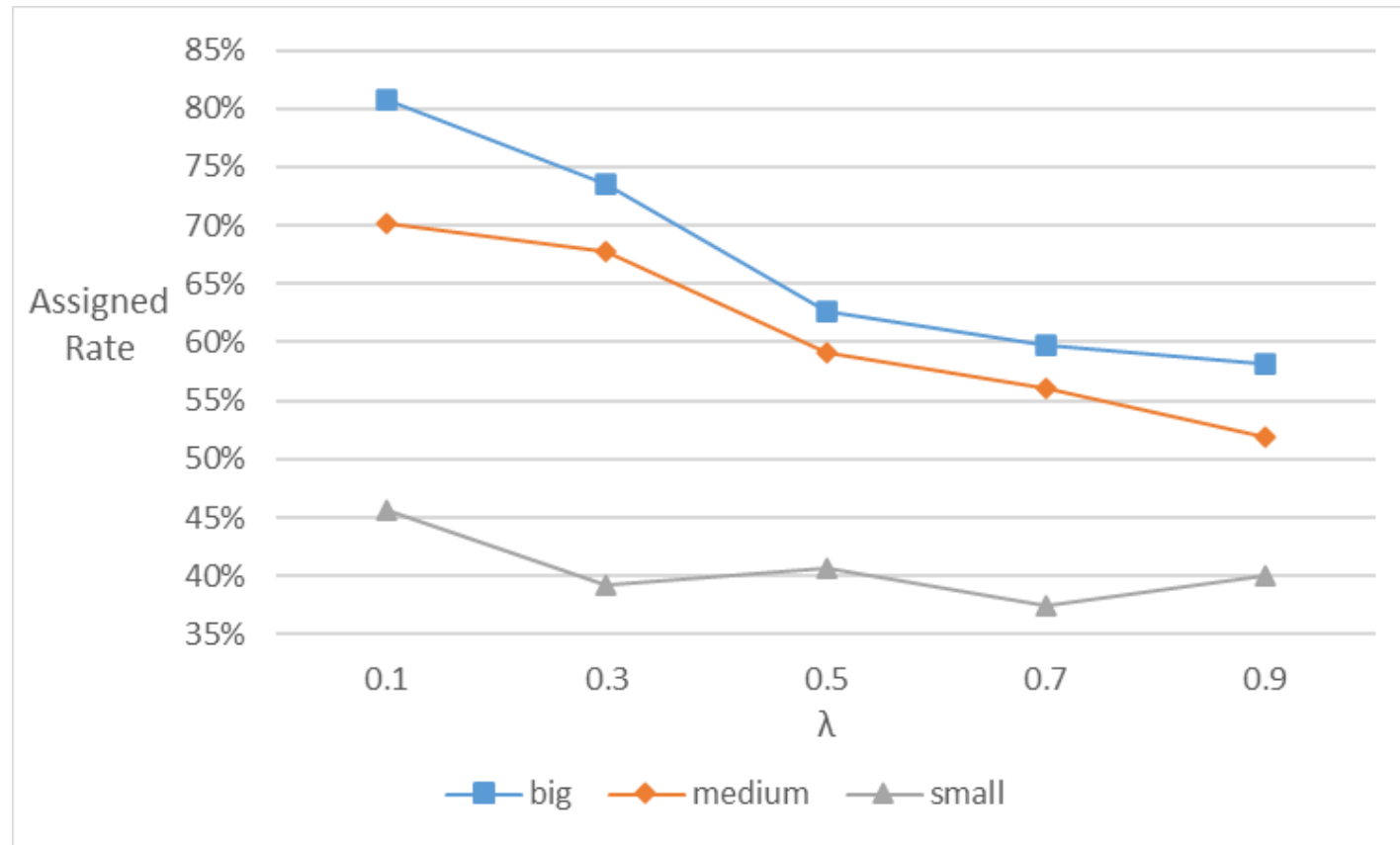


Level of freedom

n^{diff} : maximum allowed difference between # tasks unassigned from driver d and # tasks assigned to driver d .

Approach: performance

$$\text{minimize } f(x_{g,d}, z_{g,d}) = \lambda \sum_{g \in G} \sum_{d \in D \cap \hat{D}} z_{g,d} + (1 - \lambda) \sum_{d \in D^u} x_{g,d}$$





Gäller ej körning med ATC-besked

Gäller ej körning med ATC-besked

O-12a

Tack så mycket!

Mälartåg

008

STOCKHOLM C