

0. Study at a Glance

- Question:** How should trucks and on-site repairers be co-optimized to keep bikes available while limiting emissions?
- Approach:** Time-indexed MILP + Hybrid Genetic Search with Station Budget Constraint to route trucks, load/unload bikes, and allocate repair time.
- Context:** 2-hour horizon, trucks carry usable and broken bikes, repairers fix on-site; dissatisfaction and CO₂ drive the objective.
- Performance:** 4–33% better solutions than Gurobi in 6–51s; scales to 500 stations.
- Key takeaway:** On-site repairers pay off when damage ratio $\geq 30\%$ and repair time < 10 minutes; otherwise trucks alone suffice.

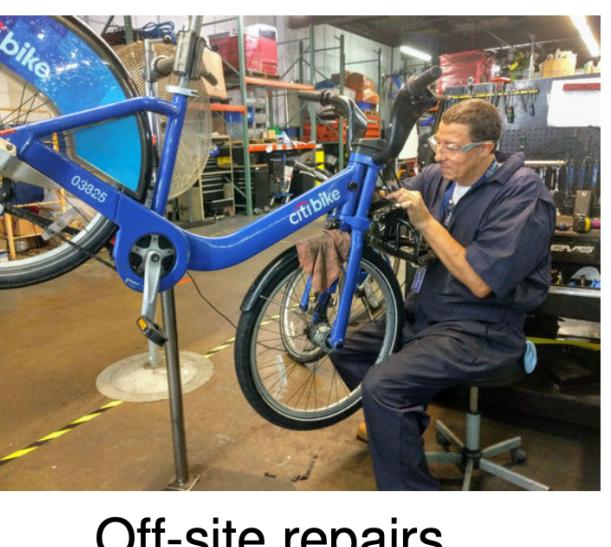
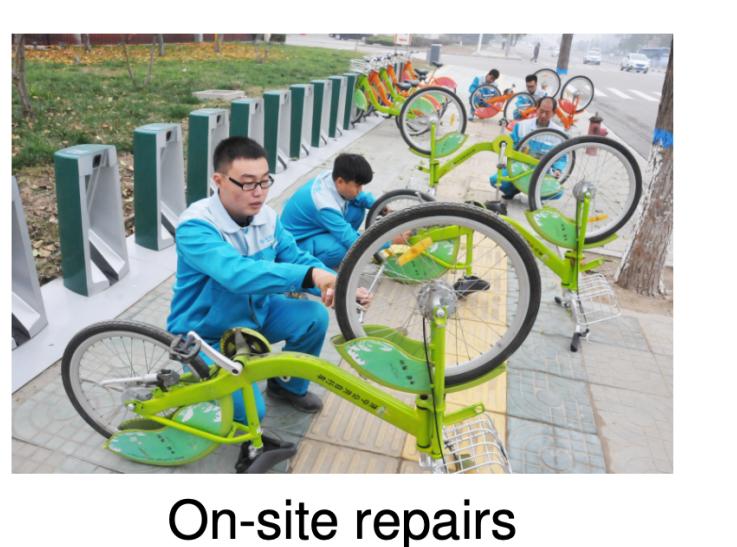
1. Background & Motivation

The Problem: Bike Sharing Systems (BSS) face two critical challenges:

- Demand-Supply Imbalance:** Stations become empty or full, causing user dissatisfaction
- Broken Bikes:** Barcelona: 12% severe + 55% light damage; NYC: 2% need daily repairs

Research Gap: No existing study jointly optimizes:

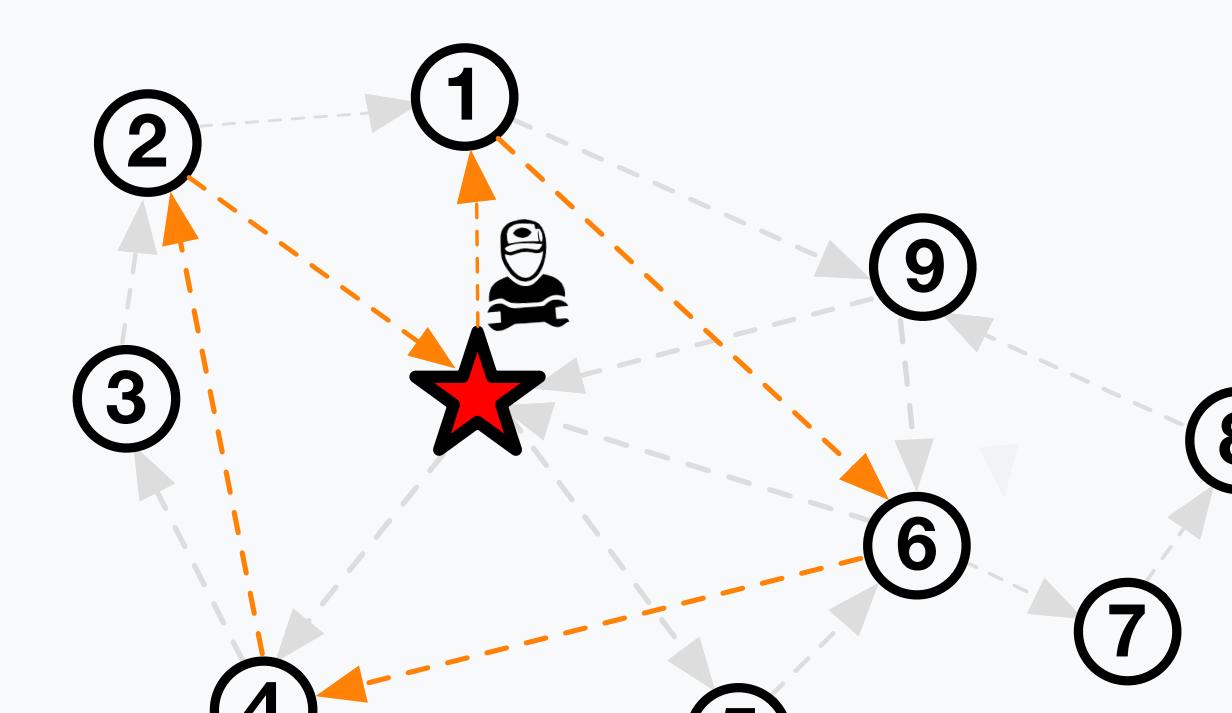
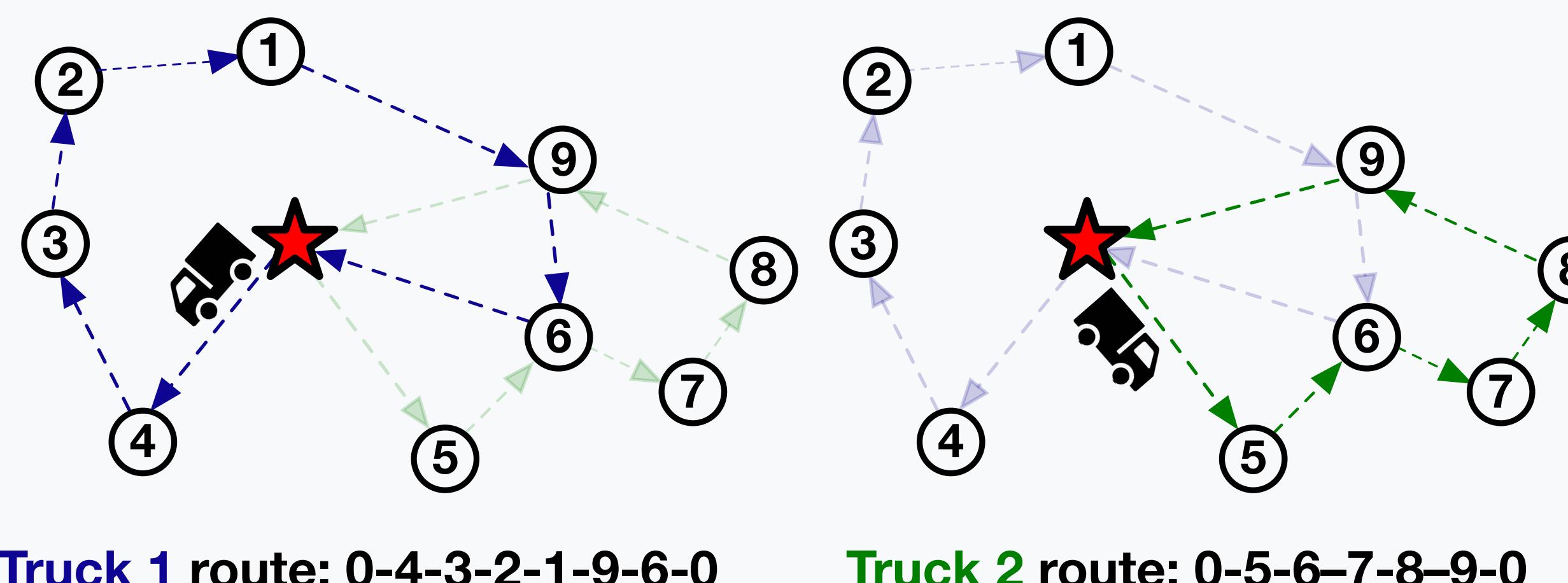
- Vehicle-based repositioning (usable & broken bikes)
- Labor-based on-site repairs



2. Contributions

- Novel Problem:** First study to integrate on-site repairs with vehicle-based repositioning operations
- MILP Model:** Time-indexed formulation minimizing user dissatisfaction + CO₂ emissions
- HGSADC-SBC:** Hybrid Genetic Search algorithm with Station Budget Constraint for large-scale instances
- Managerial Insights:** Cost-effectiveness analysis framework for repairer deployment decisions

3. Problem Illustration



6. HGSADC-SBC Algorithm

Hybrid Genetic Search with Adaptive Diversity Control + Station Budget Constraint

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Algorithm 1: HGSADC-SBC
1 Initialization: Generate a population  $Pop_t$  consisting of a feasible subpopulation  $Pop_f$  and an infeasible subpopulation  $Pop_i$ 
2 Set  $iter\_no\_imp = 0$  and  $current\_best\_value = +\infty$ 
3 while  $iter\_no\_imp < I_{\max}$  and  $cpu\_time < T_{\max}$  do
4   Randomly select two individuals  $Ind_1$  and  $Ind_2$  from  $Pop$ 
5   Generate the routes of offspring  $C_1$  and  $C_2$  using ordered crossover on the routes of  $Ind_1$  and  $Ind_2$ , then apply the SBC heuristic on the routes to determine the (un)loading and repairing quantities for  $C_1$  and  $C_2$ 
6   Educate on both  $C_1$  and  $C_2$  by applying local search procedures to generate new routes and using the SBC heuristic to determine the (un)loading and repairing quantities
7   if  $C_i$  ( $i = 1, 2$ ) is feasible then
8     Add  $C_i$  to the feasible subpopulation  $Pop_f$ 
9   else
10    Add  $C_i$  to the infeasible subpopulation  $Pop_i$  and repair  $C_i$  with the probability  $P_{\text{repair}}$  using local search procedures and the SBC heuristic
11   end if
12   if the individuals become feasible after repairs then
13     The repaired individuals are added to the feasible subpopulation  $Pop_f$ 
14   else
15     The repaired individuals are discarded, and the original infeasible solution is kept in the infeasible subpopulation  $Pop_i$ 
16   end if
17   if the size of  $Pop_f$  or  $Pop_i$  reaches the maximum subpopulation size then
18     Select survivors from the corresponding subpopulation
19   end if
20   Set  $bs$  = the solution with the best fitness value from  $Pop_f$  and set  $bv$  = the fitness value of  $bs$ 
21   if  $bv \geq current\_best\_value$  then
22     Set  $iter\_no\_imp = iter\_no\_imp + 1$ 
23   Else
24     Set  $current\_best\_value = bv$ ,  $current\_best\_sol = bs$ , and  $iter\_no\_imp = 0$ 
25   end if
26   if  $iter\_no\_imp \geq I_{\max}$  then
27     Diversify the population consisting of  $Pop_f$  and  $Pop_i$  and adjust the penalty parameter
28   end if
29   end while
30 return  $current\_best\_sol$ ,  $current\_best\_value$ 

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7. Station Budget Concept (SBC)

Key Innovation: Prevents over-allocation of service time to early stations

Benefit-Cost Ratio Function (BCRF):

$$BCRF_s^t(p, b) = \frac{F_s(p, b) - F_s(q, 0)}{t^{load} \cdot (|q - p| + b)}$$

- Quantifies dissatisfaction reduction per unit service time
- Higher BCRF stations receive larger time budgets
- Unused time redistributed dynamically to later stations

8. BCRF Illustration

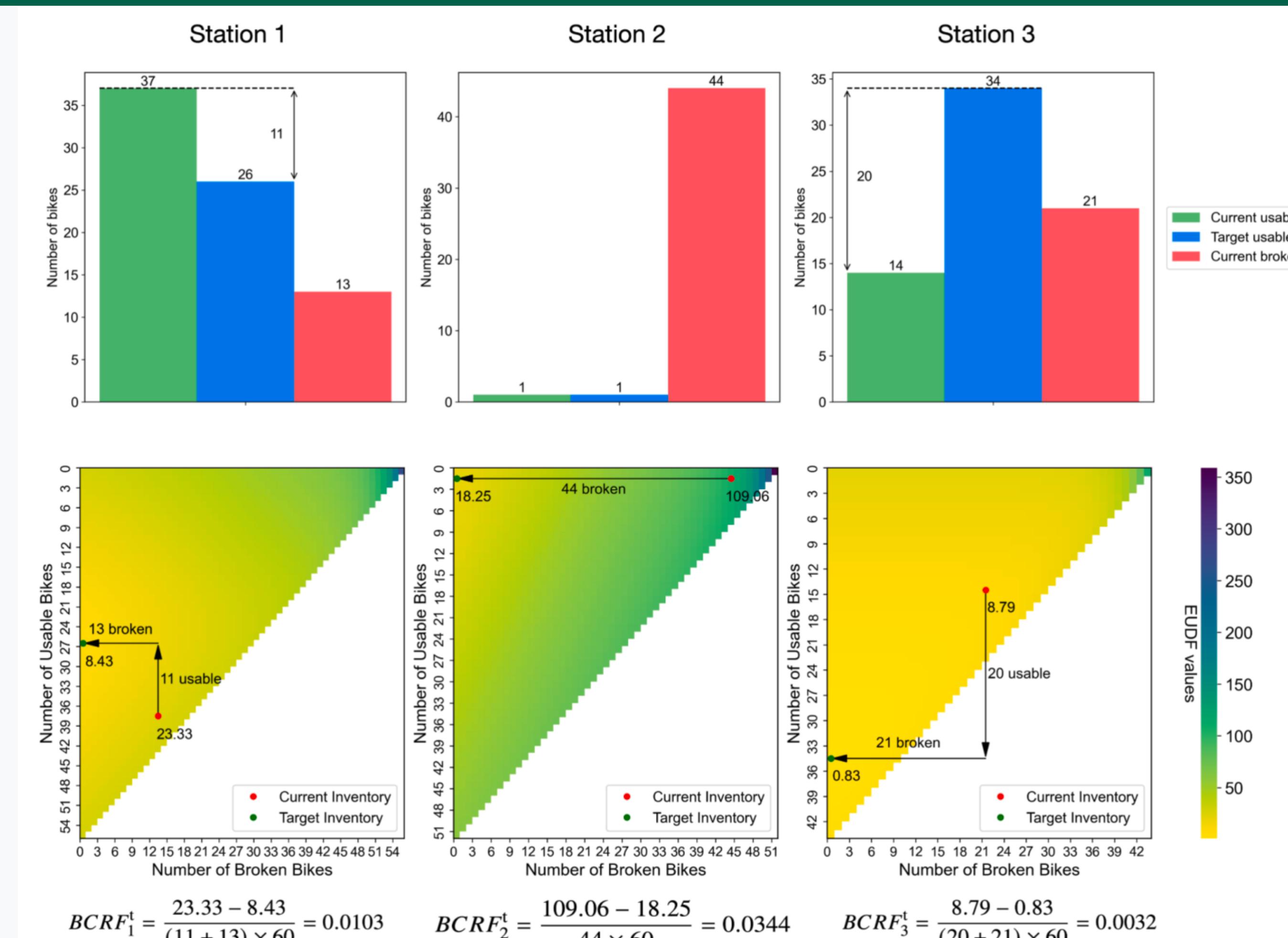


Fig. 4.1: BCRF calculation for 3 stations. Top: inventory status; Bottom: EUDF heatmaps.

9. Computational Results

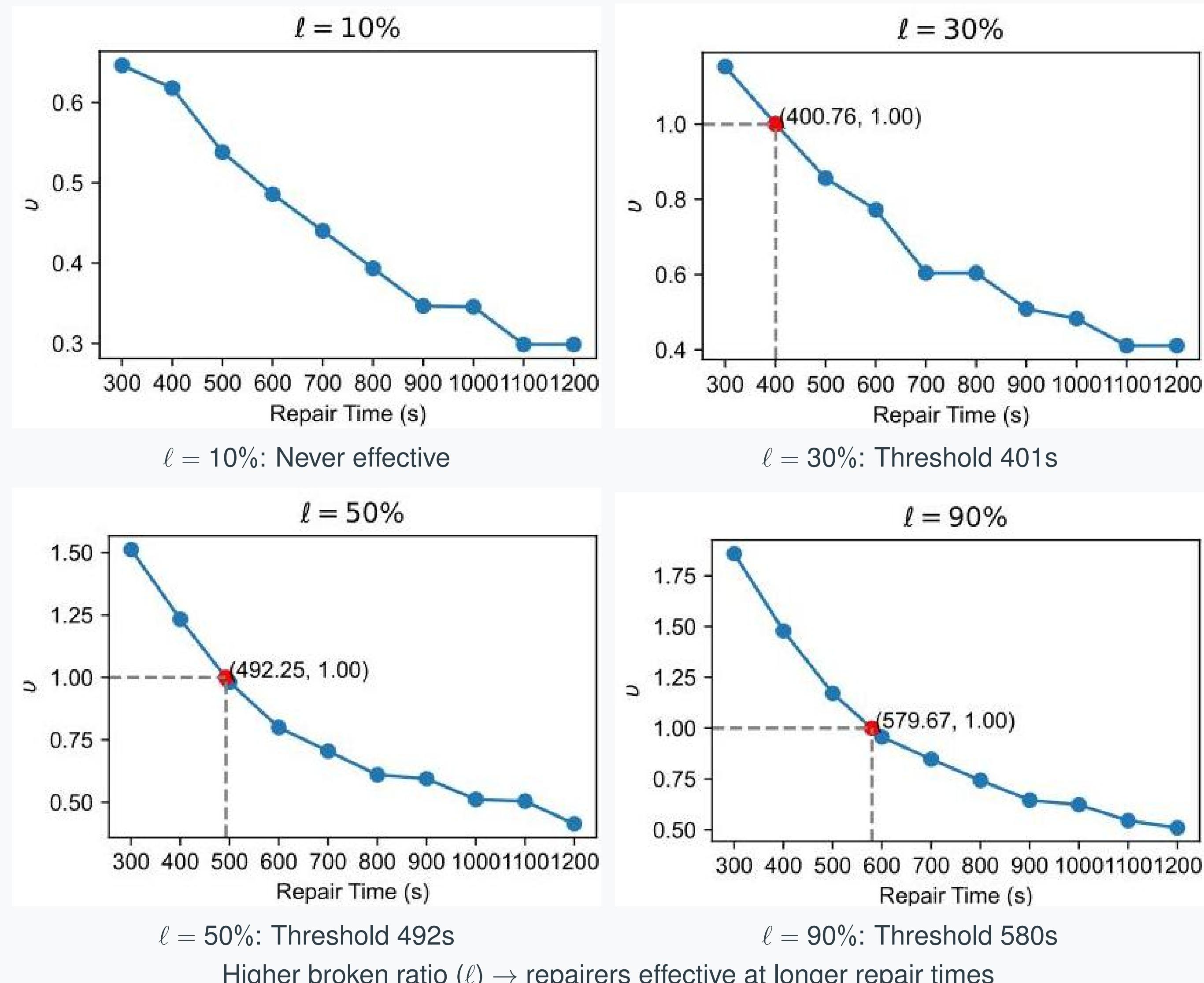
Comparison vs Gurobi (T=2h, $\tau=600$ s, 20 runs)

10. Cost-Effectiveness of Repairers

Indicator ν : Ratio of dissatisfaction reduction to repairer wage

$$\nu = \frac{c^p \sum_{i \in S} (\hat{F}_i - \bar{F}_i)}{\omega \cdot T / 3600}$$

$\nu > 1$: Cost-effective $\nu < 1$: Not cost-effective



11. Key Managerial Insights

- Critical Threshold:** Repair time > 645 s \Rightarrow never cost-effective
- Low damage ($\ell < 20\%$):** Trucks alone sufficient; repairers not justified
- High damage ($\ell \geq 30\%$):** On-site repairers become economically justified
- Preventive maintenance** crucial to keep repair times below threshold

12. Conclusions

- HGSADC-SBC is **scalable** for real-world BSS (up to 500 stations)
- Station Budget Constraint effectively balances time allocation across route
- Dynamic repairer deployment should adapt to daily damage levels

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References

Hu, R., Szeto, W.Y., & Ho, S.C. (2025). Repositioning in bike sharing systems with broken bikes considering on-site repairs. *Transportation Research Part E*, 104155.

Scan for full paper:

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