

Incentivizing the Use of Bike Trailers for Dynamic Repositioning in Bike Sharing Systems

Supriyo Ghosh, Pradeep Varakantham

School of Information Systems
Singapore Management University



The 27th International Conference on Automated
Planning and Scheduling (ICAPS, 2017)

Motivation: Bike Sharing System

■ Bike Sharing Systems

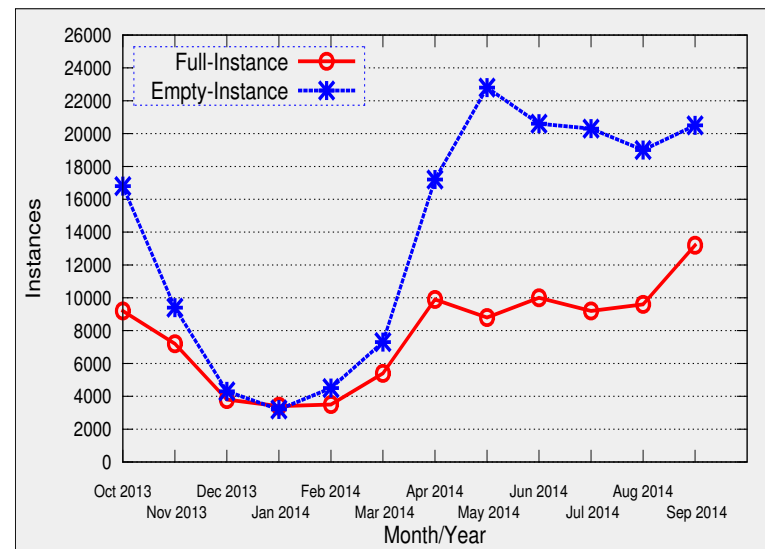
- 1,070 active systems all over the world.
- Attractive alternative to private vehicles
- Reduce traffic congestion, green house gas emission and air pollution.



■ **Problem:** Starvation or congestion of bikes at stations

- Increase usage of private vehicle and carbon emission.

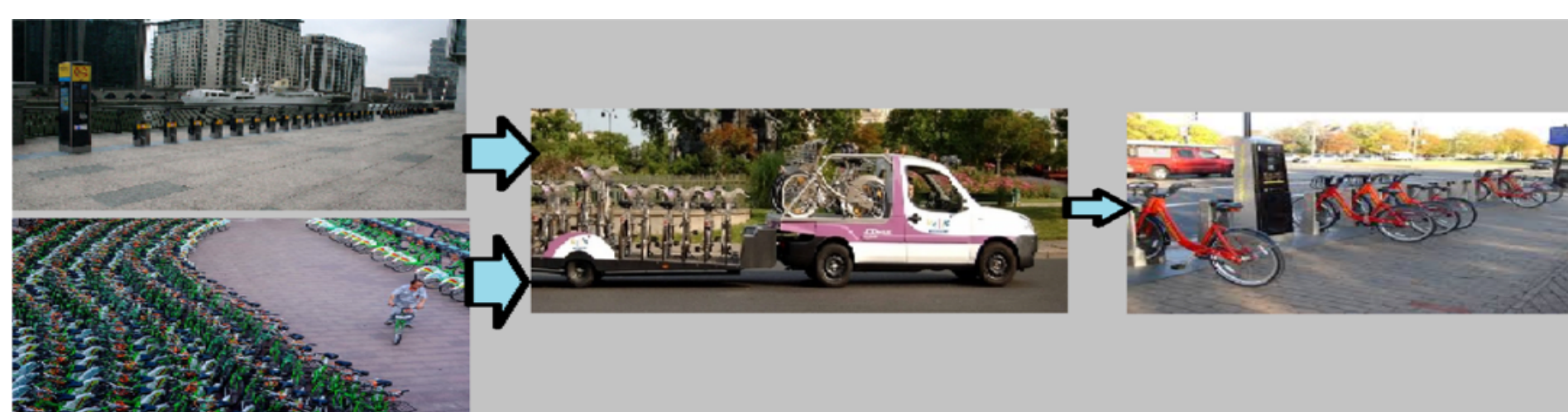
■ **Goal:** Repositioning of bikes during the day to address availability issues.



Starvation/congestion in Capitalbikeshare

Background

- Repositioning using trucks
 - Static Redeployment - once at the end of day
 - Dynamic Redeployment - matching of producer and consumer station
 - Problems with trucks for repositioning:
 - Incur substantial routing and labor costs
 - Increase carbon footprint
 - Limited number of trucks



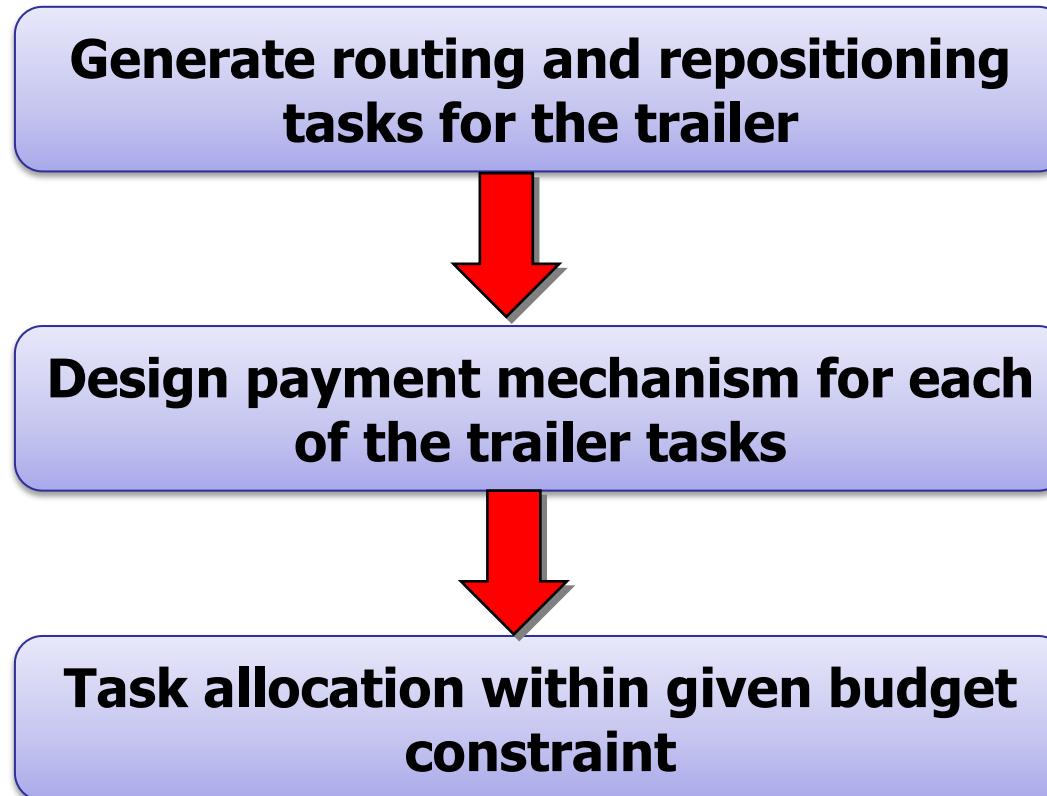
Contributions

- Repositioning with bike-trailers – Moving beyond trucks
- **Challenges:**
 - Physical limitations of the routes for trailers.
 - Limited budget: Employing staffs for trailer is not feasible.
- **Overall Goal:**
 - Develop a self-sufficient system of rebalancing using bike trailers
 - Crowdsource repositioning tasks to customers within a given budget



Solution Approach

- Dynamic Routing and Repositioning Problem using Trailers (DRRPT)
 - Inputs: $\langle S, \mathcal{V}, C^\#, C^*, D^{\#,0}, \{\sigma_v^0\}, P, F, B \rangle$
 - Outputs: Repositioning strategy for trailers & allocation of tasks



Task Generation for Bike Trailers

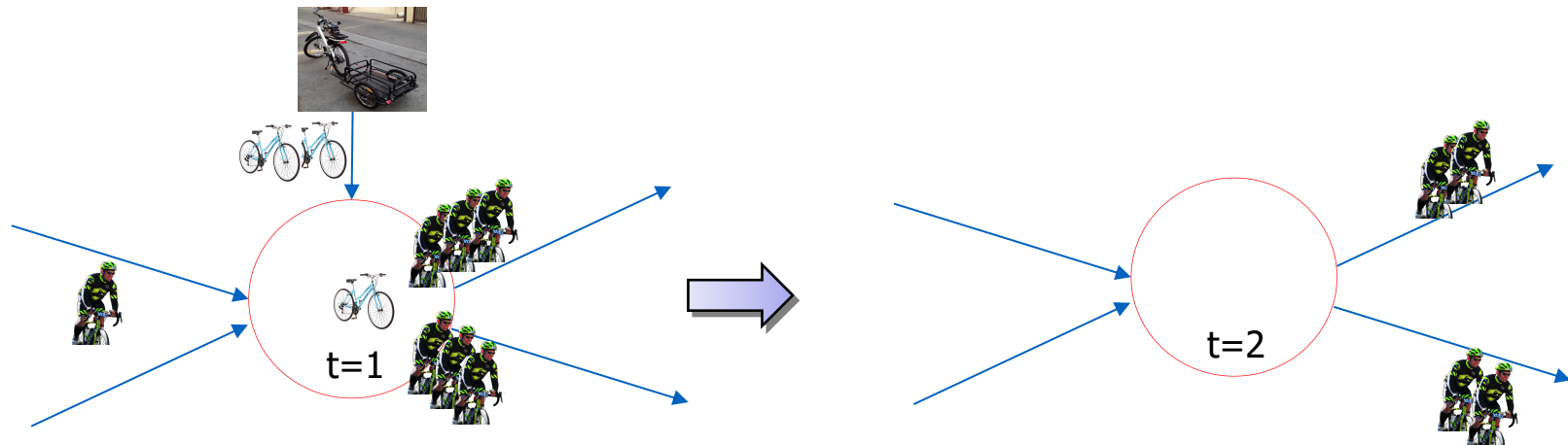
$$\min_{y^+, y^-} \sum_{k, s} L_s^k$$

Minimise Lost demand over k scenarios

Constraints:

- Compute lost demand

$$s.t. \quad L_s^k \geq \underbrace{\sum_{s'} F_{s, s'}^k}_{\text{Demand}} - \underbrace{\left(d_s^\# + \sum_v (y_{s, v}^- - y_{s, v}^-) \right)}_{\text{Supply}}, \forall k, s$$



$$\text{Lost Demand} = 6 - 4 = 2$$

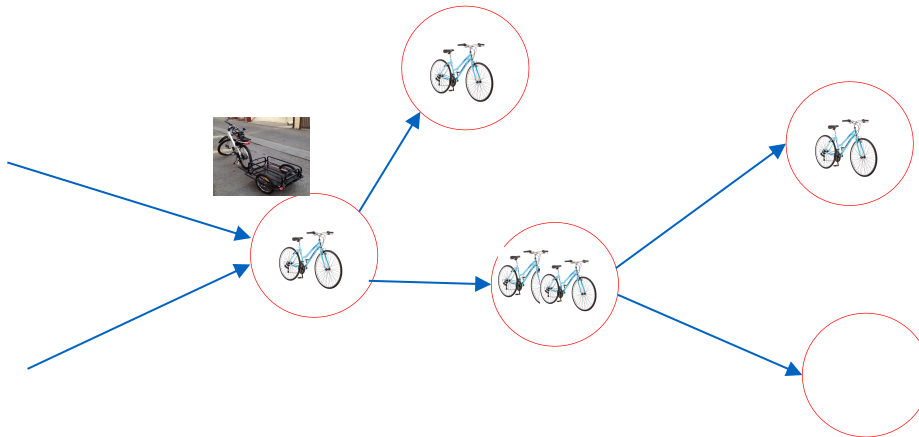
Task Generation for Bike Trailers

$$\min_{y^+, y^-} \sum_{k, s} L_s^k$$

Minimise Lost demand over k scenarios

■ Constraints:

- Compute lost demand
- Trailer should pickup bikes from the neighbor of it origin station



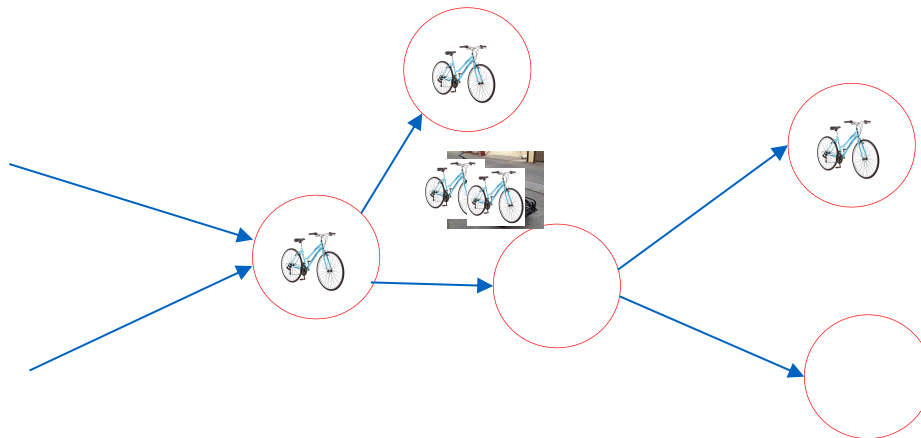
Task Generation for Bike Trailers

$$\min_{y^+, y^-} \sum_{k, s} L_s^k$$

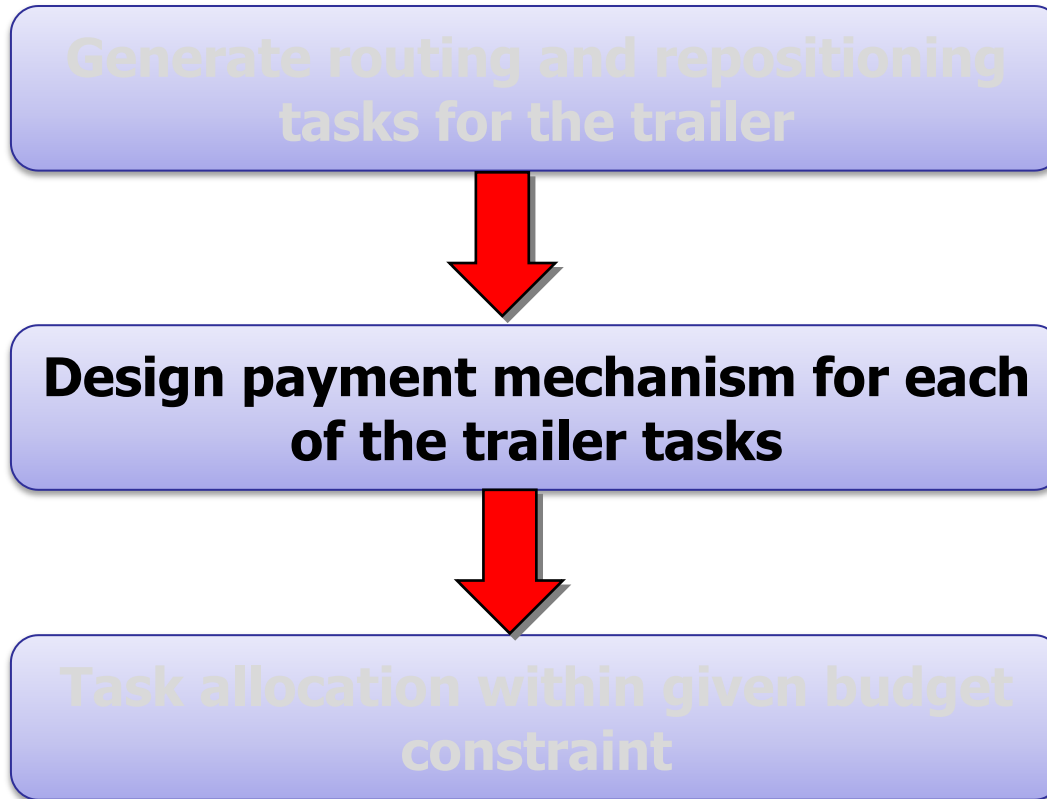
Minimise Lost demand over k scenarios

■ Constraints:

- Compute lost demand
- Trailer should pickup bikes from the neighbor of it origin station
- Trailer should drop-off exact number of bikes at its destination
 - While satisfies the physical limitation of the trailer routes



Solution Approach



Mechanism Design

- Compute the value for task of trailer v

$$U(v) = \frac{\xi}{K} \sum_{k,s} \left[\underbrace{\min \left(\max \left(\sum_{s'} F_{s,s'}^k - d_s^\#, 0 \right), y_{s,v}^+ \right)}_{\text{Lost Demand Saved}} - \underbrace{\min \left(\max \left(y_{s,v}^- - \left(d_s^\# - \sum_{s'} F_{s,s'}^k \right), 0 \right), y_{s,v}^- \right)}_{\text{Lost Demand Created}} \right]$$

Unit value of LD (points to ξ)
 #Scenarios (points to K)
 Lost Demand (points to $d_s^\#$)

Assumption: Set of bidders for different tasks are **pairwise disjoint**

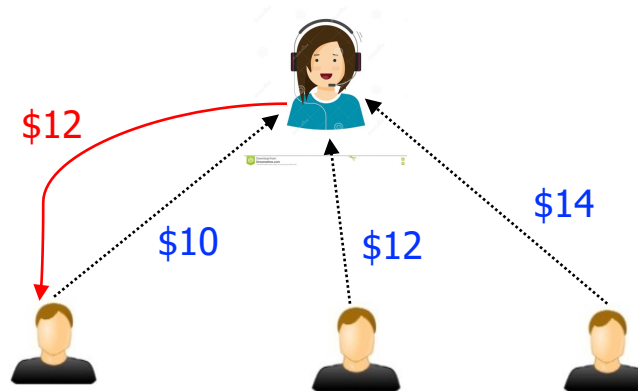
Observation: Tasks are primarily independent but coupled by the central budget constraint.

Mechanism Design

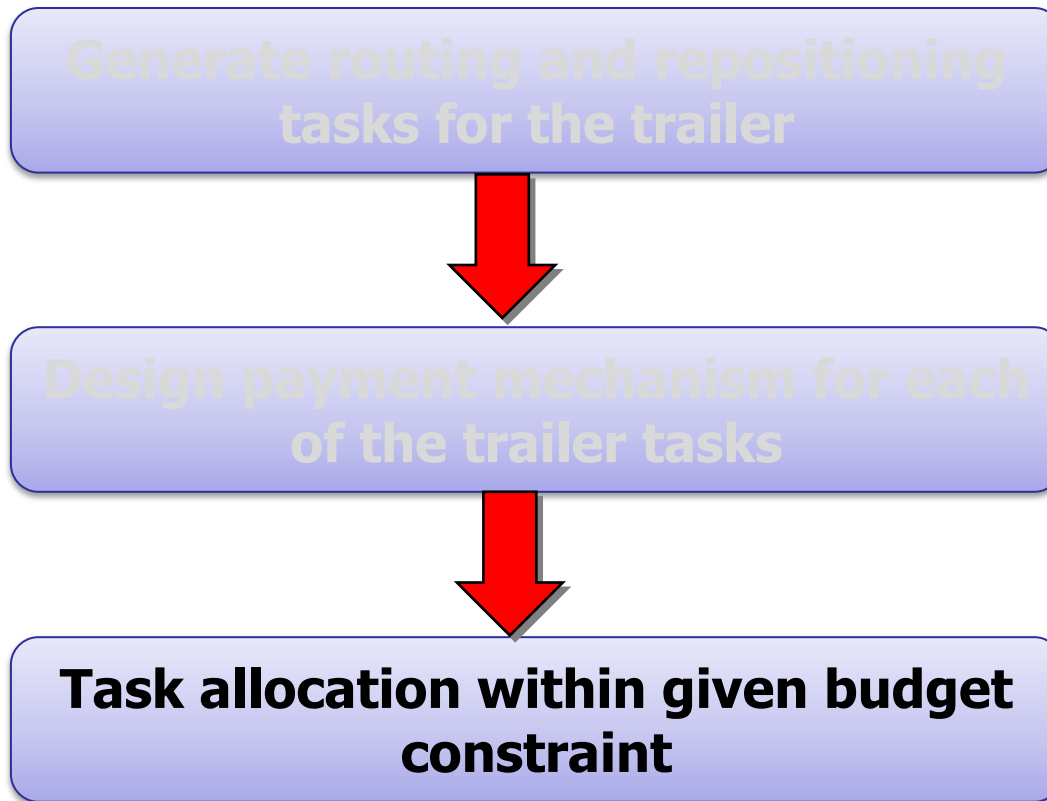
- Mechanism for single task v (N_v users bid for the task):
 - Collect the bids from user i , $C_i(v)$ privately
 - Make payment using standard VCG mechanism

$$\lambda_i^*(v) = \left\{ \begin{array}{ll} 1 & \text{if } i = \operatorname{argmax}_{j \in N_v} [U(v) - C_j(v)] \\ 0 & \text{Otherwise} \end{array} \right\}$$

$$P_i(v) = \lambda_i^*(v) \left[\min_{i \neq j} C_j(v) \right] \leftarrow \text{Second lowest bid}$$



Solution Approach



Task Allocation within Budget Constraint

- Goal: Allocate a set of tasks $\mathcal{T} = \{1, \dots, \mathcal{V}\}$, each having a valuation of $U(v)$ and payment of $P(v)$, within a central budget B .
- Exactly equivalent to binary knapsack problem
 - $x(v)$: Set to 1 if task v is allocated the payment and 0 otherwise.

$$\max_x \sum_{v \in \mathcal{T}} x(v) \cdot U(v) \leftarrow \text{Maximise the total valuation of center}$$

$$s.t. \sum_{v \in \mathcal{T}} x(v) \cdot P(v) \leq B \leftarrow \text{Ensure that the total payment is bounded by the central budget}$$

Proposition: The mechanism for task allocation for the trailers in bike sharing system is incentive compatible (IC) and truthful.

Experimental Setup



- **Dataset:**

- Hubway (95 base stations & 3 vehicles)
 - 1 quarter of trip history data
 - Planning period: 6AM-12PM (each decision epoch is 30 minutes)

- **Demand Scenarios:**

- Real-world data for 60 weekdays (Training/Testing: 20/40)
- Demand follows *Poisson* at origin station (Training/Testing: 30/70)
- Demand follows *Poisson* for each OD pair (Training/Testing: 30/70)

- **Evaluation Metrics:** Average lost demand over testing scenarios

- **Benchmark approaches:**

- **Static repositioning:** Redeployment at the end of day
- **Online repositioning using truck:** Adapted from *Ghosh et. al., 2016*

Experimental Results

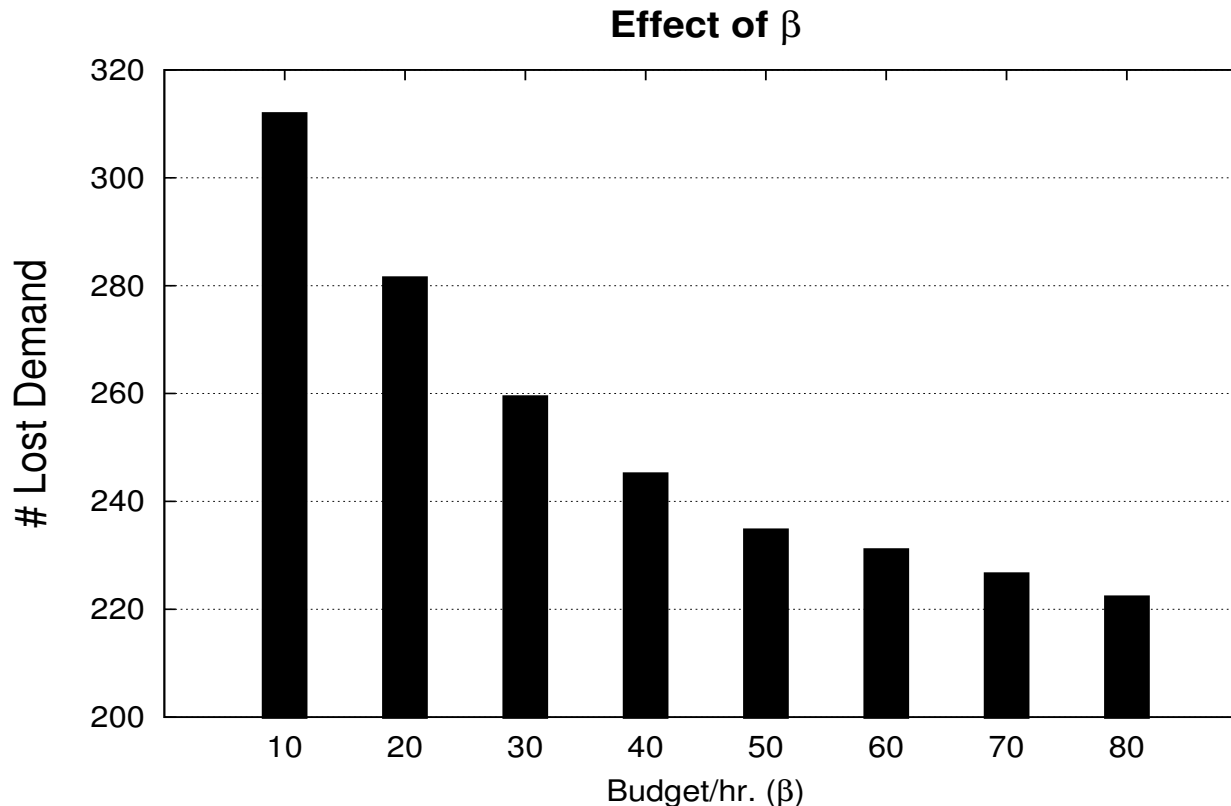
- Effect of bidding parameters on lost demand (LD)
 - Hourly Budget
 - % of users interested in bidding
 - Ratio of lower and upper bounds of bid (α)

- Performance comparison with the benchmark approaches
 - Reduction in expected lost demand over three sets of demand scenarios.

Runtime performance.

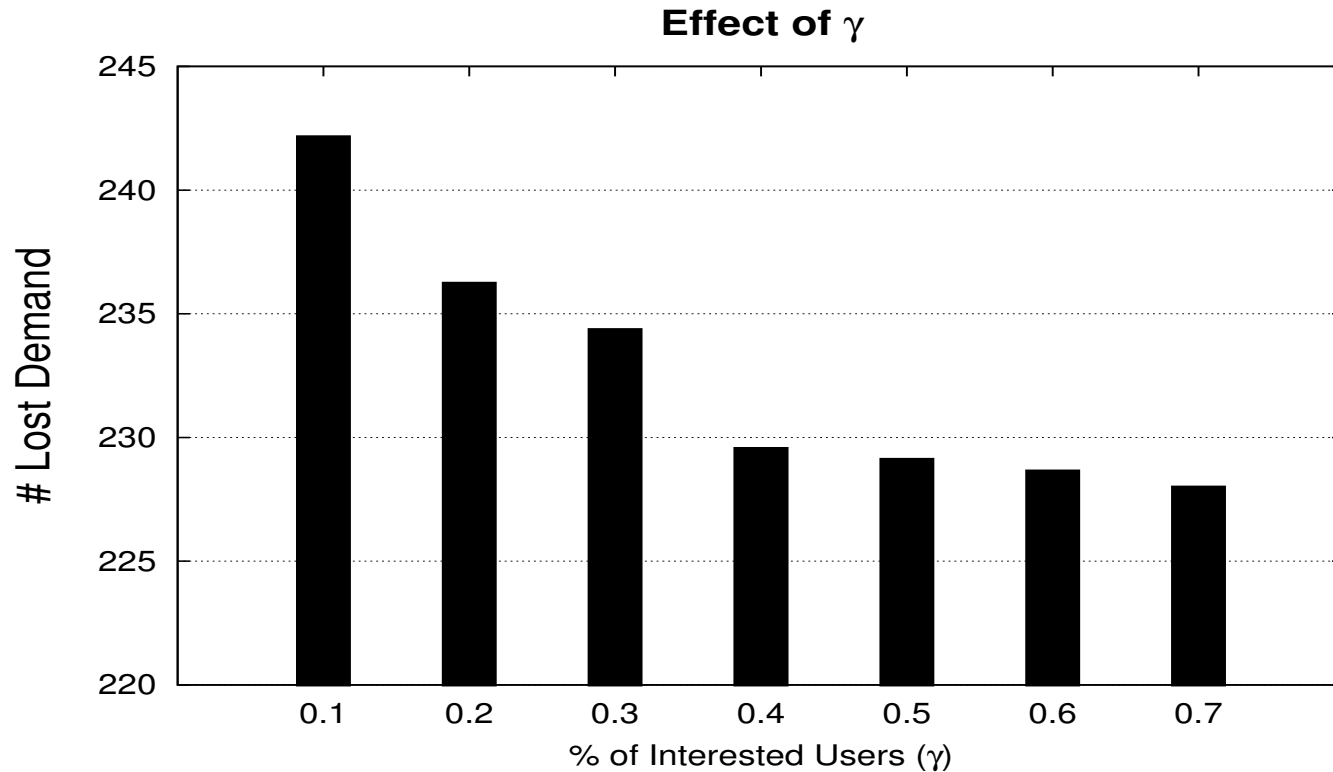
Experimental Results

- Effect of bidding parameters on lost demand (LD)
 - Hourly Budget: LD decreases as the budget increases



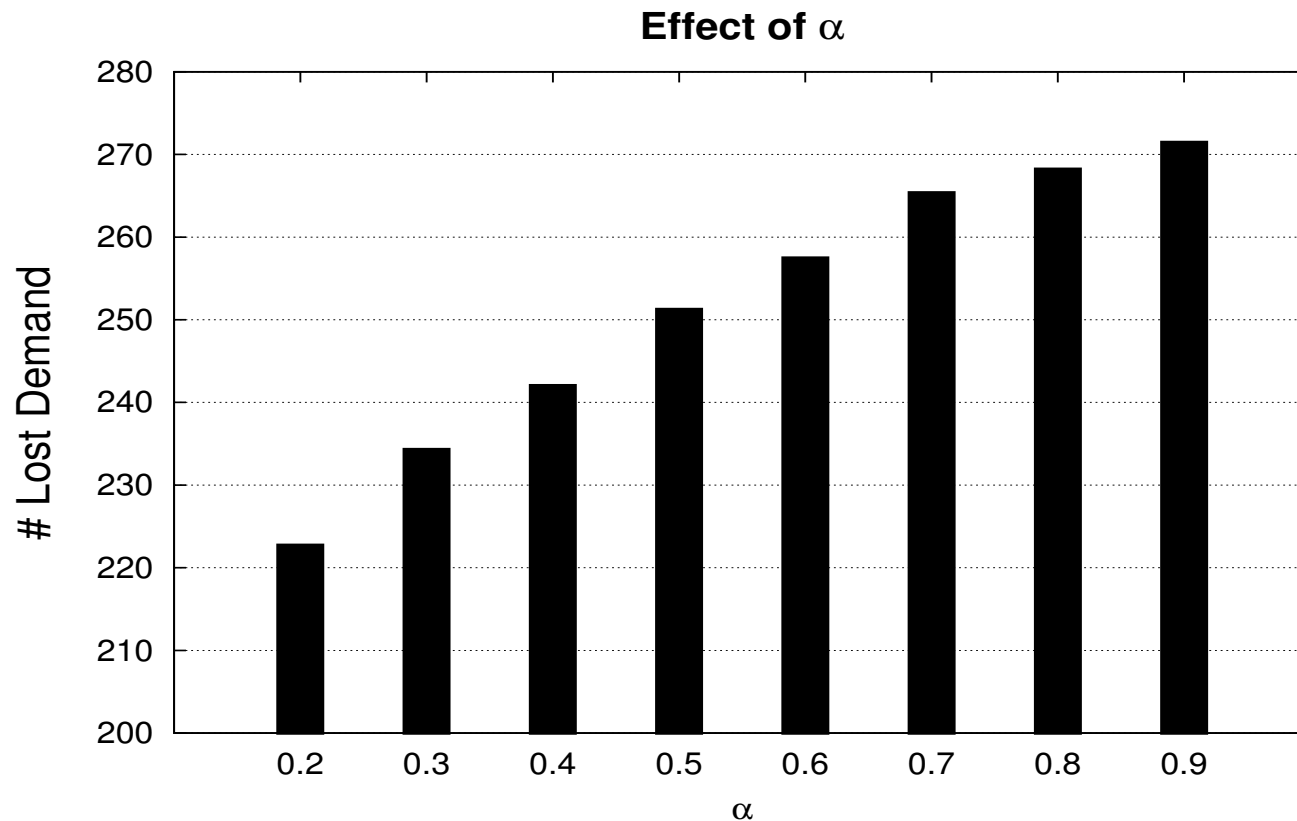
Experimental Results

- Effect of bidding parameters on lost demand (LD)
 - % of users interested in bidding: LD decreases as number of bids increases.



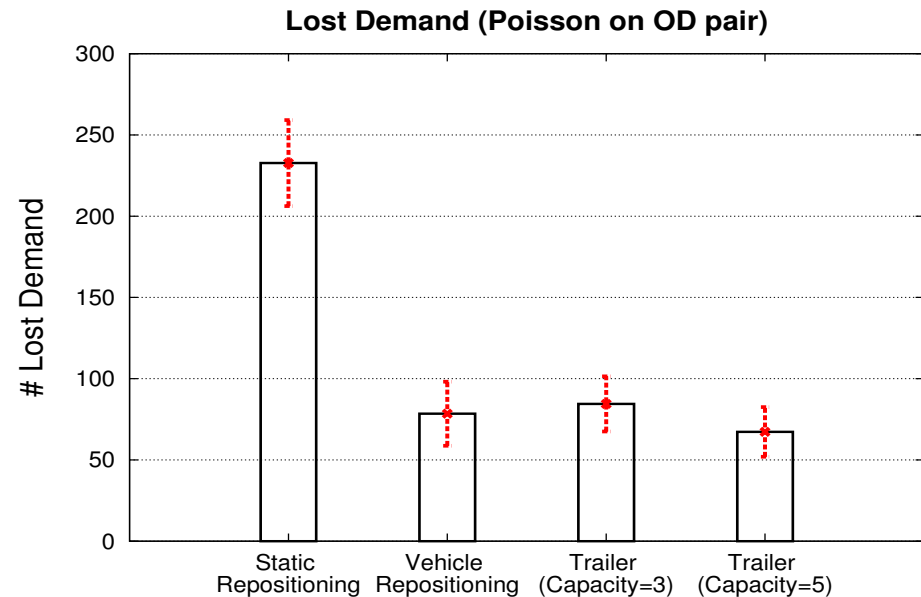
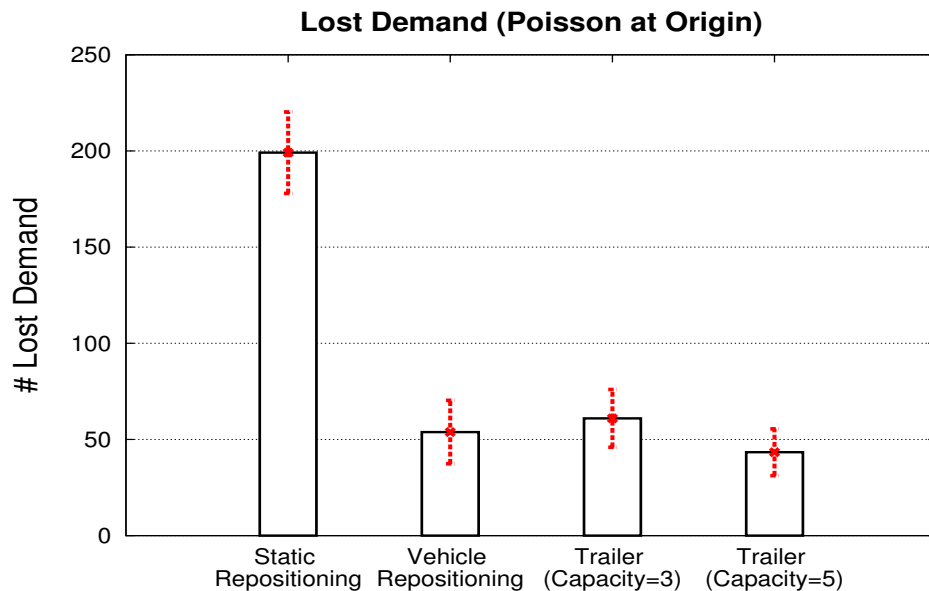
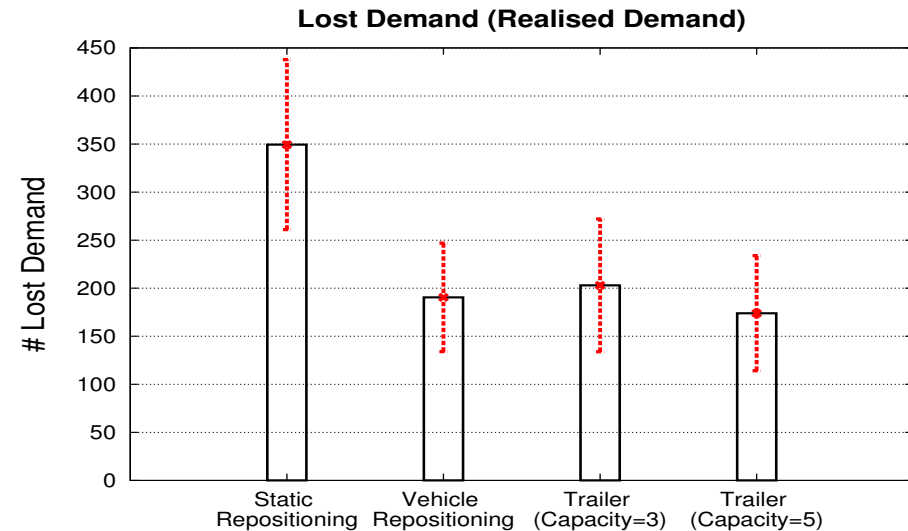
Experimental Results

- Effect of bidding parameters on lost demand (LD)
 - Ratio of lower and upper bounds of bid (α): LD increases monotonically with α



Experimental Results

- Performance comparison
 - 10 trailers with capacity 3 reduces the lost demand by 41%
 - 10 trailers with capacity 5 perform better than repositioning with 3 vehicles

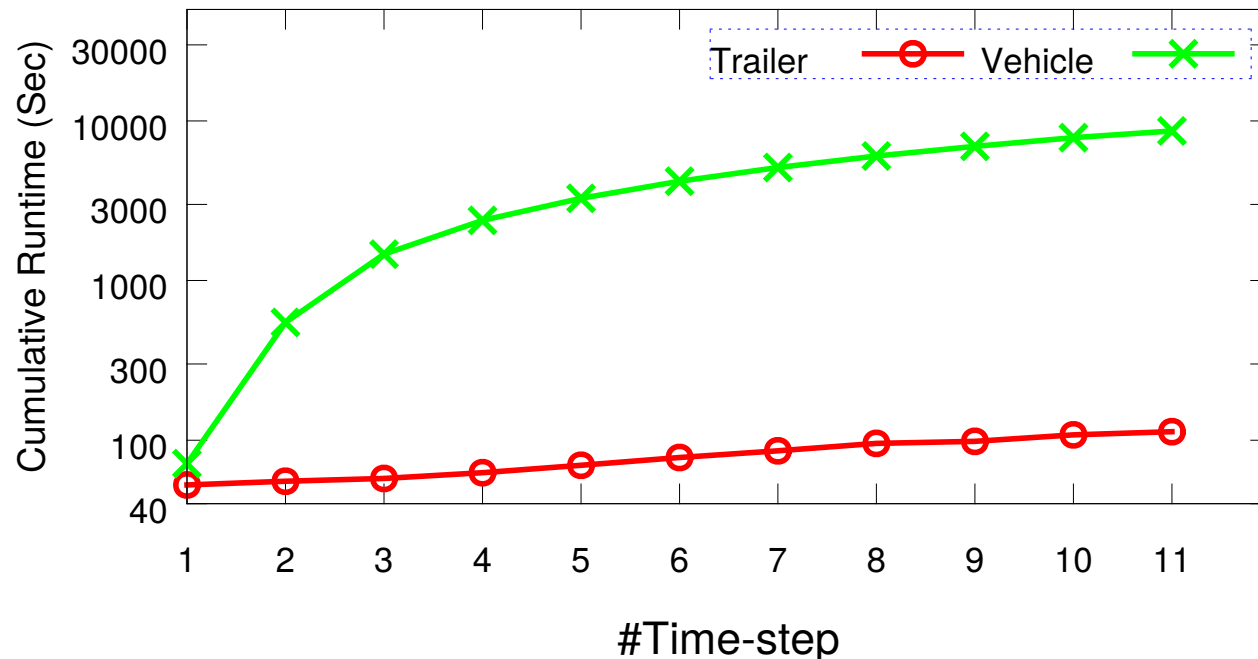


Experimental Results

■ Runtime Comparison:

- Runtime for the repositioning problem of trucks is around 15 minutes for each decision epoch.
- Tasks for the trailers can be generated within a minute

(Cumulative) Runtime Comparison



Summary

- **Dynamic Routing & Repositioning Problem using Bike-Trailers**
 - Self-sustaining and green mode of repositioning
 - We propose a unique combination of optimization and mechanism design to crowdsource the trailer tasks.
 - Experimental results show that trailers can replace the trucks.
- **Future Direction:**
 - Develop mechanism by considering the uncertainties in completion time of the trailer tasks.
 - Jointly consider the repositioning problem of trucks and trailers while considering the central budget constraints.

Q & A



supriyog.2013@phdis.smu.edu.sg

Thank you!



Supplementary Slides

Task Generation: DRRPT

$$\min_{\mathbf{y}} \sum_{s,k} L_s^k$$

Minimize total lost demand

$$\text{s.t. } L_s^k \geq \sum_{s'} F_{s,s'}^k - (d_s^{\#,t} + \sum_v (y_{s,v}^- - y_{s,v}^+)) \quad \forall k, s$$

Compute lost demand

$$y_{s,v}^+ \leq b_{s,v}^+ \cdot \min(d_s^{\#,t}, C_v^*), \quad \forall s, v$$

Pickup restrictions by a trailer from a station

$$\sum_v y_{s,v}^+ \leq d_s^{\#,t}, \quad \forall s$$

$$\sum_v y_{s,v}^- \leq C_s^{\#} - d_s^{\#,t}, \quad \forall s$$

Drop-off restrictions by a trailer at a station

$$y_{s,v}^- = b_{s,v}^- \cdot \sum_s y_{s,v}^+, \quad \forall s, v$$

$$(b_{s,v}^+ + b_{s',v}^- - 1) \cdot P_{s,s'} \leq P_{max} \quad \forall s, v$$

Physical limitations of trailer route

$$\sum_s b_{s,v}^+ = 1, \quad \forall v$$

$$\sum_{s \notin G_v} b_{s,v}^+ = 0, \quad \forall v$$

Trailer should pick up from one station and drop-off at other station

$$\sum_s b_{s,v}^- = 1, \quad \forall v$$

$$b_{s,v}^+, b_{s,v}^- \in \{0, 1\}, 0 \leq y_{s,v}^+, y_{s,v}^- \leq C_v^*, L_s^k \geq 0$$