

Asymptotic properties of bike-sharing systems

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SICSA workshop – Edinburgh, May 2016

1. j.w. with Christine Fricker (Inria), Vincent Jost (CNRS), Ariel Waserhole (ENSTA)

Question : What is your experience of bike-sharing systems ?

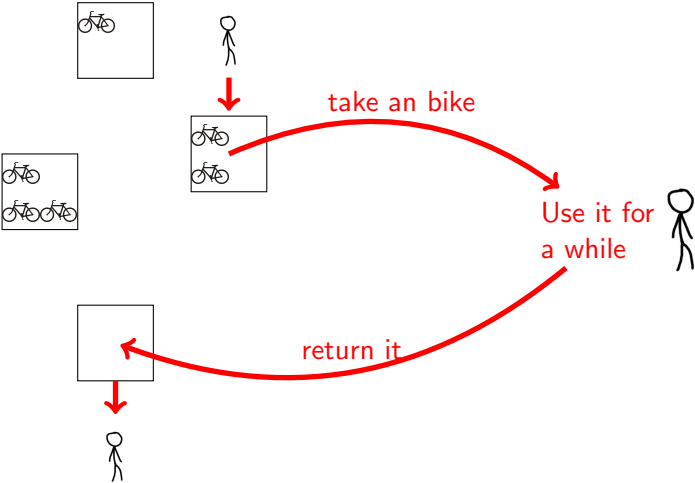
Question : What is your experience of bike-sharing systems ?

- ▶ Problems : lack of resources.

Bike-sharing systems



Bike-sharing systems



I will focus on large bike-sharing systems

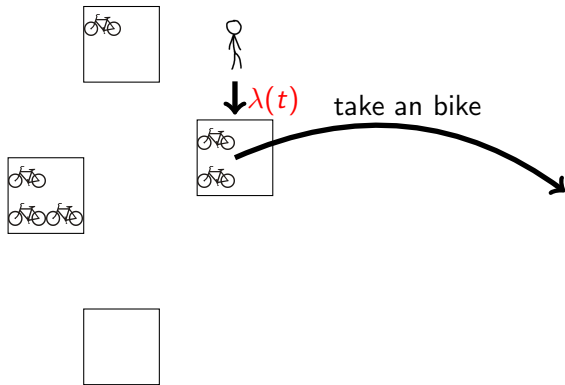


Map of Velib' stations in Paris (France).

Example of Velib' :

- ▶ 20 000 bikes
- ▶ 1 200 stations.

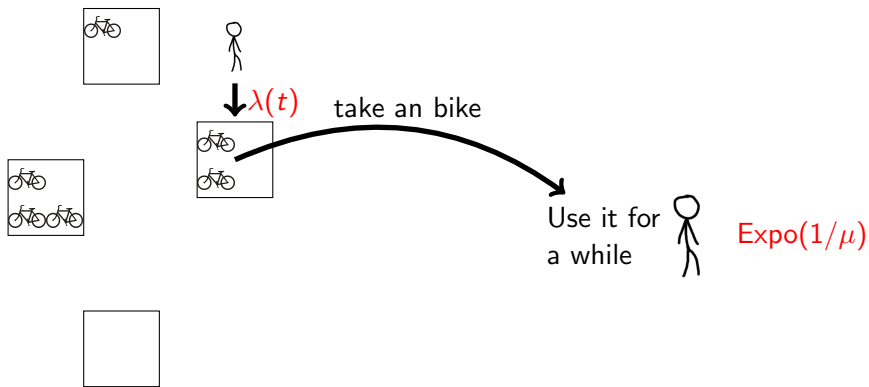
Goal : model the randomness of BSSs



Closed-queuing networks

Scaling : $N \rightarrow \infty$ stations, s bikes per station.

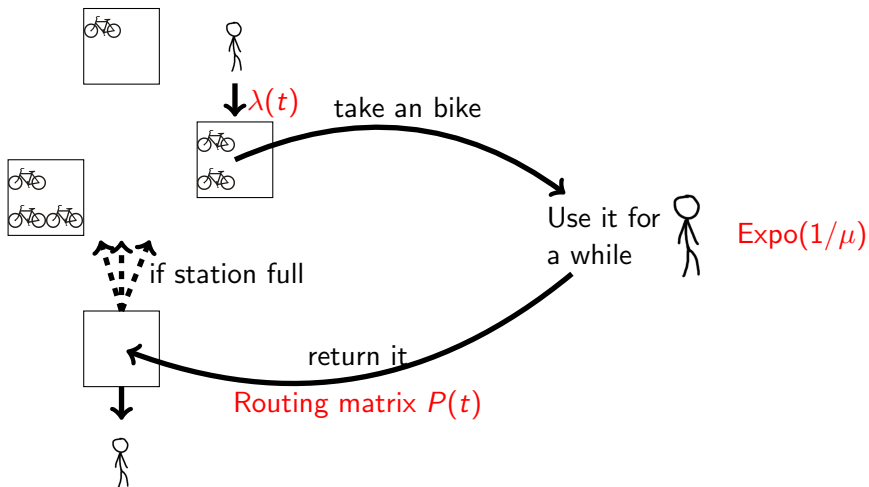
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Closed-queuing networks

Scaling : $N \rightarrow \infty$ stations, s bikes per station.

A few questions...

- ▶ Are there some typical regimes?
- ▶ What is the optimal fleet sizes?
- ▶ What should be the station capacity?
- ▶ What is the impact of redistribution or incentives?

Is the performance monotone?

Main message

Theoretical results : When the system is large :

- ▶ if the stations have finite capacities, the performance is continuous in the fleet size.
- ▶ if the stations have infinite capacities, there are problems of concentration.

Practical considerations :

- ▶ Performance is poor, even for a symmetric city (but simple incentives like a two-choice rule can help a lot).
- ▶ Frustrating users can help :
 - ▶ It is better to have stations of finite capacities.
 - ▶ Frustrating some users can improve the overall usage.
 - ▶ We show that the optimal fleet size is not

Outline

Detailed study of the homogeneous case

Adding some heterogeneity

Improvement by frustrating some demand

Conclusion and future work

The homogeneous model

- ▶ All stations are identical.

Motivation :

- ▶ Impact of random choices
- ▶ Close-form results
- ▶ “Best-case analysis”

“Theorem”

Asymptotically, stations are independent and behaves as a $M/M/1/K$.

Distribution of x_i , the fraction of station with i bikes

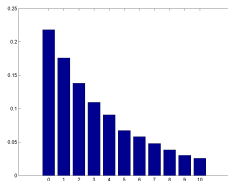
Theorem

There exists ρ , such that *in steady state*, as N goes to infinity :

$$x_i \propto \rho^i.$$

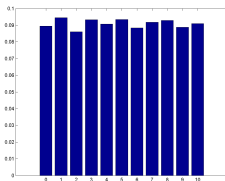
$\rho \leq 1$ iff $s \leq \frac{C}{2} + \frac{\lambda}{\mu}$ where s be the average number of bikes per stations.

$$s < \frac{C}{2} + \frac{\lambda}{\mu}$$



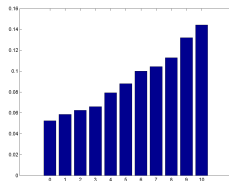
$$\rho < 1$$

$$s = \frac{C}{2} + \frac{\lambda}{\mu}$$



$$\rho = 1$$

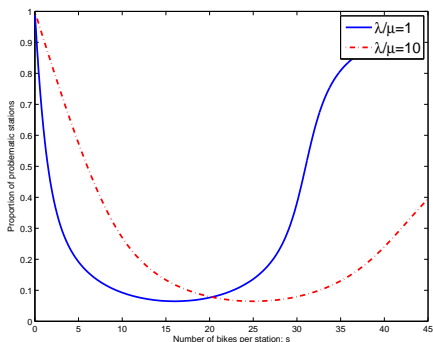
$$s > \frac{C}{2} + \frac{\lambda}{\mu}$$



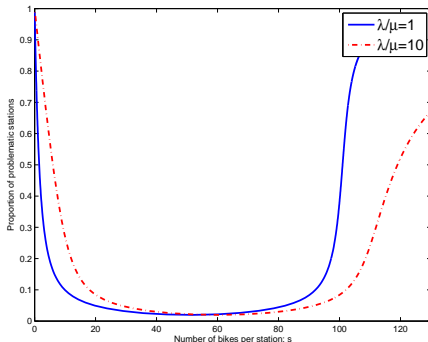
$$\rho < 1$$

Consequences : optimal performance for $s \approx C/2$

y-axis : Prop. of problematic stations. x-axis : number of bikes/station s .



(a) $C = 30$.



(b) $C = 100$.

Fraction of **problematic stations** (=empty+full) minimal for $s = \lambda/\mu + C/2$

- Prop. of problematic stations is at least $2/(C+1)$ (6.5% for $C = 30$)

Improvement by dynamic pricing : “two choices” rule

- ▶ Users can observe the occupation of stations.
- ▶ Users choose the **least loaded** among 2 stations close to destination to return the bike (ex : force by pricing)

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Paradigm known as “*the power of two choices*” :

- ▶ Comes from balls and bills [Azar et al. 94]
- ▶ Drastic improvement of service time in server farm [Vvedenskaya 96, Mitzenmacher 96]

Question : what is the effect on bike-sharing systems ?

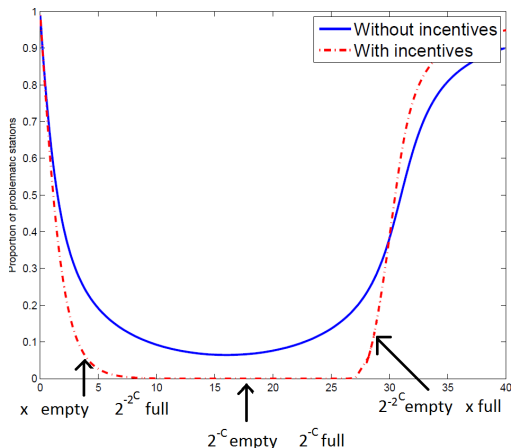
Characteristics :

1. **Finite capacity** of stations.
2. Strong **geometry** : choice among **neighbors**.

Two choices – finite capacity but no geometry

With no geometry, we can solve in close-form.

- ▶ Proof uses mean field argument.

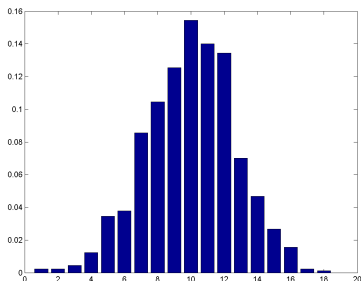


Choosing two stations at random, decreases problems from $2/C$ to $2^{-C/2}$

Two choices – taking geometry into account is hard

Mean field do not apply (geometry) :(.

- ▶ Existing results for balls and bins (see [Kenthapadi et al. 06])
- ▶ Only numerical results exists for server farms (ex : [Mitzenmacher 96])



We rely on simulation

Occupancy of stations

x-axis = occupation of station.

y-axis : proportion of stations.

Recall : with no incentives, the distribution would be uniform.

- ▶ Simulation indicate that 2D model is close to no-geometry
- ▶ Pair-approximation can be used but no close-form [Gast 2015]

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We assume that as N goes to infinity, the parameters (λ_i, p_i) of the station have a limiting distribution.

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“Theorem”

When the stations have finite capacities, a station behaves as a $M/M/1/K$.

Finite capacities regime

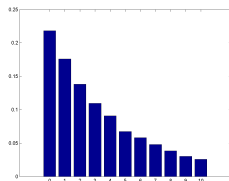
Theorem (Propagation of chaos-like result)

There exists a function $\rho(p)$ such that for all k , if stations $1, \dots, k$ have parameter p_1, \dots, p_k , then, as N goes to infinity :

$$P(\#\{\text{bikes in stations } j\} = i_j \text{ for } j = 1..k) \propto \prod_{j=1}^k \rho(p_j)^{i_j}$$

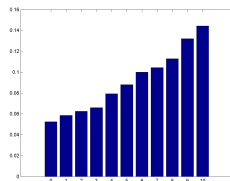
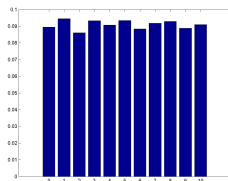
Depending on popularity, stations have different behaviors :

Popular start



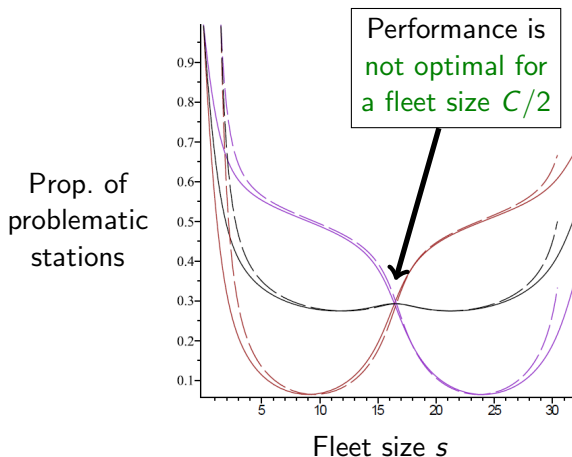
→

Popular destination



Finite-capacity : numerical example

Two types of stations : popular and non-popular for arrivals : $\lambda_1/\lambda_2 = 2$.



Infinite capacities can worsen the situation

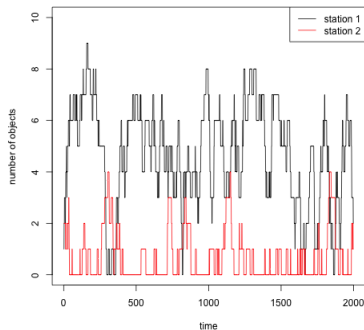
Infinite capacities can worsen the situation

Theorem (Malyshev-Yakovlev 96)

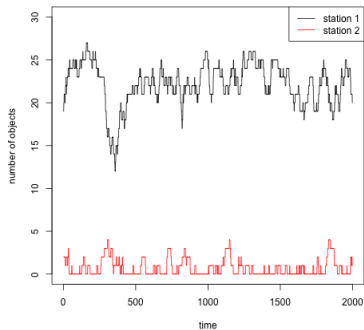
When the stations have infinite capacity, then there exists s_c :

- ▶ if $s < s_c$, a station behaves as a $M/M/1/K$.
- ▶ if $s > s_c$, bikes will accumulate in a few stations.

Example with $\mu = 1$, $p = (2, 1, 1, 1, 1, 1, 1, 1, 1)/10$:



$$s = 1 < s_c$$



$$s = 3 > s_c$$

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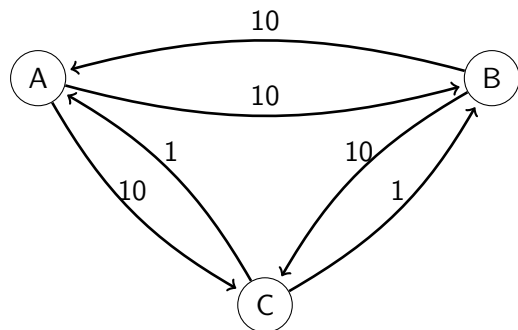
Conclusion and future work

Having finite capacities prevent saturation of the demand.
What if we could frustrate some demand ?

Model : we have a trip demand $\Lambda_{ij}(t)$ and an accepted demand $\lambda_{ij}(t)$.

- ▶ Generous policy : $\lambda_{ij}(t) := \Lambda_{ij}(t)$
- ▶ Possible control $\lambda_{ij}(t) \leq \Lambda_{ij}(t)$

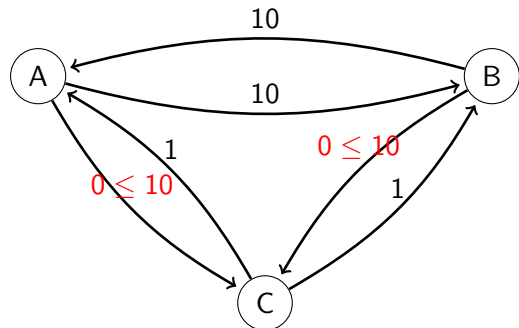
Frustrating demand can improve the balance of bikes



Users want to go to C.
Almost nobody wants
to go to A or B.

	Rate of trips (infinite capacities, infinite vehicles)
Generous policy	≈ 6 trips / time unit

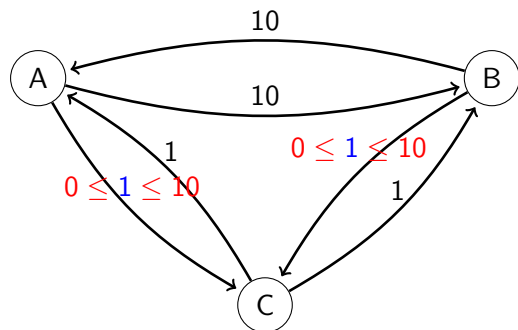
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	Rate of trips (infinite capacities, infinite vehicles)
Generous policy	≈ 6 trips / time unit
Frustrating policy	20 trips / time unit

Frustrating demand can improve the balance of bikes

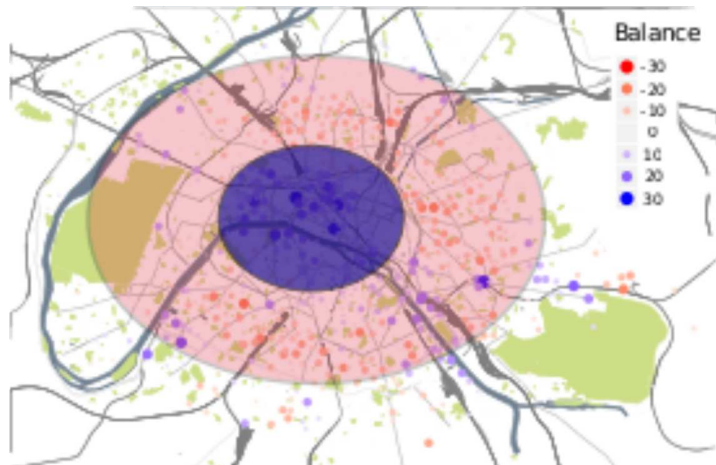


Users want to go to C.
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	Rate of trips (infinite capacities, infinite vehicles)
Generous policy	≈ 6 trips / time unit
Frustrating policy	20 trips / time unit
Optimal circulation	24 trips / time unit

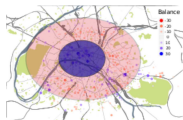
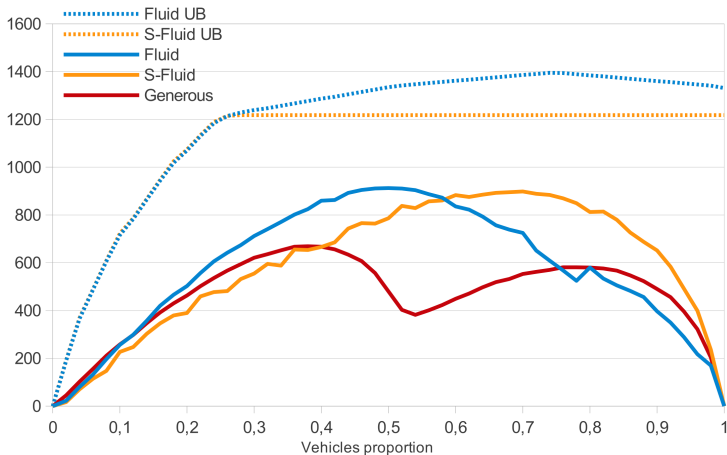
We can explore dynamic scenarios [Waserhole/Jost 2012]

Tides in Paris



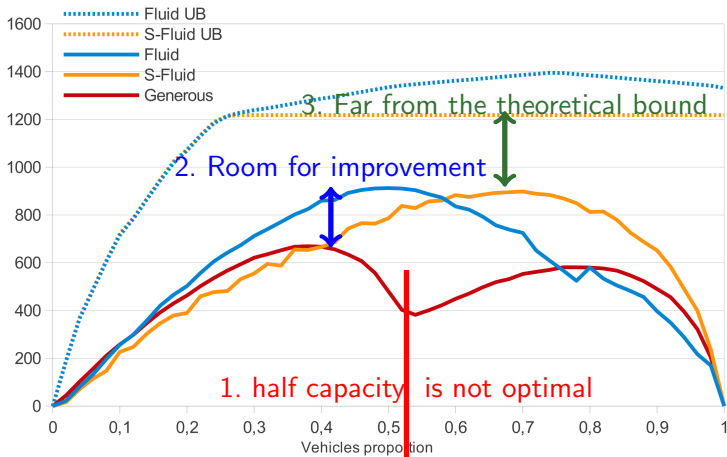
Simulation results : Static time-varying frustration of user can improve the situation

Trips per Second



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Trips per Second



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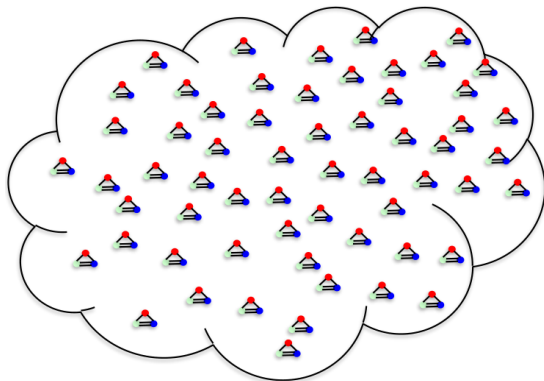
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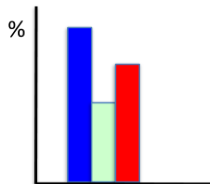
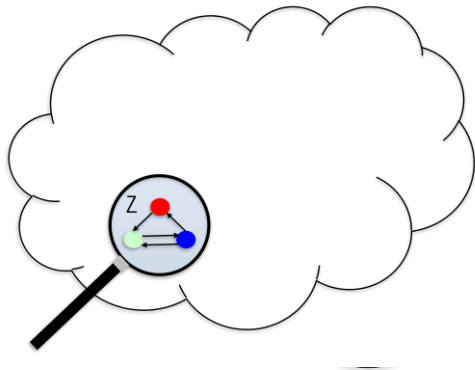
Conclusion and future work

Methodological comments : the asymptotic method comes from statistical mechanics (mean-field approximation)



- ▶ Basic models are reversible.
 - ▶ Saddle-points methods can also be used.

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- ▶ Basic models are reversible.
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Summary

Asymptotic results for a large class of bike-sharing network.

- ▶ **Performance poor**, even for symmetric : $1/C$ problematic stations.
- ▶ Simple incentives can help a lot : 2^{-C} .
- ▶ Frustrating some users improves overall usage.

Possible extensions of this model

- ▶ **Optimal regulation rate** : λ/C .
- ▶ Reservation : increases congestion.

Discussion

- ▶ Metrics are not easy to define.
- ▶ Visualization of traces and Influence of geometry?

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- ▶ Metrics are not easy to define.
- ▶ Visualization of traces and Influence of geometry ?

If an ideal symmetric system works poorly, do not expect perfect service in a real system ;)

References

- ▶ C Fricker, N Gast. *Incentives and redistribution in homogeneous bike-sharing systems with stations of finite capacity*. EURO Journal on Transportation and Logistics. 2014.
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- ▶ Gast, N. and Massonnet, G. and Reijsbergen, D. and Tribastone, M. *Probabilistic forecasts of bike-sharing systems for journey planning*. ACM CIKM 2015