# Jérôme, Algorithmic Game Theory and Strategyproof Facility Location

### **Dimitris Fotakis**



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Based on joint work with Laurent Gourvès (LAMSADE), Jérôme Monnot (LAMSADE) and Panagiotis Patsilinakos (NTUA)

Homage to Jérôme Monnot, Univ. Paris-Dauphine, December 2021

Spanning Trees Constructed by Selfish Agents [Gourvés M., WINE 08]

- Agents pay for **first edge** of path to root.
- Pure Nash equilibrium iff strong PNE.
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Strong Equilibria of (k-)Cut Games [Gourvés, M., WINE 09, TAMC 10]

- Cut games admit strong PNE and SPoA = 2/3.
- *k*-Cut games admit 3-strong PNE.

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Congestion Games with Capacited Resources [Gourvés, M., Morreti, Thang, SAGT 12, ToCS 2015]

- PNE exists, m = 2 resources, NP-hard to tell for  $m \ge 3$ .
- PNE for singleton games, by elegant 2-dimensional potential!

#### Selfish Transportation [F. Gourvés M., SOFSEM 17]

- **Bus route** selection, based on flow time from entering bus until destination.
- PNE exist (and computed efficiently), if buses have "identical routes", or *m* = 2 and metric distances, or distances ∈ {1,2}.
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### And Much More ...

- Coordination mechanisms for scheduling selfish tasks with setup times [Gourvés, M., Telelis, WINE 09]
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- **Project** games: how competitive agents select tasks and when decide to cooperate [Biló, Gourvés, M., CIAC 19]
- Fair allocation of indivisible goods: approximation of the **maximin share** for *n* = 3 and matroids [Gourvés, M., CIAC 17]

## Our Focus: k-Facility Location Games

Public Good Allocation for Strategic Agents on the Line

- Agents  $N = \{1, ..., n\}$  on the real line.
- Agent *i* wants a facility close to *x<sub>i</sub>*, which is private information.



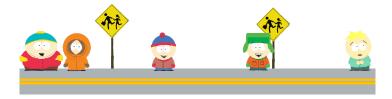
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#### (Randomized) Mechanism

**Mechanism** *F* maps reported ideal locations  $y = (y_1, ..., y_n)$  to (probability distribution over) set(s) of *k* facilities.



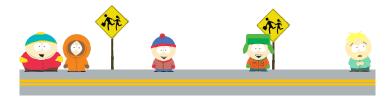
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- Each agent *i* **reports** *y*<sub>*i*</sub> that may be **different** from *x*<sub>*i*</sub>.

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## Preferences and Truthfulness

### **Connection Cost**

(Expected) distance of agent *i*'s **true location** to the **nearest** facility:  $cost[x_i, F(y)] = dist(x_i, F(y)) = min_{c \in F(y)} |x_i - c|$ 



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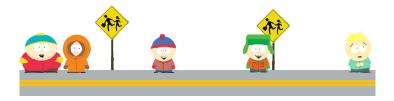
For any location profile x, agent i, and location y:  $cost[x_i, F(x)] \le cost[x_i, F(y, x_{-i})]$ 



## Variants and Social Efficiency

#### Candidate Facility Locations:

- Unrestricted : Any point (esp. agent locations) can be facility.
- **Restricted** : Facilities selected from *m* candidate locations *C* Motivation from multi-winner elections: Chamberlin-Courant.



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### Social Objective

F(x) should optimize (or approximate) a given **objective function**.

- Social Cost: minimize  $\sum_{i=1}^{n} \operatorname{cost}[x_i, F(\mathbf{x})]$
- Social Welfare : maximize  $\sum_{i=1}^{n} (L \text{cost}[x_i, F(x)])$



## 1-Facility Location on the Line

#### Median Mechanism

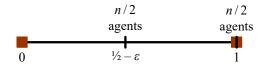
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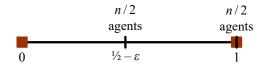
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  - Median social  $\cos t \approx 3n/4$ . Median social welfare  $\approx n/4$ .
- Anonymity and truthfulness iff generalized median [Moulin 80]



## *k*-Facility Location on the Line, $k \ge 2$

Percentile Mechanisms [Sui Boutilier Sandholm, IJCAI 13]

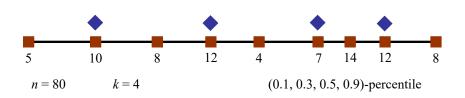
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- $vote(\ell) = #agents preferring \ \ell \in C$  to other candidates in C.
- *j*-th facility at leftmost ℓ ∈ C with ≥ α<sub>j</sub> fraction of vote on ℓ and its left.
  - Median is 0.5-percentile. Two-Extremes is (0,1)-percentile.

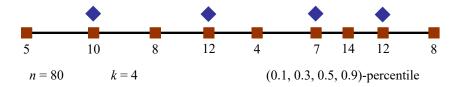


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- Percentile mechanisms are anonymous and truthful.



(*k*, *q*)-Conference Program Design [Caragiannis Gourvés M., IJCAI 16], [F. Gourvés M., WINE 16]

**Robust** version of (*kq*)-Facility Location where each agent receives **utility from** *q* **different** facilities.

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- For any  $q \ge 1$ ,  $(1/(2k), 3/(2k), \dots, (2k-1)/(2k))$ -percentile applied to q-tuples is **truthful** and (2k-3)/(2k-1)-**approximate**.

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- Optimal solution through generalization of LP-based approach in [Hajiaghayi et al., SODA 14]

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- Focus on instances with stable optimal clustering.

## Perturbation Stability for k-Facility Location

### Perturbation Stability in Clustering [Bilu Linial, ITCS 10]

γ-stability: scaling down any distances by factor ≤ γ (while maintaining metric property) does not affect optimal solution.



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- Real-world instances are **stable**: "Clustering is hard when it doesn't *matter*" [Roughgarden 17]



### Question

Assume that "true" instances are indeed stable.

How much stability for truthfulness and reasonable approximation?

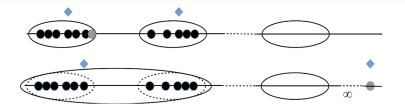
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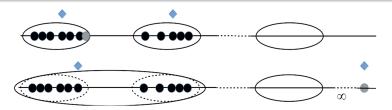
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- For k ≥ 3, deterministic anonymous truthful mechanisms for (√2 − ε)-stable instances have unbounded approximation (based on [F. Tzamos, ICALP 13])



- **Optimal** clustering  $(C_1, \ldots, C_k)$  due to bounded approximation.
- Stability verification (necessary cond.): allocate facilities only if max{diam(C<sub>i</sub>), diam(C<sub>i+1</sub>)} < d(C<sub>i</sub>, C<sub>i+1</sub>)

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- For 5-stable instances, facility at random agent in each optimal cluster is truthful and 2-approximate.

## Optimal Mechanism for Stable k-Facility Location

### Optimal Mechanism and Approach to Truthfulness

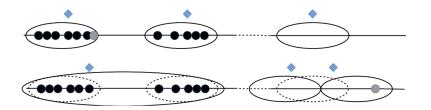
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- Key deviation: rightmost agent of *C<sub>i</sub>* deviates to *C<sub>j</sub>*, causing *C<sub>j</sub>* to split and *C<sub>i</sub>* to merge with *C<sub>i+1</sub>*.
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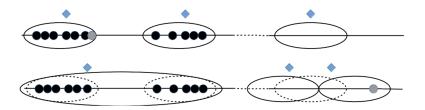


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- "Simulate" increase in cost of C<sub>j</sub> by γ-perturbation and decrease in cost of C<sub>j</sub> by agent's cost improvement.
- Stability: optimal clustering not affected by deviation.



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- Extension to trees and study of general metrics.
- Possibility of truthfulness **for all** instances and bounded approximation only **for stable**? (conjecture: **no**)
- Complexity of **determining** whether a *k*-Facility Location instance is *γ*-stable, esp. for line and trees?

## Thank You for Everything and Goodbye

We, your many friends, deeply miss your kindness, openness, collaboration, passion for research and life, warm smile and true love for people, and so many things we kept learning from you.



# Thank you and goodbye, Jérôme