Algorithmic Game Theory and Applications

Congestion Games

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Braess' Paradox (Pigou 1920)

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<u>Reasonable Assumption</u>: The cost c_r of a resource is *non-decreasing* in the number of players that use it (*monotonicity*).

But it is not unreasonable to not have this in some cases, e.g., the *El Farol Bar problem*.



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For example: $c_e(x)$ could be a linear function $c_e(x) = \alpha_e x + \beta_e$
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U	-1, 1	1, -1	-2, -2
M	1, -1	-1, 1	-2, -2
D	-2, -2	-2, -2	2, 2

<u>Recall</u>: Best response := a strategy s_i that maximises the utility of Player igiven the strategies s_{-i} of the other players.

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The theorem also gives us an *algorithm* to find a PNE:

- start from any arbitrary strategy profile,
- run the best response dynamics until we reach a PNE.

Potential Games

<u>Definition</u>: A game is an (exact) potential game if there exists a potential function $\Phi : S_1 \times ... \times S_n \to \mathbb{R}$ such that for all $i \in N$, all $s_{-i} \in S_{-i}$ and $s_i, s'_i \in S_i$, we have that

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In particular, this also holds for $s' = (s'_i, s_{-i})$. Since the game is a potential game, this means that $\text{cost}_i(s^*) \ge \text{cost}_i(s')$, and hence s^* is a pure Nash equilibrium.

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- 1. N is a set of n players.
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3. $S = S_1 \times \ldots \times S_n$, where $S_i \subseteq 2^R \setminus \{0\}$ is the set of (pure) strategies of Player *i*.

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Best Response Dynamics in Congestion Games

Theorem (Rosenthal 1973): In any congestion game, the best response dynamics always converges to a pure Nash equilibrium.

In particular, this implies that every congestion game has a pure Nash equilibrium.

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How do we prove that the algorithm will terminate?

Whenever a player best responds, the player's utility is increased by $u_i(s'_i, s_{-i}) - u_i(s_i, s_{-i}) = \alpha$.

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We would have to show two things:

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Unary representation: $5_{10} \rightarrow 11111$

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- The cost functions for each agent can be represented in space $O(m \cdot n \cdot \log \max r_j(n))$, where we represent the function $r_j(\cdot)$ using a binary representation.

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Intuition: If the cost functions are represented with fairly small numbers, then it is a fast algorithm.

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<u>Theorem (Babichenko and Rubinstein 2021)</u>: Computing a MNE of a congestion game is PPAD \cap PLS - complete.

