

# Congestion Games

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- ▶ We'll study a simple, natural dynamic, and show it converges to Nash equilibrium.
- ▶ Our first “computationally plausible” set of predictions in a large interaction.

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**A:** We need  $k^n$  numbers just to encode a single utility function.

Unreasonable to expect anyone to understand such an object.  
So: we need to think about structured, concisely defined games.



# Example 1: Traffic Routing

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3. For each player  $i$ , a set of actions  $A_i$ . Each action  $a_i \in A_i$  represents a subset of the facilities:  $a_i \subseteq F$ .

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4. For each facility  $j \in F$ , a cost function  $\ell_j : \{0, \dots, n\} \rightarrow \mathbb{R}_{\geq 0}$ .  $\ell_j(k)$  represents “the cost of facility  $j$  when  $k$  players are using it”.

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Player costs are then defined as follows. For action profile  $a = (a_1, \dots, a_n)$  define  $n_j(a) = |\{i : j \in a_i\}|$  to be the number of players using facility  $j$ . Then the cost of agent  $i$  is:

$$c_i(a) = \sum_{j \in a_i} \ell_j(n_j(a))$$

## Example 2: Network Creation



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# Congestion games

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- ▶ Do they have pure strategy Nash equilibria?
- ▶ Can computationally bounded, uncoordinated players find one?
- ▶ i.e. are pure strategy Nash equilibria computationally plausible predictions?
- ▶ Lets study a simple dynamic...

# Best (Better) Response Dynamics

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2. In arbitrary order, players take turns changing their action if doing so can improve their utility.
3. Forever...

# Best (Better) Response Dynamics

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## Algorithm 1 Best Response Dynamics

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**Initialize**  $a = (a_1, \dots, a_n)$  to be an arbitrary action profile.

**while** There exists  $i$  such that  $a_i \notin \arg \min_{a \in A_i} c_i(a, a_{-i})$  **do**

**Let**  $a'_i$  be such that  $c_i(a'_i, a_{-i}) < c(a)$ .

**Set**  $a_i = a'_i$ .

**end while**

**Halt** and return  $a$ .

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## Algorithm 2 Best Response Dynamics

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## Algorithm 3 Best Response Dynamics

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## Claim

*If best response dynamics halts, it returns a pure strategy Nash equilibrium.*

## Proof.

Immediate from halting condition – by definition, every player must be playing a best response. □

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*Best response dynamics always halt in congestion games.*

## Corollary

*All congestion games have at least one pure strategy Nash equilibrium.*

# Analysis of BRD in Congestion Games

1. Consider the *potential function*  $\phi : A \rightarrow \mathbb{R}$ :

$$\phi(\mathbf{a}) = \sum_{j=1}^m \sum_{k=1}^{n_j(\mathbf{a})} \ell_j(k)$$

(Note: *not* social welfare)

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3. Well... We know it must have decreased player  $i$ 's cost:

$$\begin{aligned} \Delta c_i &\equiv c_i(b_i, a_{-i}) - c_i(a_i, a_{-i}) \\ &= \sum_{j \in b_i \setminus a_i} \ell_j(n_j(a) + 1) - \sum_{j \in a_i \setminus b_i} \ell_j(n_j(s)) \\ &< 0 \end{aligned}$$

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4. And hence BRD halts in congestion games...

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2. Therefore, the change in potential is strictly *negative*
3. So... since  $\phi$  can take on only finitely many values, this cannot go on forever.
4. And hence BRD halts in congestion games...
5. Which proves the *existence* of pure strategy Nash equilibria!

# Efficiency

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Our proof gives only an exponential convergence bound... And it might really take that long!

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might really take that long!  
Lets consider approximation...

# Approximation

## Definition

An action profile  $a \in A$  is an  $\epsilon$ -approximate pure strategy Nash equilibrium if for every player  $i$ , and for every action  $a'_i \in A_i$ :

$$c_i(a_i, a_{-i}) \leq c_i(a'_i, a_{-i}) + \epsilon$$

i.e. nobody can gain more than  $\epsilon$  by deviating.

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# Approximate Best Response Dynamics

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**Algorithm 4** FindApproxNash( $\epsilon$ )

---

**Initialize**  $a = (a_1, \dots, a_n)$  to be an arbitrary action profile.

**while** There exists  $i, a'_i$  such that  $c_i(a'_i, a_{-i}) \leq c_i(a_i, a_{-i}) - \epsilon$  **do**

**Set**  $a_i = \arg \min_{a \in A_i} c_i(a, a_{-i})$

**end while**

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**Algorithm 5** FindApproxNash( $\epsilon$ )

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**Set**  $a_i = \arg \min_{a \in A_i} c_i(a, a_{-i})$

**end while**

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## Claim

*If FindApproxNash( $\epsilon$ ) halts, it returns an  $\epsilon$ -approximate pure strategy Nash equilibrium*



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**Algorithm 6** FindApproxNash( $\epsilon$ )

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**while** There exists  $i, a'_i$  such that  $c_i(a'_i, a_{-i}) \leq c_i(a_i, a_{-i}) - \epsilon$  **do**

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## Claim

*If FindApproxNash( $\epsilon$ ) halts, it returns an  $\epsilon$ -approximate pure strategy Nash equilibrium*

## Proof.

Immediately, by definition. □

# Analysis

## Theorem

*In any congestion game, FindApproxNash( $\epsilon$ ) halts after at most:*

$$\frac{n \cdot m \cdot c_{\max}}{\epsilon}$$

*steps, where  $c_{\max} = \max_{j,k} \ell_j(k)$  is the maximum facility cost.*

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We revisit the potential function  $\phi$ . Recall that  $\Delta c_i = \Delta \phi$  on any round when player  $i$  moves.

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Observe also that at every round,  $\phi \geq 0$ , and

$$\phi(a) = \sum_{j=1}^m \sum_{k=1}^{n_j(a)} \ell_j(k) \leq n \cdot m \cdot c_{max}$$

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By definition of the algorithm, we have  $\Delta c_i = \Delta \phi \leq -\epsilon$  at every round, and so the theorem follows.

Thanks!

See you next class!