

# When do Best Response Dynamics Converge?

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- ▶ Is that it? How much further can we push it?
- ▶ Today: study a couple more games in which BRD converges, and try to abstract what is needed.
- ▶ Characterize *exactly* when BRD is guaranteed to converge.

# Load Balancing Games on Identical Machines

## Definition

A *load balancing game on identical machines* models  $n$  players  $i \in P$  scheduling jobs of size  $w_i > 0$  on  $m$  identical machines  $F$ .

The game has:

1. Action space  $A_i = F$  for each player
2. For each machine  $j \in F$ , a load  $\ell_j(a) = \sum_{i:a_i=j} w_i$

The cost of player  $i$  is the load of the machine he plays on:

$$c_i(a) = \ell_{a_i}(a).$$

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The cost of player  $i$  is the load of the machine he plays on:

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*Almost* a congestion game — but facility costs depend on *which* players are using them.

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## Theorem

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## Corollary

*Load balancing games on identical machines have pure strategy Nash equilibria*

## Proof

Define  $\phi(a) = \frac{1}{2} \sum_{j=1}^m \ell_j(a)^2$ . Suppose player  $i$  switches from machine  $j$  to machine  $j'$ . Then we have:

$$\Delta c_i(a) \equiv c_i(j', a_{-i}) - c_i(j, a_{-i})$$

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Note:  $\Delta c_i \neq \Delta \phi$ .

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## Definition

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1. The players are vertices  $P = V$ .
2. Each edge  $e = (i, j) \in E$  has weight  $w_e$
3. Actions  $A_i = \{-1, 1\}$  (read {red, blue})
4.  $u_i(a) = \sum_{e=(i,j) \in E} w_e \cdot a_i \cdot a_j = \sum_{j:a_i=a_j} w_{i,j} - \sum_{j:a_i \neq a_j} w_{i,j}$

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“Everyone picks an affiliation, and obtains utility equal to the weight of friends who pick the same affiliation, and disutility equal to the weight of friends who don't.”

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*The Red-State/Blue-State game always has a pure strategy Nash equilibrium.*

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## Definition

A function  $\phi : A \rightarrow \mathbb{R}_{\geq 0}$  is an *exact potential function* for a game  $G$  if for all  $a \in A$ , all  $i$ , and all  $a_i, b_i \in A_i$ :

$$\phi(b_i, a_{-i}) - \phi(a_i, a_{-i}) = c_i(b_i, a_{-i}) - c_i(a_i, a_{-i})$$

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## Definition

$\phi : A \rightarrow \mathbb{R}_{\geq 0}$  is an *ordinal potential function* for a game  $G$  if for all  $a \in A$ , all  $i$ , and all  $a_i, b_i \in A_i$ :

$$(c_i(b_i, a_{-i}) - c_i(a_i, a_{-i}) < 0) \Rightarrow (\phi(b_i, a_{-i}) - \phi(a_i, a_{-i}) < 0)$$

i.e. the change in utility is always equal *in sign* to the change in potential.

# A Characterization

## Theorem

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*So our proof technique is without loss of generality!*

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2. Consider a graph  $G = (V, E)$ :
  - 2.1 Let each  $a \in A$  be a vertex in the graph: i.e.  $V = A$ .
  - 2.2 For each pair of vertices  $a, b \in V$ , add a directed edge  $(a, b)$  if it is possible to get to  $a$  from  $b$  via a best response move – i.e. if there is some index  $i$  such that  $b = (b_i, a_{-i})$ , and  $c_i(b_i, a_{-i}) < c_i(a)$ .

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4. Nash Equilibria are the sinks in this graph.
5. BRD converges = there are no cycles in this graph.

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4. We need: for any edge  $a \rightarrow b$ ,  $\phi(b) < \phi(a)$ .
5. Its true!  $\phi(a) \geq \phi(b) + 1$ . (why?)



# Thanks!

See you next class — stay healthy!