

# A Game Theoretic Approach to Network Coding

Jason R. Marden  
California Institute of Technology

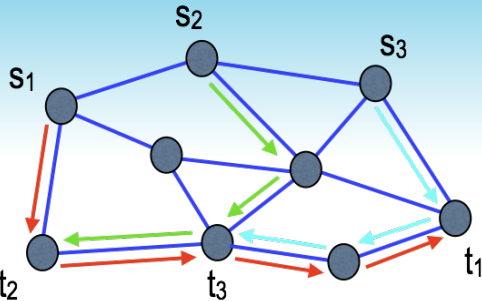
Michelle Effros  
California Institute of Technology

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# A Game Theoretic Approach to Network Coding

Marden and Effros

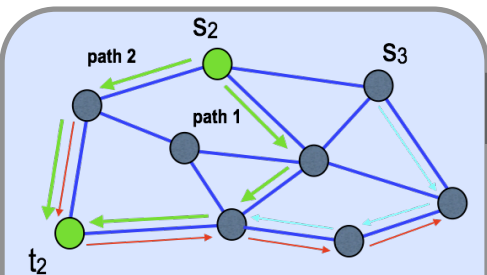
**STATUS QUO**



**Global Objective:** Efficiently use network using network coding

**Approach:** Centralized solutions. (e.g., opportunistic coding) Fix paths, use coding opportunities if available

**NEW INSIGHTS**



What about **distributed** solutions?  
What if flows were allowed to select path in response to local "cost"?

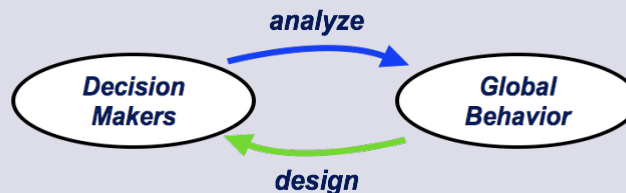
**Goal:** Let users create coding opportunities to improve efficiency

## ACHIEVEMENT DESCRIPTION

### MAIN ACHIEVEMENT:

Introduced game theory as a distributed tractable mechanism to obtain good network performance

*game theory for social sciences:*  
"descriptive agenda"



*game theory for engineering:*  
"prescriptive agenda"

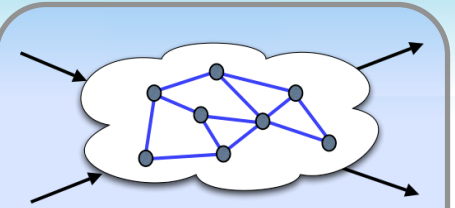
### HOW IT WORKS:

- Model interactions as a non-cooperative game
  - players (unicast flows)
  - actions (available paths)
- Assign each player a "cost" function
- Analyze efficiency of equilibrium behavior

### ASSUMPTIONS AND LIMITATIONS:

- Limited form of network coding (reverse carpool)
- Players have knowledge of available paths
- Players equilibrate faster than network changes

**IMPACT**



Approach provides guarantees **independent** of network structure.

Guarantees existence of an equilibrium that achieves a system cost of at most 50% higher than the optimal.

This offers an improvement over opportunistic coding.

**NEXT-PHASE GOALS**

*Understand the potential of game theory in network coding problems*

*Establish **desirable** distributed learning algorithms with good convergence rates*

*Extend game theoretic approach to more general network coding problems*

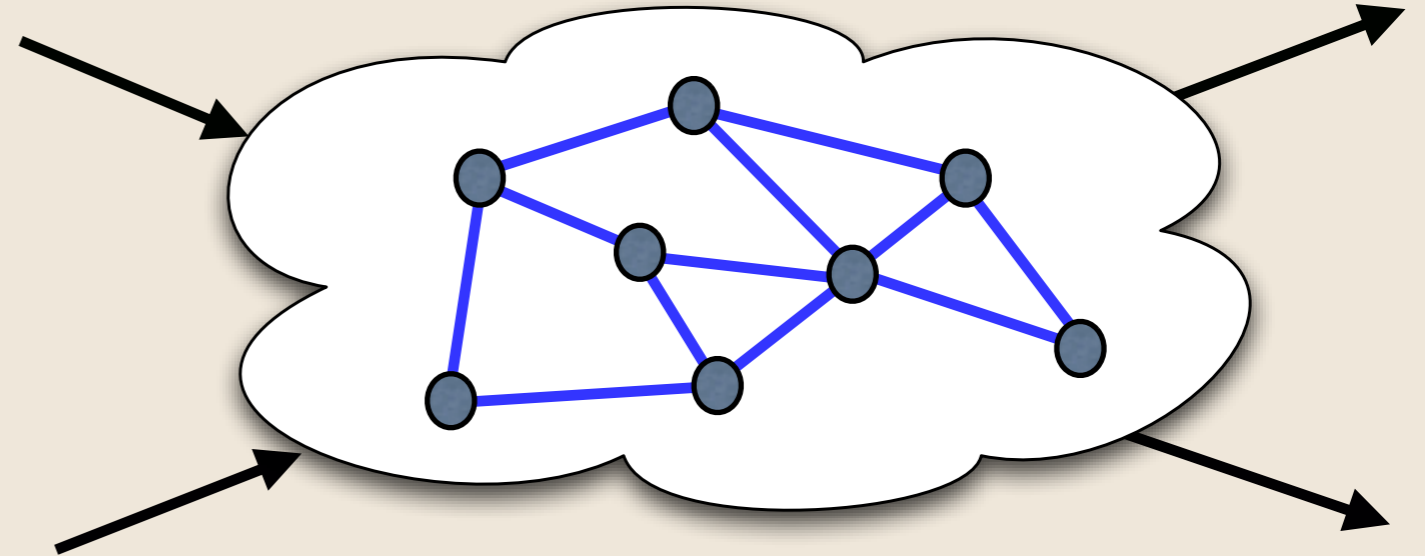
**Game theory is an applicable tool for distributed optimization in network coding**

## Example: Network coding

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### Features:

- large common network
- large # of users
- different network demands



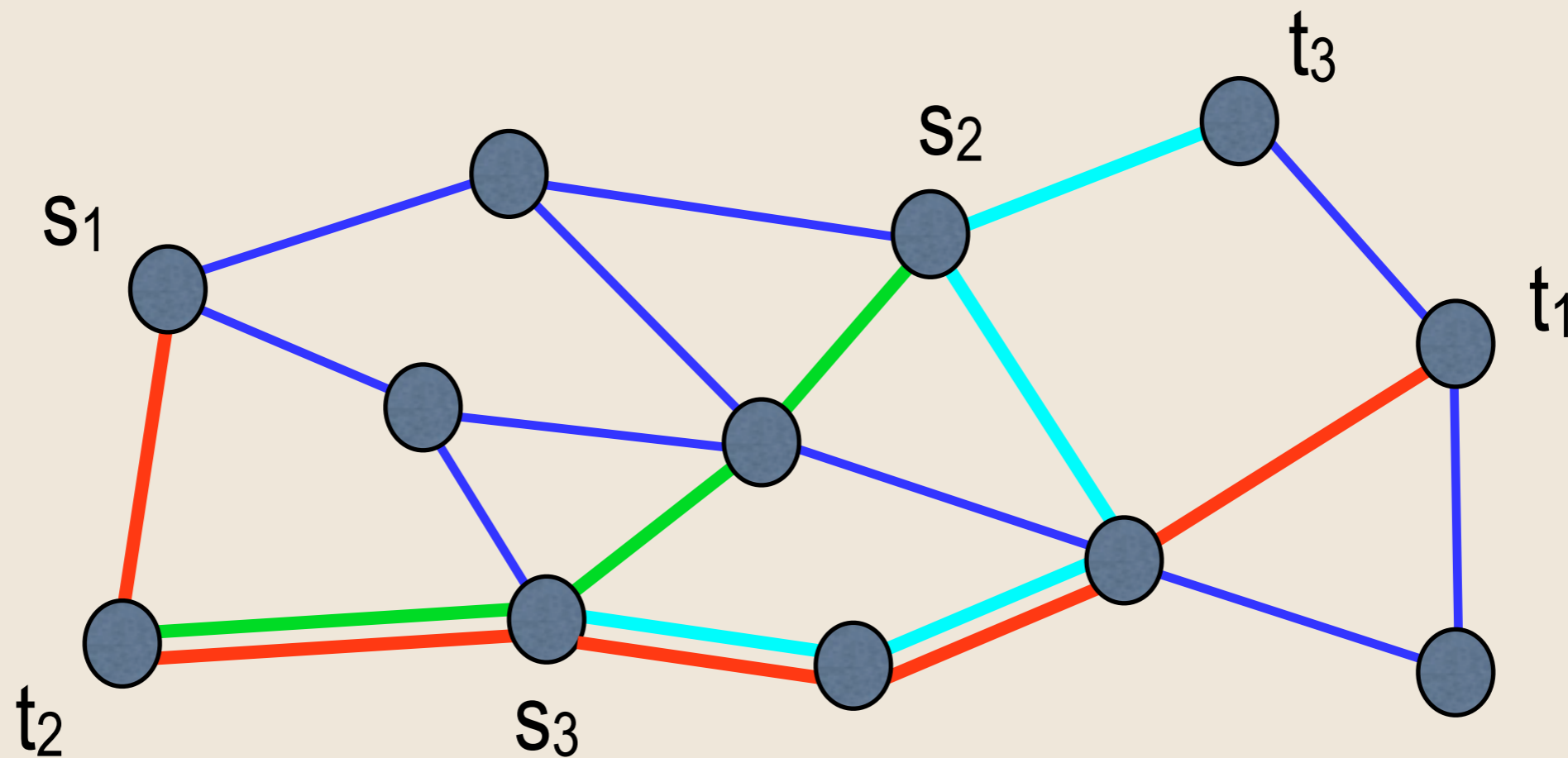
**Global objective:** Allocate users efficiently over network (utilizing network coding)

- maximize throughput
- minimize # of transmissions

### Challenges:

- centralized optimization is not feasible
- network coding capacity is unsolved

**Multiple unicast flows in shared network environment**

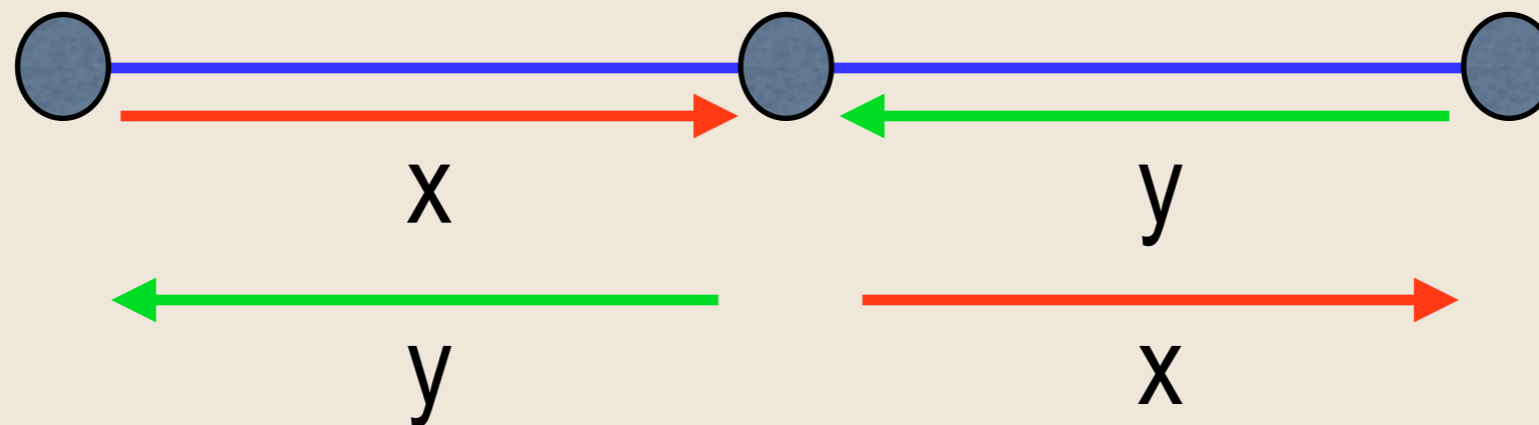


*possible transmissions highlighted by edges on graph*

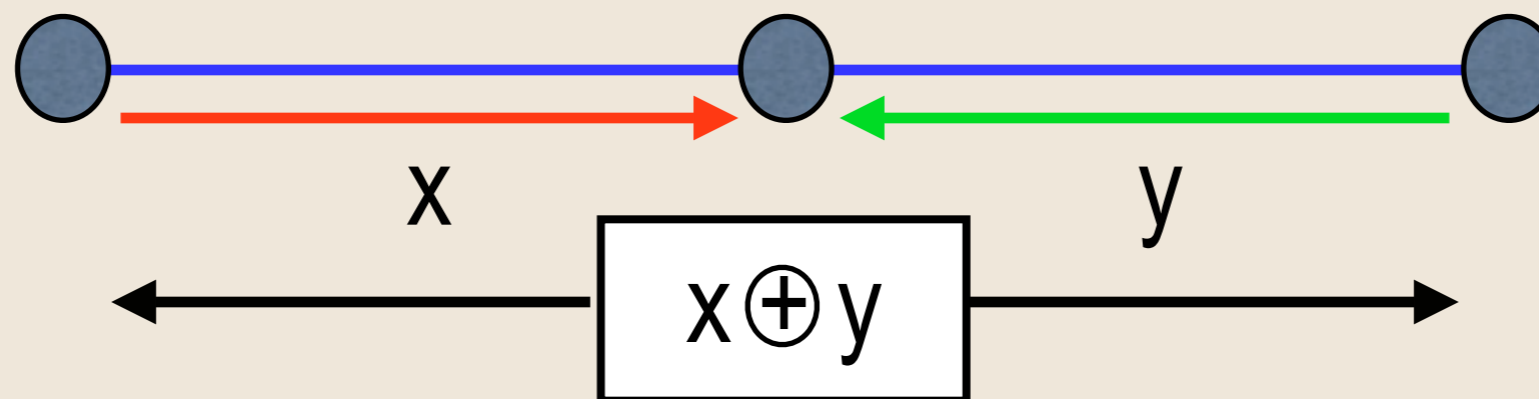
**Cost of allocation = number of transmission**

## Limited form of network coding: “Reverse carpooling”

*opportunity for network coding arises when two unicasts traverse the same node in opposite directions*



*without network coding - 4 transmissions required*



*with network coding - 3 transmissions required*

**Approach:** Model interactions as a non-cooperative game

- Players (unicast flows):  $\{1, 2, \dots, n\}$   $(s_i, t_i)$
- Actions (available paths):  $a_i \in \mathcal{A}_i$   
 $\mathcal{A} = \mathcal{A}_1 \times \dots \times \mathcal{A}_n$
- System cost:  $C(a) = \sum_e \max\{\#_e^0(a), \#_e^1(a)\}$

**Goal:** Design local players' cost functions so that equilibrium behavior is desirable

**Cost functions:**

$$J_i(a) = J_i(a_i, a_{-i})$$

**Equilibrium behavior:** Pure Nash equilibrium

$$J_i(a_i^*, a_{-i}^*) \leq J_i(a_i, a_{-i}^*)$$

Let

$\mathcal{E}(G) := \{a \in \mathcal{A} : a \text{ is a Nash equilibrium of game } G\}$

$a^{\text{opt}} \in \arg \min_{a \in \mathcal{A}} C(a)$

## Price of Anarchy

$$POA = \sup_G \max_{a \in \mathcal{E}(G)} \frac{C(a)}{C(a^{\text{opt}})}$$

worst case performance of any NE

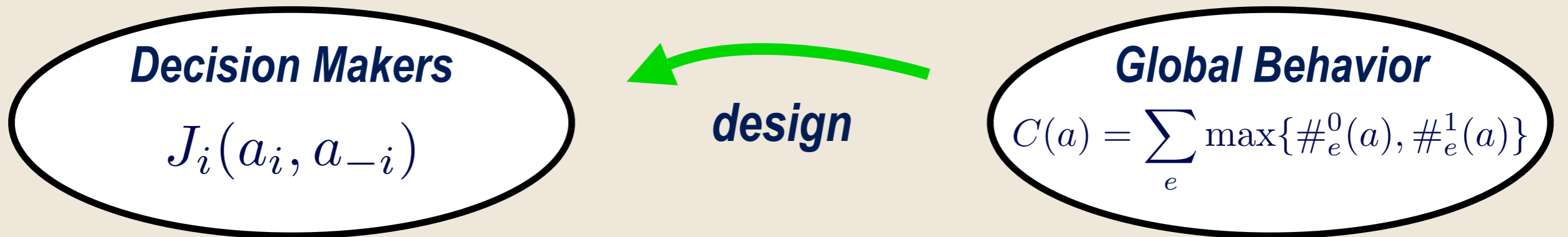
## Price of Stability

$$POS = \sup_G \min_{a \in \mathcal{E}(G)} \frac{C(a)}{C(a^{\text{opt}})}$$

worst case performance of best NE

*(independent of network structure or demands)*

# Cost design: Wonderful life design



**Wonderful Life:**  $J_i(a) = C(a) - C(a_i^0, a_{-i})$

**Positives**

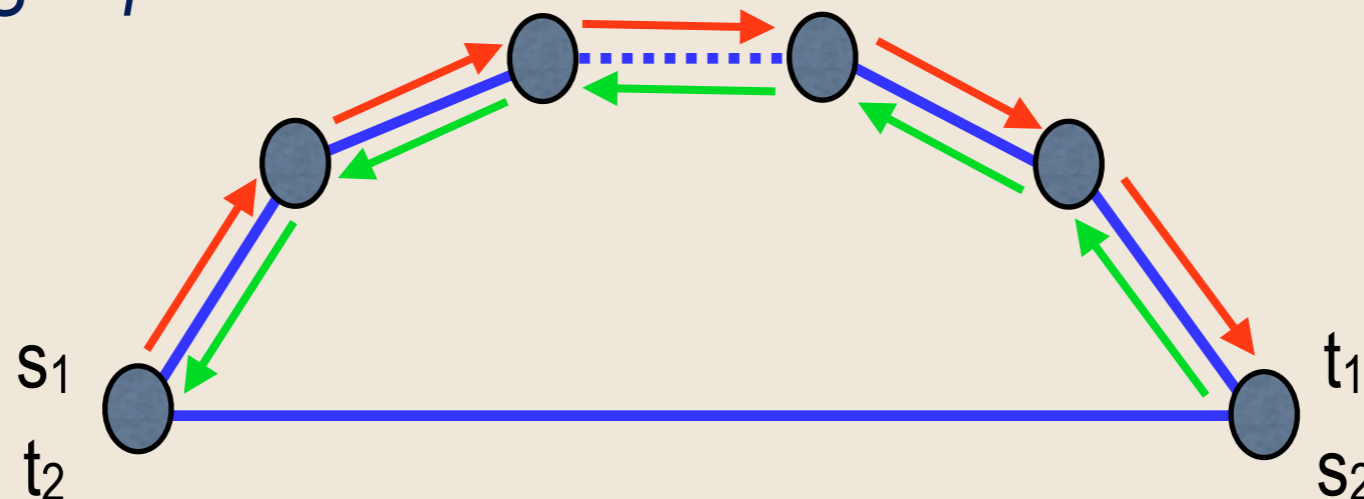
*local*

*NE exists (minimizer C)*

*POS = 1*

**Negatives**

*POA unbounded*





**Select any global cost**  $\phi : \mathcal{A} \rightarrow \mathcal{R}$

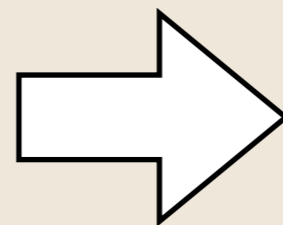
$$J_i(a) = \phi(a) - \phi(a_i^0, a_{-i})$$

**Positives**

*NE exists (minimizer  $\phi$ )*

*local*

$$\phi(a) = C(a)$$



**Negatives**

*POS = 1*

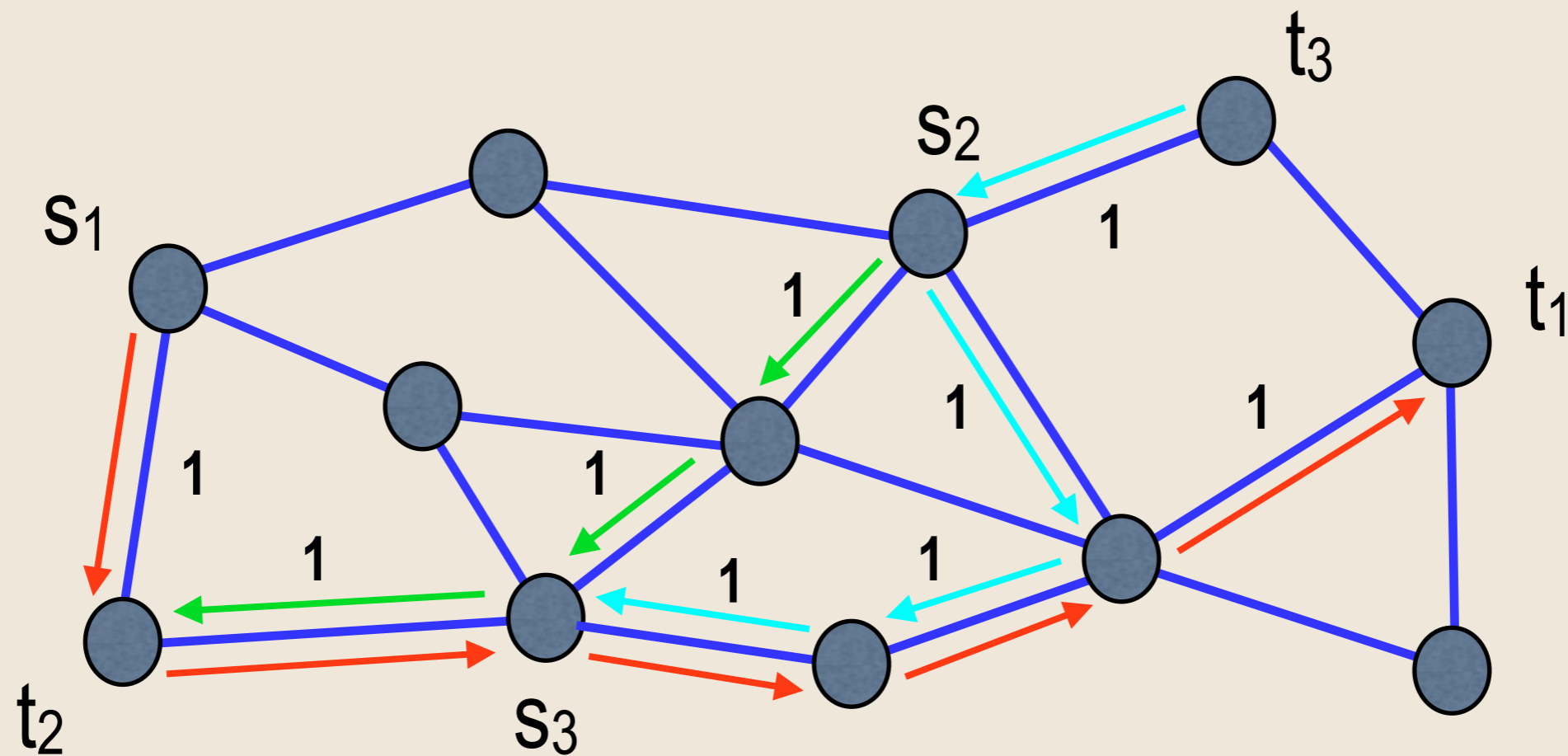
*POA unbounded*

**Can we choose a cost that gives us better equilibrium efficiency?**

**Consider**  $\phi(a) = (\alpha - 1)C(a) + \sum_i |a_i| \quad \alpha \geq 0$

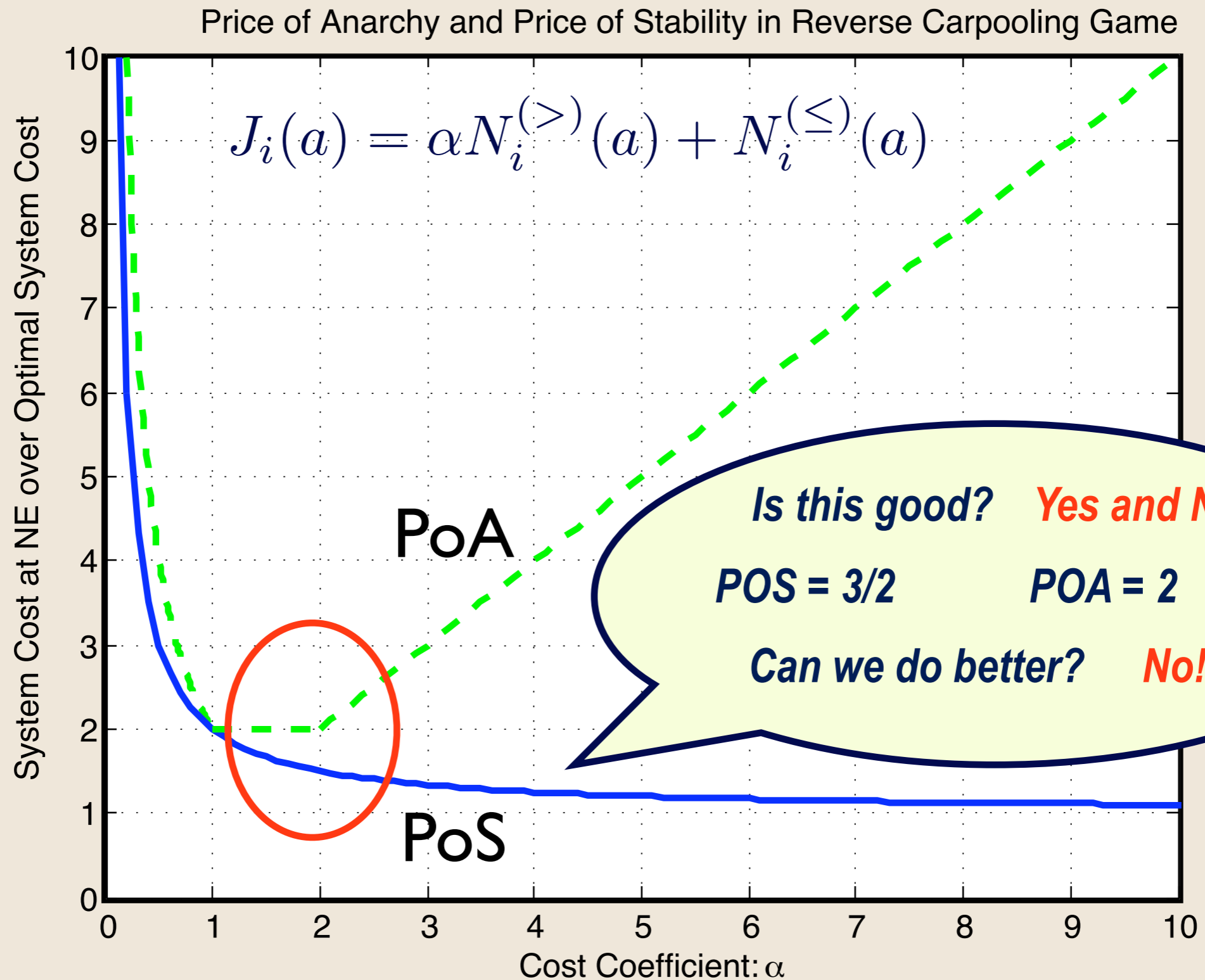
**New design:**  $J_i(a) = \phi(a) - \phi(a_i^0, a_{-i})$   
 $= \alpha N_i^{(>)}(a) + N_i^{(\leq)}(a)$

**Old design:**  $J_i(a) = C(a) - C(a_i^0, a_{-i})$   
 $= N_i^{(>)}(a)$



What is the cost of this assignment for **player 1**?

$$\begin{aligned}
 J_i(a) &= \phi(a) - \phi(a_i^0, a_{-i}) \\
 &= \alpha N_i^{(>)}(a) + N_i^{(\leq)}(a) = 2\alpha + 3
 \end{aligned}$$



## **Recap:**

- *Derived “optimal” cost functions for a simple network coding problem (WL)*
- *The results on efficiency have implications beyond simple setup*

## **Goals:**

- *Illustrate potential of game theory as mechanism for distributed optimization*
- *Results stem from recent research connecting “potential games” and cooperative control problems*

J. Marden, G. Arslan, and J. Shamma, “Connections between cooperative control and potential games,” 2008.

## **Left unsaid:**

- *How do players reach equilibrium in a distributed fashion?*
- *Existing literature on learning in games provides some algorithms*