

# Games with pure equilibria

Spring 2024

1 Best response dynamics

2 Potential games

3 Congestion games

4 References

## 1 Best response dynamics

## 2 Potential games

## 3 Congestion games

## 4 References

# Best response dynamics

Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$  with  $N = \{1, \dots, n\}$

- PNE are defined as the fix point among mutually best responses.

# Best response dynamics

Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$  with  $N = \{1, \dots, n\}$

- PNE are defined as the fix point among mutually best responses.
- It seems natural to consider variants of the process of local changes to try to get a PNE.

## Best response dynamics

Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$  with  $N = \{1, \dots, n\}$

- PNE are defined as the fix point among mutually best responses.
- It seems natural to consider variants of the process of local changes to try to get a PNE.
- Consider the algorithm:
  - choose  $s \in A_1 \times \dots \times A_n$
  - while  $s$  is not a NE do
    - choose  $i \in \{1, \dots, n\}$  such that  $s_i \notin BR(s_{-i})$
    - Set  $s_i$  to be an action in  $BR(s_{-i})$
- The process looks similar to local search algorithms. Is there any difference?

## Best response graph

- The **Nash dynamics** or **Best Response graph**  $G = (V, E)$  of  $\Gamma$ :
  - $V = A_1 \times \dots \times A_n$
  - $(s, s') \in E$  iff  $s' = (s_{-i}, s'_i)$  for  $i \in N$ ,  $s_i \notin BR(s_{-i})$  and  $s'_i \in BR(s_{-i})$ .
- Performing local search on the best response graph
  - Does it produce a PNE?
  - If so, how much time?
  - Let's look to some examples.

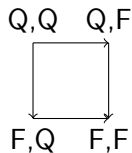
# Prisoner's dilemma

	Quiet	Fink
Quiet	2,2	0,3
Fink	3,0	1,1



# Prisoner's dilemma

	Quiet	Fink
Quiet	2,2	0,3
Fink	3,0	1,1

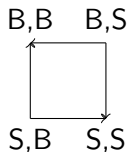


# Bach and Stravinsky

	Bach	Stravinsky
Bach	2,1	0,0
Stravinsky	0,0	1,2

# Bach and Stravinsky

	Bach	Stravinsky
Bach	2,1	0,0
Stravinsky	0,0	1,2

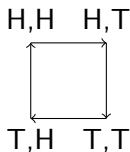


# Matching Pennies

	Head	Tail
Head	1,-1	-1,1
Tail	-1,1	1,-1

# Matching Pennies

	Head	Tail
Head	1,-1	-1,1
Tail	-1,1	1,-1



## Other games

What about Congestion Games?

- In those games we cannot get the best response graph in polynomial time.
- However we can perform a local improvement step in polynomial time.
- Although, even assuring convergence, it might take exponential time to reach a NE.

## Best response graph: Properties

- A NE is a sink, a node with out-degree 0, in the best response graph.
- The existence of a cycle in the best response graph does not rule out the existence of a PNE.
- If the best response graph is acyclic, the game has a PNE.

## Best response graph: Properties

- A **NE** is a **sink**, a node with out-degree 0, in the best response graph.
- The existence of a **cycle** in the best response graph does not rule out the existence of a PNE.
- If the **best response graph** is **acyclic**, the game has a **PNE**.
- If **best response dynamics converges** to a PNE, maybe with a lot of time.



- 1 Best response dynamics
- 2 Potential games**
- 3 Congestion games
- 4 References

# Potential games

(Monderer and Shapley 96)

- Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$ .  
Let  $S = A_1 \times \cdots \times A_n$ .

# Potential games

(Monderer and Shapley 96)

- Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$ .  
Let  $S = A_1 \times \cdots \times A_n$ .
- A function  $\Phi : S \rightarrow \mathbb{R}$  is an **exact potential function** for  $\Gamma$  if

$$\forall i \in N, \forall s \in S, \forall s'_i \in A_i : u_i(s) - u_i(s_{-i}, s'_i) = \Phi(s) - \Phi(s_{-i}, s'_i)$$

# Potential games

(Monderer and Shapley 96)

- Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$ .  
Let  $S = A_1 \times \cdots \times A_n$ .
- A function  $\Phi : S \rightarrow \mathbb{R}$  is an **exact potential function** for  $\Gamma$  if

$$\forall i \in N, \forall s \in S, \forall s'_i \in A_i : u_i(s) - u_i(s_{-i}, s'_i) = \Phi(s) - \Phi(s_{-i}, s'_i)$$

- A function  $\Phi : S \rightarrow \mathbb{R}$  is a **potential function** for  $\Gamma$  if

$$\forall i \in N, \forall s \in S, \forall s'_i \in A_i,$$

$$u_i(s) - u_i(s_{-i}, s'_i) = \Phi(s) - \Phi(s_{-i}, s'_i) = 0$$

$$\text{or } (u_i(s) - u_i(s_{-i}, s'_i))(\Phi(s) - \Phi(s_{-i}, s'_i)) > 0$$

# Potential games

(Monderer and Shapley 96)

- Consider a strategic game  $\Gamma = \langle N, (A_i)_{i \in N}, (u_i)_{i \in N} \rangle$ .  
Let  $S = A_1 \times \cdots \times A_n$ .
- A function  $\Phi : S \rightarrow \mathbb{R}$  is an **exact potential function** for  $\Gamma$  if

$$\forall i \in N, \forall s \in S, \forall s'_i \in A_i : u_i(s) - u_i(s_{-i}, s'_i) = \Phi(s) - \Phi(s_{-i}, s'_i)$$

- A function  $\Phi : S \rightarrow \mathbb{R}$  is a **potential function** for  $\Gamma$  if

$$\forall i \in N, \forall s \in S, \forall s'_i \in A_i,$$

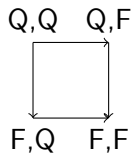
$$u_i(s) - u_i(s_{-i}, s'_i) = \Phi(s) - \Phi(s_{-i}, s'_i) = 0$$

$$\text{or } (u_i(s) - u_i(s_{-i}, s'_i))(\Phi(s) - \Phi(s_{-i}, s'_i)) > 0$$

- $\Gamma$  is a **potential game** if it admits a potential function.

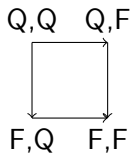
# Prisoner's dilemma

	Quiet	Fink
Quiet	2,2	0,3
Fink	3,0	1,1



# Prisoner's dilemma

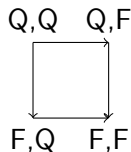
	Quiet	Fink
Quiet	2,2	0,3
Fink	3,0	1,1



$\phi$	Quiet	Fink
Quiet	1	2
Fink	2	3

# Prisoner's dilemma

	Quiet	Fink
Quiet	2,2	0,3
Fink	3,0	1,1



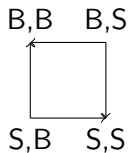
$\Phi$	Quiet	Fink
Quiet	1	2
Fink	2	3

$\Phi$  is an exact potential function



# Bach and Stravinsky

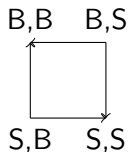
	Bach	Stravinsky
Bach	2,1	0,0
Stravinsky	0,0	1,2



# Bach and Stravinsky

	Bach	Stravinsky
Bach	2,1	0,0
Stravinsky	0,0	1,2

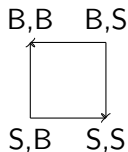
$\phi$	Bach	Stravinsky
Bach	2	1
Stravinsky	0	2



# Bach and Stravinsky

	Bach	Stravinsky
Bach	2,1	0,0
Stravinsky	0,0	1,2

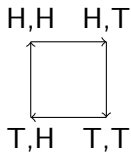
$\Phi$	Bach	Stravinsky
Bach	2	1
Stravinsky	0	2



$\Phi$  is an exact potential function

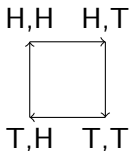
# Matching Pennies

	Head	Tail
Head	1,-1	-1,1
Tail	-1,1	1,-1



# Matching Pennies

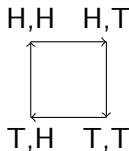
	Head	Tail
Head	1,-1	-1,1
Tail	-1,1	1,-1



This is not a potential game

# Matching Pennies

	Head	Tail
Head	1,-1	-1,1
Tail	-1,1	1,-1



This is not a potential game

The property on  $\Phi$  cannot hold along a cycle in the best response graph.

# Potential games

## Theorem

*A strategic game is a potential game iff the best response graph is acyclic*

# Potential games

## Theorem

*A strategic game is a potential game iff the best response graph is acyclic*

## Proof.

Let  $G$  be the best response graph of  $\Gamma$ .



# Potential games

## Theorem

*A strategic game is a potential game iff the best response graph is acyclic*

## Proof.

Let  $G$  be the best response graph of  $\Gamma$ .

- If  $G$  is acyclic, a topological sort of the graph provides a potential function for  $\Gamma$ .

# Potential games

## Theorem

*A strategic game is a potential game iff the best response graph is acyclic*

## Proof.

Let  $G$  be the best response graph of  $\Gamma$ .

- If  $G$  is acyclic, a topological sort of the graph provides a potential function for  $\Gamma$ .
- The existence of a potential function  $\Phi$  and the fact that, for each pair of connected strategy profiles in  $G$ , at least one player improves, implies the non existence of cycles in  $G$ .



# Potential games

## Theorem

*Any potential game has a PNE*

# Potential games

## Theorem

*Any potential game has a PNE*

## Proof.

As the best response graph is acyclic it must have a sink. □

# Potential games

## Theorem

*Any potential game has a PNE*

## Proof.

As the best response graph is acyclic it must have a sink. ☐

We have a way to show that a game has a PNE by showing that it is a potential game.

1 Best response dynamics

2 Potential games

3 **Congestion games**

4 References

# Congestion games

A **congestion game** is defined by

- A finite set  $E$  of  $m$  resources
- A finite set  $N$  of  $n$  players
- A **delay function**  $d: E \times \mathbb{N} \rightarrow \mathbb{Z}$
- For each player  $i \in N$ :
  - A set of actions  $A_i \in 2^E$
  - A cost function  $c_i$ :

$$c_i(a_1, \dots, a_n) = \left( \sum_{e \in a_i} d(e, f(a_1, \dots, a_n, e)) \right)$$

being  $f(a_1, \dots, a_n, e) = |\{i \mid e \in a_i\}|$ .

# Congestion games

A **congestion game** is defined by

- A finite set  $E$  of  $m$  resources
- A finite set  $N$  of  $n$  players
- A **delay function**  $d: E \times \mathbb{N} \rightarrow \mathbb{Z}$
- For each player  $i \in N$ :
  - A set of actions  $A_i \in 2^E$
  - A cost function  $c_i$ :

$$c_i(a_1, \dots, a_n) = \left( \sum_{e \in a_i} d(e, f(a_1, \dots, a_n, e)) \right)$$

being  $f(a_1, \dots, a_n, e) = |\{i \mid e \in a_i\}|$ .

A **singleton congestion game** has  $A_i = \{\{r\} \mid r \in E\}$ .



# An example of a congestion game

## An example of a congestion game

- We have a factory with two end production lines, each having a cutting and a packing unit. Orders are cut down and then packed.

## An example of a congestion game

- We have a factory with two end production lines, each having a cutting and a packing unit. Orders are cut down and then packed.
- We have 3 orders that have to be send to one of the end production lines.

## An example of a congestion game

- We have a factory with two end production lines, each having a cutting and a packing unit. Orders are cut down and then packed.
- We have 3 orders that have to be send to one of the end production lines.
- The cutting machine on the first line takes 1 hour to process a single order, 2 hours to process 2 and 4 hours to process 3. The cutting machine on the second line takes 4, 5 and 9 hours respectively.

## An example of a congestion game

- We have a factory with two end production lines, each having a cutting and a packing unit. Orders are cut down and then packed.
- We have 3 orders that have to be send to one of the end production lines.
- The cutting machine on the first line takes 1 hour to process a single order, 2 hours to process 2 and 4 hours to process 3. The cutting machine on the second line takes 4, 5 and 9 hours respectively.
- The packing machine on the first line takes 2 additional hours to pack a single order, 3 hours to pack 2 and 7 hours to pack 3. The packing machine on the second line takes instead 0, 2 and 9 hours respectively.

# An example of a congestion game

- We have 4 resources  $C_1, C_2, P_1, P_2$  and 3 players  $N = \{1, 2, 3\}$
- $A_i = \{\{C_1, P_1\}, \{C_2, P_2\}\}$ ,  $i = 1, 2, 3$
- Delay functions are defined by the processing times.

	1	2	3
$C_1$	1	2	4
$C_2$	4	5	9
$P_1$	2	3	7
$P_2$	0	2	9

## An example of a congestion game

- We have 4 resources  $C_1, C_2, P_1, P_2$  and 3 players  $N = \{1, 2, 3\}$
- $A_i = \{\{C_1, P_1\}, \{C_2, P_2\}\}, i = 1, 2, 3$
- Delay functions are defined by the processing times.

	1	2	3
$C_1$	1	2	4
$C_2$	4	5	9
$P_1$	2	3	7
$P_2$	0	2	9

Does this game have a PNE?

# Rosenthal's theorem

## Theorem (Rosenthal 73)

*Every congestion game is a potential game,*



# Rosenthal's theorem

## Theorem (Rosenthal 73)

*Every congestion game is a potential game,*

- For a strategy profile  $s = (a_1, \dots, a_n)$ , define

$$\Phi(s) = \sum_{e \in r(s)} \sum_{k=1}^{f(s,e)} d(e, k)$$

where  $r(s) = \cup_{i \in N} a_i$ .

# Rosenthal's theorem

## Theorem (Rosenthal 73)

*Every congestion game is a potential game,*

- For a strategy profile  $s = (a_1, \dots, a_n)$ , define

$$\Phi(s) = \sum_{e \in r(s)} \sum_{k=1}^{f(s,e)} d(e, k)$$

where  $r(s) = \cup_{i \in N} a_i$ .

Let us show that  $\Phi$  is a potential function.

- Let  $s = (a_1, \dots, a_n)$ . Fix a player  $i$  and let  $a'_i \subseteq E$  and  $s' = (s_{-i}, a'_i)$ . We have

- Let  $s = (a_1, \dots, a_n)$ . Fix a player  $i$  and let  $a'_i \subseteq E$  and  $s' = (s_{-i}, a'_i)$ . We have

$$c_i(s) - c_i(s_{-i}, a'_i) = \left( \sum_{e \in a_i} d(e, f(s, e)) \right) - \left( \sum_{e' \in a'_i} d(e, f(s', e')) \right)$$

- Let  $s = (a_1, \dots, a_n)$ . Fix a player  $i$  and let  $a'_i \subseteq E$  and  $s' = (s_{-i}, a'_i)$ . We have

$$c_i(s) - c_i(s_{-i}, a'_i) = \left( \sum_{e \in a_i} d(e, f(s, e)) \right) - \left( \sum_{e' \in a'_i} d(e, f(s', e')) \right)$$

$$\Phi(s) - \Phi(s') = \sum_{e \in r(s)} \sum_{k=1}^{f(s,e)} d(e, k) - \sum_{e' \in r(s')} \sum_{k=1}^{f(s',e')} d(e', k)$$

# Cost difference

- Note that

# Cost difference

- Note that
  - $e \in a_i \cap a'_i$ :  $f(s, e) = f(s', e)$
  - $e \notin a_i$  and  $e \notin a'_i$ :  $f(s, e) = f(s', e)$

## Cost difference

- Note that

- $e \in a_i \cap a'_i: f(s, e) = f(s', e)$
- $e \notin a_i$  and  $e \notin a'_i: f(s, e) = f(s', e)$

$$\begin{aligned} c_i(s) - c_i(s_{-i}, s'_i) &= \left( \sum_{e \in a_i} d(e, f(s, e)) \right) - \left( \sum_{e' \in a'_i} d(e, f(s', e')) \right) \\ &= \sum_{e \in a_i, e \notin a'_i} d(e, f(s, e)) - \sum_{e \notin a_i, e \in a'_i} d(e, f(s', e')) \end{aligned}$$



# Potential difference

- Furthermore,

# Potential difference

- Furthermore,
  - $e \in a_i$  and  $e \notin a'_i$ :  $f(s, e) = f(s', e) + 1$
  - $e \notin a_i$  and  $e \in a'_i$ :  $f(s, e) + 1 = f(s', e)$

## Potential difference

- Furthermore,

- $e \in a_i$  and  $e \notin a'_i$ :  $f(s, e) = f(s', e) + 1$
- $e \notin a_i$  and  $e \in a'_i$ :  $f(s, e) + 1 = f(s', e)$

$$\begin{aligned}
 \Phi(s) - \Phi(s') &= \sum_{e \in r(s)} \sum_{k=1}^{f(s,e)} d(e, k) - \sum_{e \in r(s')} \sum_{k=1}^{f(s',e)} d(e, k) \\
 &= \sum_{e \in a_i, e \notin a'_i} \left[ \sum_{k=1}^{f(s',e)+1} d(e, k) - \sum_{k=1}^{f(s',e)} d(e, k) \right] \\
 &\quad + \sum_{e \notin a_i, e \in a'_i} \left[ \sum_{k=1}^{f(s,e)} d(e, k) - \sum_{k=1}^{f(s,e)+1} d(e, k) \right]
 \end{aligned}$$

## Potential difference

$$\begin{aligned}
 &= \sum_{e \in a_i, e \notin a'_i} \left[ \sum_{k=1}^{f(s',e)+1} d(e, k) - \sum_{k=1}^{f(s',e)} d(e, k) \right] \\
 &\quad + \sum_{e \notin a_i, e \in a'_i} \left[ \sum_{k=1}^{f(s,e)} d(e, k) - \sum_{k=1}^{f(s,e)+1} d(e, k) \right] \\
 &= \sum_{e \in a_i, e \notin a'_i} d(e, f(s', e) + 1) - \sum_{e \notin a_i, e \in a'_i} d(e, f(s, e) + 1) \\
 &= \sum_{e \in a_i, e \notin a'_i} d(e, f(s, e)) - \sum_{e \notin a_i, e \in a'_i} d(e, f(s', e)) \\
 &= c_i(s) - c_i(s_{-i}, s'_i)
 \end{aligned}$$

# Network congestion games

- A **network congestion game** is a congestion game defined by a directed graph  $G$  and a collection of pairs of vertices  $(s_i, t_i)$ .
  - The set of resources are the arcs in  $G$ .
  - The actions, for player  $i$ , are the  $s_i \rightarrow t_i$  paths on  $G$ .
- A network congestion game is **symmetric** when  $s_i = s$  and  $t_i = t$ , for  $i \in N$ .

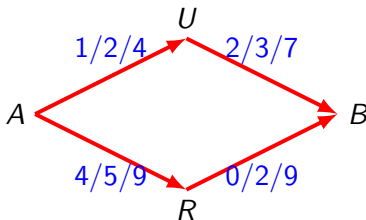
# An example of a network congestion game

# An example of a network congestion game

- There are three players.
- and a network (with a delay function on arcs)

## An example of a network congestion game

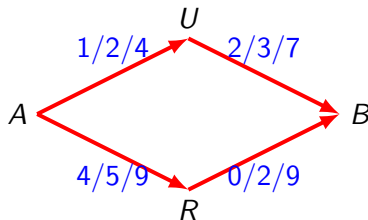
- There are three players.
- and a network (with a delay function on arcs)





# An example of a network congestion game

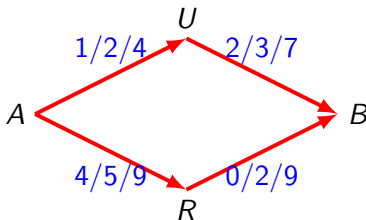
- There are three players.
- and a network (with a delay function on arcs)



- Player's objective: going from  $s = A$  to  $t = B$  as fast as possible.

## An example of a network congestion game

- There are three players.
- and a network (with a delay function on arcs)



- Player's objective: going from  $s = A$  to  $t = B$  as fast as possible.
- Strategy profiles: paths from  $A$  to  $B$ .
- A NE?

# An example of a weighted network congestion game

# An example of a weighted network congestion game

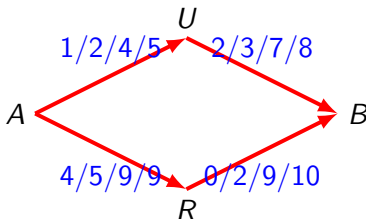
- There are three players with weights 1,1,2

# An example of a weighted network congestion game

- There are three players with weights 1,1,2
- and a network (with a delay function on arcs)

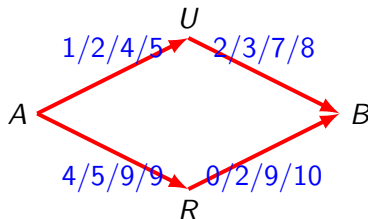
# An example of a weighted network congestion game

- There are three players with weights 1,1,2
- and a network (with a delay function on arcs)



# An example of a weighted network congestion game

- There are three players with weights 1,1,2
- and a network (with a delay function on arcs)



- Player's objective: send  $w_i$  units from  $s = A$  to  $t = B$  as fast as possible.
- Strategy profiles: paths from  $A$  to  $B$ .
- A NE?

## Results on convergence time

### Theorem (Fabrikant, Papadimitriou, Talwar (STOC 04))

*There exist network congestion games with an initial strategy profile from which all better response sequences have exponential length.*

### Theorem (Fabrikant, Papadimitriou, Talwar (STOC 04))

*There is a polynomial time algorithm for finding a PNE in symmetric network congestion games.*



## Results on convergence time

### Theorem (Fabrikant, Papadimitriou, Talwar (STOC 04))

*There exist network congestion games with an initial strategy profile from which all better response sequences have exponential length.*

### Theorem (Fabrikant, Papadimitriou, Talwar (STOC 04))

*There is a polynomial time algorithm for finding a PNE in symmetric network congestion games.*

### Theorem (leong, McGrew, Nudelman, Shoham, Sun (AAAI 05))

*In singleton congestion games all best response sequences have length at most  $n^2 m$ .*

## Complexity classification?

## Optimization problem

An **optimization problem** is defined by a structure

$\Pi = (I, \text{sol}, m, \text{goal})$ , where

- $I$  is the input set to  $\Pi$ ;
- $\text{sol}(x)$  is the set of feasible solutions for an input  $x$ .
- $m$  is an integer measure defined over pairs  $(x, y)$ ,  $x \in I$  and  $y \in \text{sol}(x)$ .
- $\text{goal}$  is the optimization criterium MAX or MIN.

An optimization problem is a functional problem whose goal, given an instance  $x$ , is to find an optimum solution

$$y = \text{goal}\{(m(x, y') \mid y' \in \text{sol}(x))\}.$$

**Example:** Given a graph and two vertices, obtain a path joining them with minimum length.

# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .

# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .
- A **local optimum** is a solution such that all its neighbors have equal or worse cost.

# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .
- A **local optimum** is a solution such that all its neighbors have equal or worse cost.

(Johnson, Papadimitriou, Yannakakis, FOCS 85)

A local search problem belongs to **PLS (Polynomial Local Search)**

# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .
- A **local optimum** is a solution such that all its neighbors have equal or worse cost.

(Johnson, Papadimitriou, Yannakakis, FOCS 85)

A local search problem belongs to **PLS (Polynomial Local Search)** if polynomial time algorithms exist for

# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .
- A **local optimum** is a solution such that all its neighbors have equal or worse cost.

(Johnson, Papadimitriou, Yannakakis, FOCS 85)

A local search problem belongs to **PLS (Polynomial Local Search)** if polynomial time algorithms exist for

- finding **initial feasible solution**  $s \in \text{sol}(x)$ ,

# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .
- A **local optimum** is a solution such that all its neighbors have equal or worse cost.

(Johnson, Papadimitriou, Yannakakis, FOCS 85)

A local search problem belongs to **PLS (Polynomial Local Search)** if polynomial time algorithms exist for

- finding **initial feasible solution**  $s \in \text{sol}(x)$ ,
- computing the **objective measure**  $m(x, y)$ ,



# PLS

- A **local search problem** is an optimization problem with a **neighborhood** structure defined on the solution set  $\mathcal{N}(\text{sol}(x))$ .
- A **local optimum** is a solution such that all its neighbors have equal or worse cost.

(Johnson, Papadimitriou, Yannakakis, FOCS 85)

A local search problem belongs to **PLS (Polynomial Local Search)** if polynomial time algorithms exist for

- finding **initial feasible solution**  $s \in \text{sol}(x)$ ,
- computing the **objective measure**  $m(x, y)$ ,
- checking whether a solution is a **local optimum** and if not finding a **better solution in the neighborhood**.

# PLS reductions

A **PLS reduction** from  $(\Pi_1, \mathcal{N}_1)$  to  $(\Pi_2, \mathcal{N}_2)$  is

- a polynomial time computable function  $f : I_{\Pi_1} \rightarrow I_{\Pi_2}$  and
- a polynomial time computable function  $g : \text{sol}_{\Pi_2}(f(x)) \rightarrow \text{sol}_{\Pi_1}(x)$ , for  $x \in I_{\Pi_1}$  such that
- if  $s_2 \in \text{sol}_{\Pi_2}(f(x))$  locally optimal then  $g(s_2)$  is locally optimal.

# PLS reductions

A **PLS reduction** from  $(\Pi_1, \mathcal{N}_1)$  to  $(\Pi_2, \mathcal{N}_2)$  is

- a polynomial time computable function  $f : I_{\Pi_1} \rightarrow I_{\Pi_2}$  and
- a polynomial time computable function  $g : \text{sol}_{\Pi_2}(f(x)) \rightarrow \text{sol}_{\Pi_1}(x)$ , for  $x \in I_{\Pi_1}$  such that
- if  $s_2 \in \text{sol}_{\Pi_2}(f(x))$  locally optimal then  $g(s_2)$  is locally optimal.

So that,

- If a local opt of  $\Pi_2$  is “easy” to find then a local opt of  $\Pi_1$  is easy to find.
- If a local opt of  $\Pi_1$  is “hard” to find then a local opt of  $\Pi_2$  is hard to find.

# PLS reductions

A **PLS reduction** from  $(\Pi_1, \mathcal{N}_1)$  to  $(\Pi_2, \mathcal{N}_2)$  is

- a polynomial time computable function  $f : I_{\Pi_1} \rightarrow I_{\Pi_2}$  and
- a polynomial time computable function  $g : \text{sol}_{\Pi_2}(f(x)) \rightarrow \text{sol}_{\Pi_1}(x)$ , for  $x \in I_{\Pi_1}$  such that
- if  $s_2 \in \text{sol}_{\Pi_2}(f(x))$  locally optimal then  $g(s_2)$  is locally optimal.

So that,

- If a local opt of  $\Pi_2$  is “easy” to find then a local opt of  $\Pi_1$  is easy to find.
- If a local opt of  $\Pi_1$  is “hard” to find then a local opt of  $\Pi_2$  is hard to find.

A PLS problem  $(\Pi, \mathcal{N})$  is **PLS-complete** if every problem in PLS is PLS-reducible to  $(\Pi, \mathcal{N})$ .

# PLS complete problems

- **MAX-SAT** (maximum satisfiability) problem
  - Given a Boolean formula in conjunctive normal form with a positive integer weight for each clause.
  - A solution is an assignment of the value 0 or 1 to all variables.
  - Its weight, to be maximized, is the sum of the weights of all satisfied clauses.
  - As neighborhood consider the **Flip-neighborhood**, where two assignments are neighbors if one can be obtained from the other by flipping the value of a single variable.

# PLS complete problems

- **MaxCut** problem.
  - Given a graph  $G = (V, E)$  with non-negative edge weights.
  - A feasible solution is a partition of  $V$  into two sets  $A$  and  $B$ .
  - The objective is to maximize the weight of the edges between the two sets  $A$  and  $B$ .
  - In the **Flip-neighborhood** two solutions are neighbors if one can be obtained from the other by moving a single vertex from one set to the other.

# PLS completeness

## Theorem

*Computing a PNE in congestion games is PLS-complete.*

# PLS completeness

## Theorem

*Computing a PNE in congestion games is PLS-complete.*

- The problem belongs to PLS taking as neighborhood the Nash dynamics because the Rosenthal's potential function can be evaluated in polynomial time.



# PLS completeness

## Theorem

*Computing a PNE in congestion games is PLS-complete.*

- The problem belongs to PLS taking as neighborhood the Nash dynamics because the Rosenthal's potential function can be evaluated in polynomial time.
- We provide a reduction from MaxCut under the Flip-neighborhood.

# PLS completeness

Reduction from MaxCut under the Flip-neighborhood:

# PLS completeness

Reduction from MaxCut under the Flip-neighborhood:

Let  $(G, E, (w_e)_{e \in E})$  be an instance of MaxCut, define a congestion game as follows:

# PLS completeness

Reduction from MaxCut under the Flip-neighborhood:

Let  $(G, E, (w_e)_{e \in E})$  be an instance of MaxCut, define a congestion game as follows:

- For each edge  $e$ , we add **resources  $e^a$  and  $e^b$** , with delay 0 if used by only one player and delay  $w_e$  if used by more players.

# PLS completeness

Reduction from MaxCut under the Flip-neighborhood:

Let  $(G, E, (w_e)_{e \in E})$  be an instance of MaxCut, define a congestion game as follows:

- For each edge  $e$ , we add resources  $e^a$  and  $e^b$ , with delay 0 if used by only one player and delay  $w_e$  if used by more players.
- Players correspond to nodes  $V$ , and the action set of  $v \in V$ ,  
 $A_v = \{e^a, e^b \mid e \in E \text{ incident to } v\}$

# PLS completeness

- Each solution  $(A, B)$  of MaxCut corresponds to strategy profile  $s = (s_1, \dots, s_n)$  where for each  $v \in V$ ,  
 $s_v = \{e^a | e = (u, v) \wedge v \in A\} \cup \{e^b | e = (u, v) \wedge v \in B\}$

# PLS completeness

- Each solution  $(A, B)$  of MaxCut corresponds to strategy profile  $s = (s_1, \dots, s_n)$  where for each  $v \in V$ ,  
 $s_v = \{e^a | e = (u, v) \wedge v \in A\} \cup \{e^b | e = (u, v) \wedge v \in B\}$
- Local optima of the MaxCut instance coincide with the Nash equilibria of the congestion game.

- 1 Best response dynamics
- 2 Potential games
- 3 Congestion games
- 4 References**



# Reference

B. Vöcking, Congestion Games: Optimization in Competition