

# Coalition games on interaction graphs

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joint work with Zhentao Li and Adrian Vetta

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- Unfortunately, people are **selfish** : if it is more interesting for them, they will create a project of their own.
- Solution : distribute payoff in such a way people do not want to leave the **grand coalition**.

## Coalition games

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- A set  $I$  of  $n$  agents.
- A valuation function  $v : 2^n \rightarrow \mathbb{N}$ . (the value generated by any group deciding to work on another project)

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A subset  $S$  of agents is a **coalition** if  $v(S)$  is positive.

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## Definition (coalition)

A subset  $S$  of agents is a **coalition** if  $v(S)$  is positive.

## Goal :

Distribute money to the agents in such a way, every coalition  $S$  received at least  $v(S)$  euros.

⇒ No one wants to leave the *grand coalition* and work on its own project.

**Definition (core)**

The **core** of the coalition game is the **set of vectors of payoff** satisfying the following constraints :

$$\begin{aligned} \sum_{i \in I} x_i &= v(I) \\ \text{and } \sum_{i \in S} x_i &\geq v(S) & \forall S \subseteq I \\ x_i &\geq 0 & \forall i \in I \end{aligned}$$

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## Why Linear Programming ?

- Real Numbers Linear Programming can be solved in **polynomial** time.
- Integral Linear Programming are **NP-hard** to solve.

## Relative cost of stability

### **Another approach (relative cost of stability) :**

How much money must be injected by an external authority to stabilize the system ?

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If not, what can be generated by the whole set of agents ?

$$\begin{aligned} \nu(\mathcal{G}) = \max \quad & \sum_{S: S \subseteq I} v(S) \cdot y_S \\ \text{s.t.} \quad & \sum_{S \subseteq I: i \in S} y_S \leq 1 \quad \forall i \in I \\ & y_S \text{ is integral and } \geq 0 \quad \forall S \subseteq I \end{aligned}$$

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**Remark :** It is an (integral) LP.

## Our expenses

**Our expenses :** (Covering-LP)

$$\begin{aligned} \tau(\mathcal{G})^* = \min \quad & \sum_{i \in I} x_i \\ \text{s.t.} \quad & \sum_{i: i \in S} x_i \geq v(S) \quad \forall S \subseteq I \\ & x_i \geq 0 \end{aligned}$$

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**Simplification :**

$\Rightarrow$  Since  $\tau^* - \nu$  is not “stable” (by disjoint copy for instance), we consider  $\frac{\tau^*}{\nu}$  instead.

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The relative cost of stability represents the ratio between the minimum payment stabilizing the system and the total wealth the grand coalition can generate.

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By Strong Duality Theorem, we have :

$$\nu(\mathcal{G}) \leq \nu^*(\mathcal{G}) = \tau^*(\mathcal{G}) \leq \tau(\mathcal{G})$$

Thus

$$\frac{\tau^*}{\nu} = \frac{\nu^*}{\nu} \leq \frac{\tau}{\nu}$$

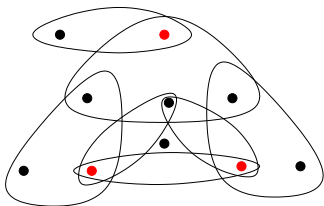
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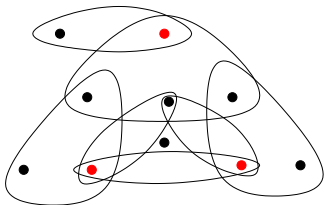


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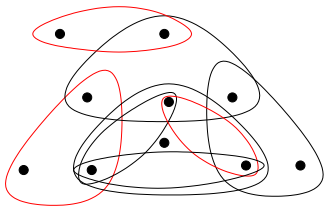
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$$\text{s.t. } \sum_{S \subseteq I: i \in S} y_S \leq 1 \quad \forall i \in I$$





## Worst case

### Lemma

The Packing-Covering ratio  $\frac{\tau(\mathcal{G})}{\nu(\mathcal{G})}$  (and both integrality gaps) can be **arbitrarily large**.

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### Example :

- A set of agents  $I = \{1, \dots, n\}$
- A subset  $S$  of  $I$  is a coalition if and only if  $|S| > |I|/2$ .
- Maximum Packing : **1**.
- Minimum Covering :  $\geq \frac{|I|}{2} - 1$ .

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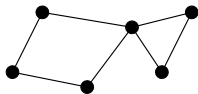
### Question

When can we bound  $\frac{\tau}{\nu}$  ?

# Graphs

A **graph** is a pair  $(V, E)$  where :

- $V$  is a set of “points” called **vertices** ;
- $E$  is a set of links between points called **edges**.

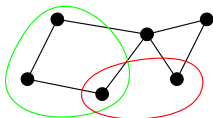


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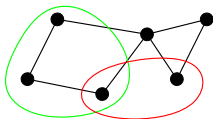


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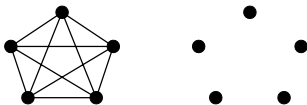
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A **clique** is a graph where vertices are pairwise incident.

A **stable set** is a graph where vertices are pairwise not incident.



## Interaction graph

Myerson proposed the following model :

### Definition (interaction graph)

Let  $G$  be a graph where the vertices of  $G$  are the agents of the coalition game  $\mathcal{G}$ .

The game  $\mathcal{G}$  has **interaction graph**  $G$  if every coalition is **connected** (i.e., if  $v(S) > 0$  then  $S$  is connected).

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**Interpretation :** The agents must be able to communicate if they want to create a coalition.

### Examples :

- $G$  is a **clique** : any coalition may exist.
- $G$  is a **stable set** : coalitions have size one.

## Treewidth and coalition game

### Theorem (Meir et al.)

Let  $G$  be a graph. We have the following inequality :

$$\frac{\tau(\mathcal{G})}{\nu(\mathcal{G})} \leq_{\forall} tw(G) + 1$$

Moreover there exist graphs for which this bound is tight.

By  $\leq_{\forall}$ , we mean that every game  $\mathcal{G}$  on interaction graph  $G$  satisfies this inequality.

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**Our work :** Improve this result, and bounds on the relative cost of stability.

## Treewidth

A tree  $T$  and a (bag) function  $f : T \rightarrow 2^V$  is a **tree decomposition** of  $G = (V, E)$  if :

- For every  $v \in V$ , the set of nodes containing  $v$  in their bags is a subtree  $T_v$  of  $T$ .
- For every edge  $uv$ ,  $T_u$  and  $T_v$  **intersects**.

The **width** of a decomposition is the maximum size of a bag of the tree-decomposition minus one.

### Definition (treewidth)

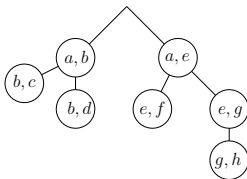
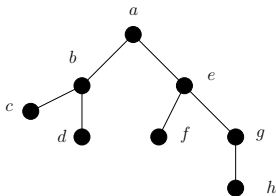
The **treewidth** of  $G$ ,  $tw(G)$ , is the minimum width of a tree-decomposition of  $G$ .

## Examples

- $K_n$  has a tree decomposition of width  $n - 1$  (all the vertices are in the same bag).

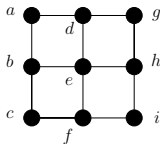
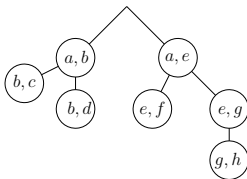
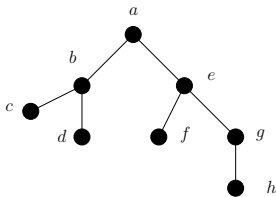
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- The  $d \times d$  grid has a tree decomposition of width  $d$ .



## A new invariant : vinewidth

A tree  $T$  and a function  $f : T \rightarrow 2^V$  is a **vine decomposition** of  $G = (V, E)$  if :

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- For every edge  $uv$ ,  $T_u$  and  $T_v$  **intersects or share an edge**.

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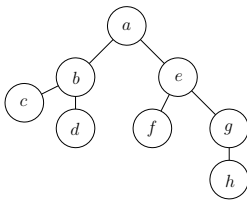
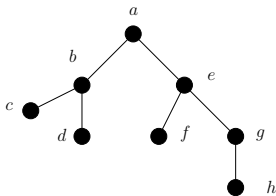


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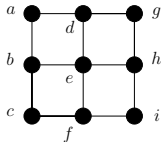
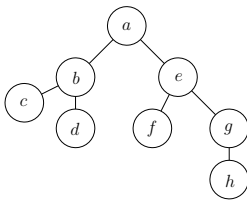
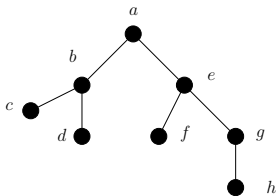
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## Link between treewidth and vwidth

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Every graph  $G$  satisfies

$$\frac{tw(G) + 1}{2} \leq vw(G) \leq tw(G) + 1$$

Moreover both inequalities are tight.

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### Proof sketch :

- Any tree decomposition is a vine decomposition.  
 $\Rightarrow vw(G) \leq tw(G) + 1.$
- We make the “union” of every pair of adjacent bags to be sure that  $T_u$  and  $T_v$  intersect.  
 $\Rightarrow tw(G) \leq 2 \cdot vw(G) - 1.$

## Main statement

$$\tau = \min \sum_{i \in I} x_i$$

$$\text{s.t. } \sum_{i: i \in S} x_i \geq v(S) \quad \forall S \subseteq I$$

$$\nu = \max \sum_{S: S \subseteq I} v(S) \cdot y_S$$

$$\text{s.t. } \sum_{S \subseteq I: i \in S} y_S \leq 1 \quad \forall i \in I$$

### Theorem (BLV'14)

For every graph  $G$ , we have :

$$vw(G) \leq_{\exists} \frac{\tau(G)}{\nu(G)} \leq_{\forall} vw(G)$$

By  $\leq_{\forall}$ , we mean that every game  $\mathcal{G}$  on interaction graph  $G$  satisfies this inequality.

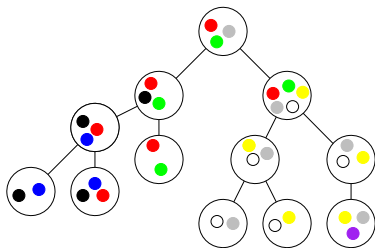
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Bottom-up from the leaves of a (rooted) vine decomposition.

- If all the vertices of a coalition  $C$  are in the bag  $B_f$  of a leaf  $f$ , add vertices of  $B_f$  in the covering and  $C$  in the packing. Delete the coalitions containing one vertex of  $B_f$ .
- Otherwise, delete all the leaves of  $T$ .

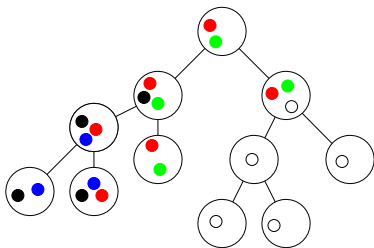




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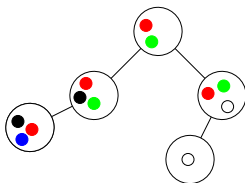
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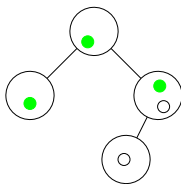
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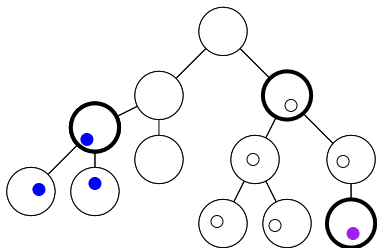
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- We have selected  $k$  disjoint coalitions :  $\nu(G) \geq k$ .
- We have deleted at most  $vw(G) \cdot k$  vertices.

## Relative cost of stability

### Reminder of the motivation :

We want to bound the relative cost of stability  $\frac{\tau^*}{\nu}$ .

#### Theorem

Let  $G$  be a graph. There exists  $\delta > 0$  such that

$$vw(G)^\delta \leq \exists \frac{\nu^*(\mathcal{G})}{\nu(\mathcal{G})} \leq \forall vw(G)$$

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Follows from the previous theorem since  $\frac{\nu^*(\mathcal{G})}{\nu(\mathcal{G})} \leq \frac{\tau(\mathcal{G})}{\nu(\mathcal{G})} \leq \forall vw(G)$ .

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**Proof of  $vw(G)^\delta \leq \exists \frac{\nu^*(\mathcal{G})}{\nu(\mathcal{G})}$  :**

- Prove that the gap is linear for grids.
- Use grid minor theorem.

## Conclusion

- What is the **best constant**  $\delta$  (we cannot beat  $\frac{1}{2}$ , on cliques) ?
- Does it exist a “good” invariant which **characterizes** the relative cost of stability ?
- On which interaction graph can we obtain a **linear bound** in terms of vinewidth ?



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**Thanks for your attention !**