

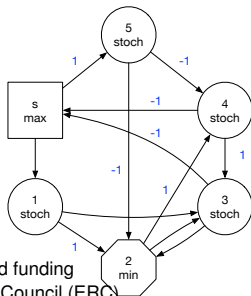
# The Connection between Stochastic Games and Constraint Satisfaction

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# Mean Payoff Games

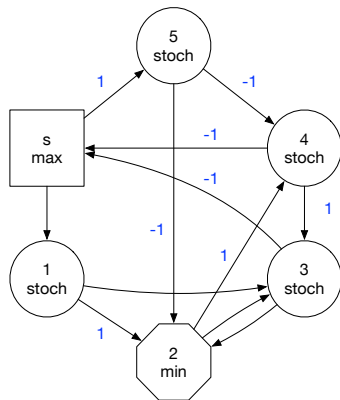
Stochastic games: Shapley 1953

A **stochastic mean payoff game** is given by

- a finite directed graph  $(V, E)$ ;
- a partition  $V = V_{\max} \uplus V_{\min} \uplus V_{\text{stoch}}$ ;
- a start vertex  $s$ ;
- a **payoff function**  
 $w: E \rightarrow \{-W, \dots, 0, \dots, +W\}$ ;
- two players **min** and **max** (a  $2\frac{1}{2}$ -game).

A **mean payoff game (MPG)**

has no stochastic vertices  
(a 2-player game).



## The rules of the game

The players move a token along edges from  $E$ .

On nodes from  $V_{\max}$  it is max' turn.

On nodes from  $V_{\min}$  it is min's turn.

On nodes from  $V_{\text{stoch}}$  an out-neighbour is chosen uniformly at random.

The game continues forever.

The **long-run average payoff** (or **limiting average payoff**) is defined to be

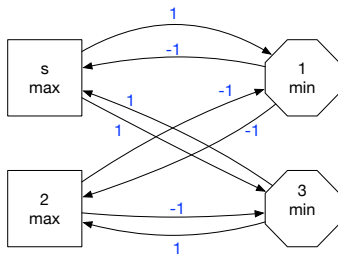
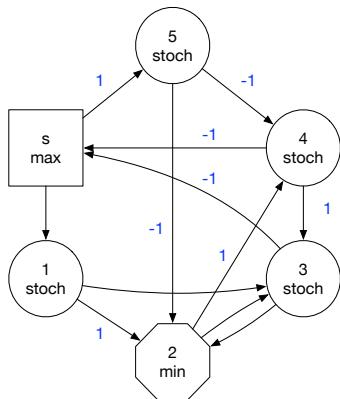
$$\liminf_{t \rightarrow \infty} \frac{1}{t} \sum_{i=1}^t w(u_{i-1}, u_i)$$

Goal of max (min): maximise (minimize) the long-run average.

**MGP as a decision problem:** decide whether max has a **winning strategy**, i.e., a strategy that guarantees non-negative long-run average.

# What is the computational complexity of mean payoff games?

# Example



# Memoryless Strategies

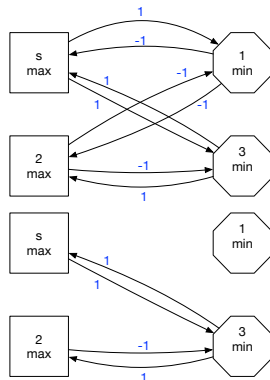
A strategy is called **memoryless** (or **positional**) if the next move only depends on the current position (and not on the history of the previous moves).

**Theorem (Ehrenfeucht+Mycielski).**

Both players in a mean payoff game have optimal memoryless strategies.

**Corollary:** Mean payoff games is in NP.

- Guess a memoryless strategy  $\sigma$  for max.
- Remove all edges not chosen by max; let  $G_\sigma$  be the resulting graph.
- min is looking in  $G_\sigma$  for a path leading to a negative cycle.
- This can be done in time  $O(|V| \cdot |E|)$  by the Bellman-Ford algorithm.



# Containment in coNP

**Corollary:** Mean payoff games are in coNP.

- Guess a memoryless strategy  $\sigma$  for min.
- Remove all edges not chosen by min; let  $G_\sigma$  be the resulting graph.
- max is looking in  $G_\sigma$  for a reachable cycle with maximum mean weight.
- This can be done in time  $O(|V| \cdot |E|)$  by Karp's algorithm for the maximum mean weight cycle and testing reachability.

**Remarks.**

- Mean payoff games not known to be in  $P$ !
- Problems in  $NP \cap coNP$  that are not known to be in  $P$  are rare.
- Already the special case of **parity games** is not known to be in  $P$ . Parity games are equivalent to model-checking for the  **$\mu$ -calculus**. Recent breakthrough (Calude, Jain, Khoussainov, Li, Stephan'17): quasipolynomial algorithm (in  $O(n^{\log(W)+6})$ ).

# Exercise

Play with one of your neighbours:

- 1** Invent on paper a mean payoff game  $G$  including a start vertex  $s$ .
- 2** Give the paper to your neighbour.
- 3** Your neighbour decides whether (s)he wants to be player max or player min in this game.
- 4** Expose a memoryless strategy.
- 5** Your neighbour exposes a memoryless strategy.
- 6** Check who is the winner.

# Max-Atoms

And/Or Precedence Constraints (Möhring+Stork+Skutella'04)  
aka Max-Atom Problem (Bezem+Nieuwenhuis+Rodriguez-Charbonell'08)  
equiv to Emptiness of Tropical Polyhedra (Akian+Gaubert+Guterman'11)  
equiv to Solvability of Tropical Linear Systems (Grigoriev+Podelskii'15) :

## Definition (Max-Atoms).

Input: A finite set of variables  $V$ , and a finite set of constraints

$$x \leq \max(y, z) + c$$

where  $x, y, z \in V$  and  $c \in \mathbb{Z}$  is **represented in binary**.

Question: Is there a mapping  $V \rightarrow \mathbb{Q}$  that satisfies all constraints?

- MSS'04, BNR'08, AGG'11: in coNP.
- MSS'04, Atserias+Maneva'09, AGG'11:

# Max-Atoms $\equiv_P$ Mean Payoff Games



# Equivalence MPGs and Max-Atom

Max-Atoms  $\equiv_P$  Mean Payoff Games: why should we care?

- Both reductions are natural, elegant, easy
- The correctness proof of the reductions is surprisingly complicated
- Side-product: new proof for the existence of memoryless strategies!
- The correspondence is the starting point for important generalisations of both MPGs and Max-Atom.
- Personal claim: CSP techniques will lead to a proof that MGPs are in P.

# Massaging MPGs

Work with computationally equivalent variant MPG\*:  
decide whether max has a winning strategy for **every** starting vertex.

- Reduction from MPG\* to MPGs:  
introduce start vertex  $s \in V_{\min}$  with edges to each  $v \in V_{\max}$ .
- Reduction from MPGs to MPG\*:  
add an edge of weight 0 from every  $v \in V_{\max}$  to start vertex  $s$ .

# Massaging Max-Atoms

## Observations.

- $x \leq \max(y_1, y_2, y_3)$  has the primitive positive definition

$$\exists z (x \leq \max(y_1, z) \wedge z \leq \max(y_2, y_3)).$$

- $x \leq \max(y_1 + c_1, y_2 + c_2)$  has the primitive positive definition

$$\exists z (x \leq \max(y_1, z) + c_1 \wedge \underbrace{z \leq \max(y_2, y_2) + (c_2 - c_1)}_{i.e., z \leq y_2 + c_2 - c_1}).$$

- $x \leq \min(y_1 + c_1, \dots, y_k + c_k)$  has the primitive positive definition

$$x \leq \max(y_1, y_1) + c_1 \wedge \dots \wedge x \leq \max(y_k, y_k) + c_k.$$

# Max-Atoms: Polymorphisms

Let  $\phi$  be a max-atoms instance.

**Lemma.**

Suppose that  $s: V \rightarrow \mathbb{Q}$  is a solution to  $\phi$ . Then for any  $d \in \mathbb{Q}$

$$x \mapsto s(x) + d \quad (\textit{Translation})$$

is a solution to  $\phi$ , too. Suppose that  $s_1, s_2: V \rightarrow \mathbb{Q}$  are solutions to  $\phi$ . Then

$$x \mapsto \max(s_1(x), s_2(x))$$

is a solution to  $\phi$ , too.

$$R_c^{\max} := \{(x, y, z) \mid x \leq \max(y, z) + c\}$$

The structure  $(\mathbb{Q}; (R_c^{\max})_{c \in \mathbb{Q}})$  has the **polymorphisms**

$$(x, y) \mapsto \max(x, y)$$

and  $x \mapsto x + d$  for all  $d \in \mathbb{Q}$ .

# Max-Atoms: Important Properties

A max-atom instance  $\phi$  is called **left-distinct** if for any two conjuncts of  $\phi$

$$x_1 \leq \max(y_1, z_1) + c_1$$

and

$$x_2 \leq \max(y_2, z_2) + c_2$$

the **head variables**  $x_1$  and  $x_2$  are distinct.

**Lemma (Bezem+Nieuvenhuis+Rodriguez-Charbonell'08).**

$\phi$  is unsatisfiable  $\Leftrightarrow \phi$  contains a left-distinct unsatisfiable subset.

**Example.**

$$x \leq \max(y, z)$$

$$y \leq \max(x, z)$$

$$z \leq \max(x, x) + 2$$

$$z \leq \max(y, y) + 1$$

# Left-distinct unsatisfiability certificates

Let  $\phi$  be a max-atoms instance and  $x$  a variable of  $\phi$ .

Let  $\psi_1, \dots, \psi_k$  be all conjuncts of  $\phi$  with head  $x$ .

**Claim.** If  $\phi$  is unsatisfiable, then  $\phi \setminus \psi_i$  is for some  $i \leq k$  unsatisfiable.

**Proof:** Assume that for every  $i \leq k$  the formula  $\phi \setminus \psi_i$  has solution  $s_i$ .

Produce solution to  $\phi$  using translations and max:

$$x \leq \max(y, z) + 1 \wedge x \leq \max(u, v) + 2$$

$$s_1 : x = 1, y = 2, z = 3, u = ?, v = ?$$

$$s_2 : x = 2, y = ?, z = ?, u = 8, v = 3$$

$$\text{Translate } s_2 : s_3 : x = 1, y = ?, z = ?, u = 7, v = 2$$

$$\text{Max } s_1, s_3 : s_4 : x = 1, y \geq 2, z \geq 3, u \geq 8, v \geq 3$$

□

Lemma follows by induction.

# Max-Atoms: Pseudo-Polynomial Algorithm

Suppose that all conjuncts of max-atom instance  $\phi$  are of the form

$$x \leq \max(y, z) + 1.$$

There is a polynomial-time algorithm to decide satisfiability of  $\phi$ .

**Theorem (Bodirsky, Martin, Mottet'18).**

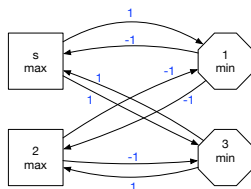
Let  $\Gamma = (\mathbb{Z}; R_1, R_2, \dots, R_m)$  be a structure with a fo-definition over  $(\mathbb{Z}; <)$ . If  $\Gamma$  has max or min or  $(\dots)$  as polymorphism, then  $\text{CSP}(\Gamma)$  is in P. Otherwise,  $\text{CSP}(\Gamma)$  is NP-complete.

**Algorithm** (in case that  $\Gamma$  has polymorphism max):

- Let  $S$  be a **sufficiently big** finite substructure of  $\Gamma$ .
- Note:  $S$  has max as polymorphism, too!
- Solve  $\phi$  as an instance of  $\text{CSP}(S)$ .
- For that, use **Arc-Consistency Algorithm**
  - **Fact:** answers correctly since  $S$  has max as polymorphism.
  - has polynomial running time in both  $\phi$  and  $S$ .

# From Mean Payoff Games To Max-Atoms

Let  $G = (V, V_{\max}, V_{\min}, w, s)$  be an MPG.



We create a max-atoms instance  $\phi_G$  with variables  $V$ :

- 1 For each  $v \in V_{\max}$  with out-neighbours  $u_1, \dots, u_k$ , add a constraint

$$v \leq \max(u_1 + w(v, u_1), \dots, u_k + w(v, u_k))$$

- 2 For each  $v \in V_{\min}$  with out-neighbours  $u_1, \dots, u_k$ , add a constraint

$$v \leq \min(u_1 + w(v, u_1), \dots, u_k + w(v, u_k))$$

(These constraints can be expressed by Max-Atom constraints.)



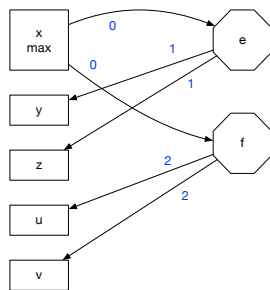
# From Max-Atoms to Mean Payoff Games

Let  $\phi$  be a max-atoms instance with variables  $V$  and inequalities  $C$ .

$$\underbrace{x \leq \max(y, z) + 1}_{=:e}, \underbrace{x \leq \max(u, v) + 2}_{=:f}$$

Create an MPG  $G_\phi$  such that  $V_{\max} := V$  and  $V_{\min} := C$ , and

- If  $e \in C$  is of the form  $x \leq \max(y, z) + k$  add edges  $(e, y), (e, z)$  of weight  $k$ ;
- For every  $x \in V$  and every  $e \in C$  of the form  $x \leq \max(y, z) + c$  add an edge  $(x, e)$  of weight 0.



# Ideas of Correctness Proofs, 1

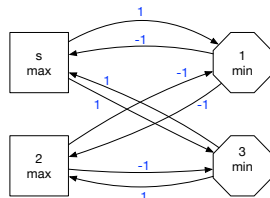
Max-atom are polynomial-time equivalent to system with  $n$  variables  $x_1, \dots, x_n$  and  $n$  constraints  $e_1, \dots, e_n$  where  $e_i$  is of the form

$$x_i \leq \max(x_{i_1} + c_1, \dots, x_{i_k} + c_k)$$
$$\text{or } x_i \leq \min(x_{i_1} + c_1, \dots, x_{i_k} + c_k).$$

## 'Max-min offset operator system'

For those systems, the correspondence is particularly nice:

$$x_s \leq \max(x_1 + 1, x_3 + 1)$$
$$x_2 \leq \max(x_1 - 1, x_3 - 1)$$
$$x_1 \leq \min(x_s - 1, x_2 - 1)$$
$$x_3 \leq \min(x_s + 1, x_2 + 1)$$



## Ideas of Correctness Proofs, 2

Have to prove:

**1** max wins on  $G$  iff  $\phi_G$  satisfiable.

**2**  $\phi$  satisfiable iff max wins on  $G_\phi$ .

**Idea:** suffices to prove correctness of 2!

Reason: reduction is such that  $G_{\phi_G} = G$ .

max wins on  $G$  iff max wins on  $G_{\phi_G}$   
iff  $\phi_G$  is satisfiable (by 2)

## Ideas of Correctness Proof, 3

Want to prove: A max-atoms instance  $\phi$  is satisfiable iff max wins on  $G_\phi$ .

**Idea:** To prove ' $\Rightarrow$ ', let  $s: V \rightarrow \mathbb{Q}$  be a solution to  $\phi$ .

For each constraint  $x \leq \max(y_1 + c_1, \dots, y_k + c_k)$

let  $i \leq k$  be such that  $y_i + c_i = \max(y_1 + c_1, \dots, y_k + c_k)$ .

**Claim:** the (memoryless!) strategy for max that always moves from  $x$  to  $y_i$  wins against any strategy of min.

**Idea:** To prove ' $\Leftarrow$ ', suppose that  $\phi$  is unsatisfiable.

Then  $\phi$  contains an unsatisfiable left-distinct subset  $\psi$ .

**Claim.**  $\psi$  corresponds to a (memoryless) strategy for min.

# Extensions?

## Questions:

- which extensions of the max-atom problem remain in  $NP \cap coNP$ ?
- which other relations over  $\mathbb{Q}$  are preserved by max and translations?
- which other relations over  $\mathbb{Q}$  are preserved by max?

# Semilinear Relations

## Definition .

$R \subseteq \mathbb{Q}^k$  is **semilinear** if  $R$  has a first-order definition in  $(\mathbb{Q}; +, 1, \leq)$ .

**Ferrante and Rackoff'75:** A relation is semilinear if and only if it is a finite intersection of finite unions of (open or closed) linear half spaces.

- $\text{CSP}(\mathbb{Q}; \leq, R_+, R_{=1})$  where  $R_+ := \{(x, y, z) \mid x = y + z\}$  and  $R_{=1} := \{1\}$  is essentially **linear program feasibility**.
- All CSPs with semi-linear constraints are in NP: guess one half space per union and verify satisfiability of resulting system of linear inequalities.
- Every finite-domain CSP falls into this class.

### **Big challenge:**

Classify the complexity of  $\text{CSP}(\mathbb{Q}; R_1, \dots, R_n)$  for all semilinear relations  $R_1, \dots, R_n$ .

# Tropically Convex Sets

**Definition (Develin-Sturmfels'04).**

A subset  $R$  of  $\mathbb{Q}^n$  is called **tropically convex** if

- $R$  is preserved by  $\max$ , and
  - $R$  is preserved by all *translations*  $x \mapsto x + c$  for all  $c \in \mathbb{Q}$ .
- 
- For every  $c \in \mathbb{Q}$  the max-atom relation  $R_c^{\max}$  is tropically convex
  - $x \leq (y + z)/2$  is tropically convex
  - $x \leq y + z$  is preserved by  $\max$ , but not tropically convex.

**Theorem (B+Mamino'15).**

Let  $S_1, \dots, S_n$  be tropically convex semilinear relations.

Then  $\text{CSP}(\mathbb{Q}; S_1, \dots, S_n)$  is in  $\text{NP} \cap \text{coNP}$ .

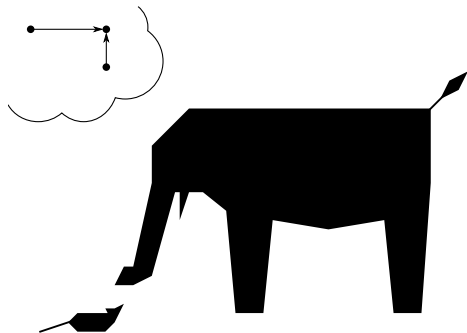
## Step 1: Syntax

Can we find a syntactic description of the relations preserved by max?

**Fact:** a relation  $R \subseteq \{0, 1\}^n$  is preserved by max **if and only if** the complement of  $R$  has a **Horn definition**.

e.g.  $(\neg x \vee \neg y \vee z) \wedge (\neg x \vee \neg u)$

For subsets of  $\mathbb{Q}^2$ , max-closed sets can be more complicated:





# Step 1: Syntax

## Theorem (B-Mamino'15).

Let  $R \subseteq \mathbb{Q}^n$  be semilinear and topologically closed. The following are equivalent:

- $R$  is preserved by max
- $R$  can be defined by a conjunction of expressions of the form

$$x \leq \max(a_1 \bar{x} + b_1, \dots, a_m \bar{x} + b_m)$$

where  $m \in \mathbb{N}$ , each of  $a_1, \dots, a_m$  is a vector from  $\mathbb{Q}_{\geq 0}^n$ , and  $b_1, \dots, b_m \in \mathbb{Q}$ ;

- $R$  has a primitive positive definition over  $(\mathbb{Q}; <, \{1\}, \{-1\}, S_1, S_2, S_3)$  where

$$S_1 := \{(x, y) : 2x \leq y\}$$

$$S_2 := \{(x, y, z) : x \leq y + z\}$$

$$S_3 := \{(x, y, z) : x \leq \max(y, z)\}$$

Theorem can be generalised to non-closed case.

# A syntactic description of tropical convexity

## Theorem (B-Mamino'15).

Let  $R \subseteq \mathbb{Q}^n$  be semilinear and topologically closed. The following are equivalent:

- $R$  is **tropically convex**.
- $R$  can be defined by a conjunction of expressions of the form

$$x \leq \max(a_1 \bar{x} + b_1, \dots, a_m \bar{x} + b_m)$$

where  $m \in \mathbb{N}$ ,  $a_1, \dots, a_m \in \mathbb{Q}_{\geq 0}^n$  are  $\sum_{i \leq n} a_{j,i} = 1$  for all  $j \leq m$ , and  $b_1, \dots, b_m \in \mathbb{Q}$ ;

- $R$  has a primitive positive definition in  $(\mathbb{Q}; \mathcal{S}_3, T_1, T_{-1}, \mathcal{S}_4)$  where

$$T_{\pm 1} := \{(x, y) : x \leq y \pm 1\}$$

$$\mathcal{S}_4 := \{(x, y, z) : x \leq (y + z)/2\}.$$

## Step 2: Duality for Max-Plus-Average Ineqs

Tropically convex CSPs can be reduced to:

$P$ : a system of  $n$  strict inequalities on  $n$  variables, each having one the following three forms

$$x_i < \max(x_{j_1} + k_1, \dots, x_{j_m} + k_m) \quad (1)$$

$$x_i < \min(x_{j_1} + k_1, \dots, x_{j_m} + k_m) \quad (2)$$

$$x_i < (\alpha_1 x_{j_1} + \dots + \alpha_m x_{j_m}) / (\alpha_1 + \dots + \alpha_m) + k \quad (3)$$

where  $\alpha_1, \dots, \alpha_m > 0$ .

The **dual**  $D$  of  $P$ : replace  $<$  by  $\geq$ .

**Theorem (B+Mamino'15).**

$P$  has a solution in  $\mathbb{Q}^n$  **if and only if**  $D$  has no solution in  $(\mathbb{Q} \cup \{+\infty\})^n \setminus \{+\infty\}^n$ .

**Consequence:** satisfiability of  $P$  is in  $\text{NP} \cap \text{coNP}$ .

**Proof:** Connection to stochastic MPG's.

# Link between CSPs and stochastic MPG

Let  $P$  be the system of  $n$  constraints on  $n$  variables, and  $D$  its dual.  
 $P$  corresponds to game:

- max inequalities  $\leftrightarrow$  max vertices.
- min inequalities  $\leftrightarrow$  min vertices.
- inequalities  $x_i < (\alpha_1 x_{j_1} + \dots + \alpha_m x_{j_m}) / (\alpha_1 + \dots + \alpha_m) + c$   
 $\leftrightarrow$  stochastic vertices with probabilities  $\alpha_1, \dots, \alpha_m$ .

$v(x)$ : **expected** limiting average payoff of game  $G$ , starting in  $x$ .

**Theorem (B+Mamino'15).**

- $D$  is satisfiable **if and only if**  $v(x_i) \leq 0$  for some vertex  $x_i$  of  $G$ .
- $P$  is satisfiable **if and only if**  $v(x_i) > 0$  for all vertices  $x_i$  of  $G$ .

This implies the duality theorem

# Overview State of the Art

