

# New Algorithms for Classes of Mean-Payoff Games

Anna Hauschild

Institute for Algorithms and Complexity

June 17, 2022

# Contents

- 1 Mean Payoff Games
  - Introduction
  - Central Idea
- 2 Algorithm
  - Implementation of the Functions
  - Example
- 3 Testing Results
  - Initialization
  - Testing Results

# Introduction to Mean Payoff Games

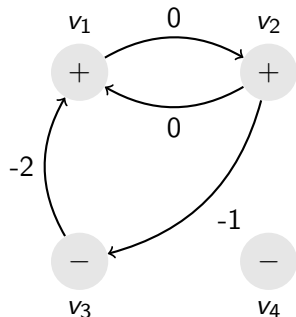


Figure: Graph  $G_1$

- Two players *Min* and *Max*
- vertex set  $V = V_+ \cup V_-$ 
  - $V_+ = \{v_1, v_2\}$
  - $V_- = \{v_3, v_4\}$

# Introduction to Mean Payoff Games

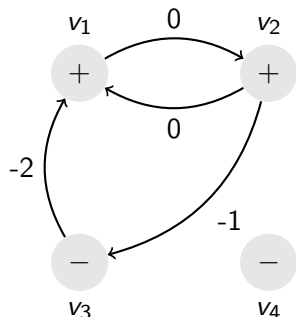


Figure: Graph  $G_1$

- Two players *Min* and *Max*
- Vertex set  $V = V_+ \cup V_-$ 
  - $V_+ = \{v_1, v_2\}$
  - $V_- = \{v_3, v_4\}$
- Edge set  $E = \{(v_1, v_2), (v_2, v_1), (v_2, v_3), (v_3, v_1)\}$
- Cost function  $\mu(v, w)$  i.e.  
 $\mu(v_2, v_3) = -1$

# Introduction to Mean Payoff Games

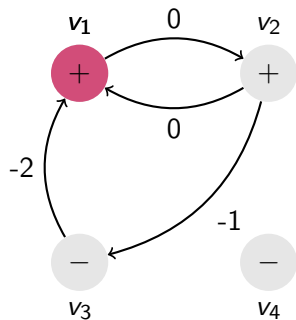


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1\}$

# Introduction to Mean Payoff Games

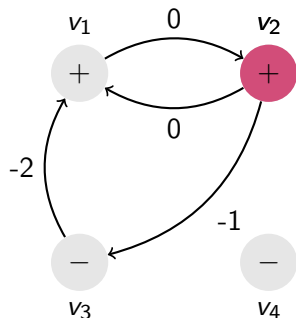


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1, v_2\}$

# Introduction to Mean Payoff Games

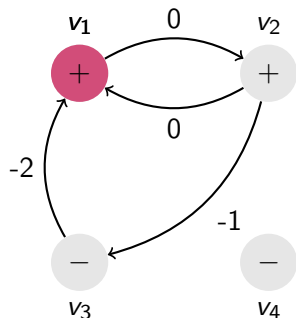


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1, v_2, v_1, \dots\}$

# Introduction to Mean Payoff Games

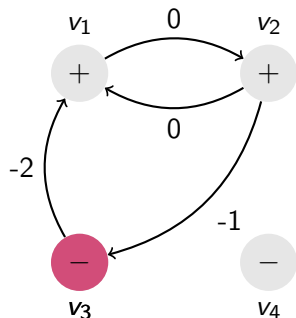


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1, v_2, v_1, \dots\}$
- Game  $\pi_2 = \{v_3\}$



# Introduction to Mean Payoff Games

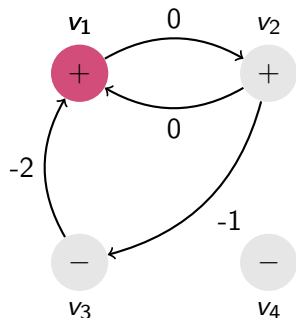


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1, v_2, v_1, \dots\}$
- Game  $\pi_2 = \{v_3, v_1\}$

# Introduction to Mean Payoff Games

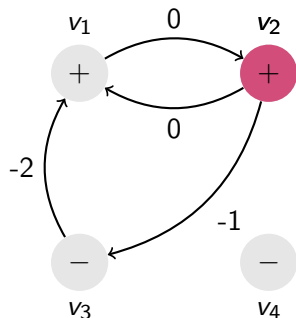


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1, v_2, v_1, \dots\}$
- Game  $\pi_2 = \{v_3, v_1, v_2\}$

# Introduction to Mean Payoff Games

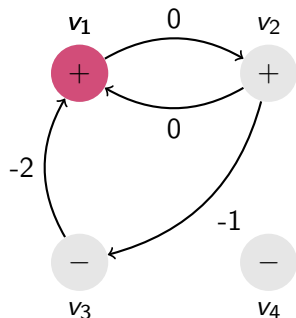


Figure: Graph  $G_1$

- Game  $\pi_1 = \{v_1, v_2, v_1, \dots\}$
- Game  $\pi_2 = \{v_3, v_1, v_2, v_1, \dots\}$

# Introduction to Mean Payoff Games

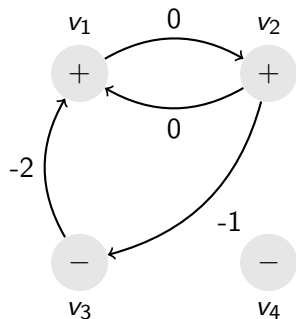


Figure: Graph  $G_1$

- *Characteristic value*

$$\chi(\pi) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \mu(v_i, v_{i+1})$$

- $\chi(\pi) > 0 \Rightarrow$  Max wins,  
 $\chi(\pi) < 0 \Rightarrow$  Min wins,  
 $\chi(\pi) = 0 \Rightarrow$  Draw

- Introduced by Ehrenfeucht and Mycielski<sup>a</sup>

---

<sup>a</sup>Ehrenfeucht and Mycielski, "Positional strategies for mean payoff games".

# Introduction to Mean Payoff Games

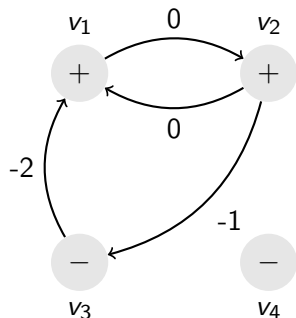


Figure: Graph  $G_1$

- *Sink* := vertex without outgoing edges
- $v_n \in V_+$  *Min* wins,  
 $v_n \in V_-$  *Max* wins

# Introduction to Mean Payoff Games

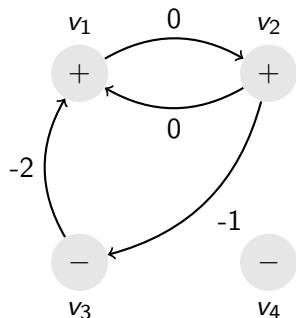


Figure: Graph  $G_1$

- Winning sets
  - $W_+$  for *Max*,
  - $W_-$  for *Min*,
  - $W_0$  for Draw
- $W_+ = \{v_4\}$
- $W_- = \emptyset$
- $W_0 = \{v_1, v_2, v_3\}$

# Central Idea

- Existing algorithm in  $\mathcal{O}(|V|^3 \cdot |E| \cdot W)$ <sup>1</sup>
- Instead of solving the whole graph solve small part of it
- Extend winning sets by applying a reachability game
- Especially games with a signature of a potential can be solved in linear time<sup>2</sup>

---

<sup>1</sup>Zwick and Paterson, “The complexity of mean payoff games on graphs”.

<sup>2</sup>Akian et al., “Mean payoff games with signature of a potential”.

- 1 Mean Payoff Games
  - Introduction
  - Central Idea
- 2 Algorithm
  - Implementation of the Functions
  - Example
- 3 Testing Results
  - Initialization
  - Testing Results



# Functions<sup>3</sup>

```

function MAIN( $V_+$ ,  $V_-$ ,  $E$ ,  $\mu$ )
   $F := \text{Obsolete}(\#V_+ \# + V_-, V_+, V_-, E, \mu)$ 
  while  $F \neq \emptyset$  do
     $E := E - F$ 
     $F := \text{Obsolete}(\#V_+ + V_-, V_+, V_-, E, \mu)$ 
  end while
   $W_+ = W_- = W_0 := \emptyset$ 
  if  $V = \emptyset$  then
     $r := 0$ 
  else
     $r := 1$ 
  end if
   $X_1 := V$ 
  while  $r > 0$  do
    ...
  end while
  return ( $W_+$ ,  $W_-$ ,  $W_0$ )
end function

```

- 1 Remove obsolete edges
- 2 Decompose graph into components
- 3 Solve one component
- 4 Check reachability for vertices outside the winning sets

<sup>3</sup>Akian et al., “Mean payoff games with signature of a potential”.

# Functions<sup>3</sup>

```

function MAIN( $V_+$ ,  $V_-$ ,  $E$ ,  $\mu$ )
...
while  $r > 0$  do
   $X := X_r - (W_+ \cup W_- \cup W_0)$ 
   $(s, X_r, \dots, X_{r+s-1}) := \text{StrConnComp}(X, E)$ 
   $r := r + s - 1$ 
   $(Y_+, Y_-, Y_0) := \text{SolveMPG}(X_r, V_+, V_-, E, \mu)$ 
   $W_0 := W_0 \cup Y_0$ 
  for  $e \in \{+, -\}$  do
     $F := \{(v, w) \in E \mid v \in Y_e, w \in Y_e \cup W_0\}$ 
     $W_0 := \text{Reach}(W_0, -e, V_+, V_-, F)$ 
    if  $Y_e \cap W_0 = \emptyset$  then
       $W_e := W_e \cup Y_e$ 
    end if
  end for
   $W_+ := \text{Reach}(W_+, +, V_+, V_-, E)$ 
   $W_- := \text{Reach}(W_-, -, V_+, V_-, E)$ 
  while  $r > 0$  and  $X_r \subset W_+ \cup W_- \cup W_0$  do
     $r := r - 1$ 
  end while
end while
return  $(W_+, W_-, W_0)$ 
end function

```

- ① Remove obsolete edges
- ② Decompose graph into components
- ③ Solve one component
- ④ Check reachability for vertices outside the winning sets

<sup>3</sup>Akian et al., “Mean payoff games with signature of a potential”.

# Functions<sup>3</sup>

```

function MAIN( $V_+, V_-, E, \mu$ )
  ...
  while  $r > 0$  do
     $X := X_r - (W_+ \cup W_- \cup W_0)$ 
     $(s, X_r, \dots, X_{r+s-1}) := \text{StrConnComp}(X, E)$ 
     $r := r + s - 1$ 
     $(Y_+, Y_-, Y_0) := \text{SolveMPG}(X_r, V_+, V_-, E, \mu)$ 
     $W_0 := W_0 \cup Y_0$ 
    for  $e \in \{+, -\}$  do
       $F := \{(v, w) \in E \mid v \in Y_e, w \in Y_e \cup W_0\}$ 
       $W_0 := \text{Reach}(W_0, -e, V_+, V_-, F)$ 
      if  $Y_e \cap W_0 = \emptyset$  then
         $W_e := W_e \cup Y_e$ 
      end if
    end for
     $W_+ := \text{Reach}(W_+, +, V_+, V_-, E)$ 
     $W_- := \text{Reach}(W_-, -, V_+, V_-, E)$ 
    while  $r > 0$  and  $X_r \subset W_+ \cup W_- \cup W_0$  do
       $r := r - 1$ 
    end while
  end while
  return  $(W_+, W_-, W_0)$ 
end function

```

- 1 Remove obsolete edges
- 2 Decompose graph into components
- 3 Solve one component
- 4 Check reachability for vertices outside the winning sets

<sup>3</sup>Akian et al., “Mean payoff games with signature of a potential”.

# Functions<sup>3</sup>

```

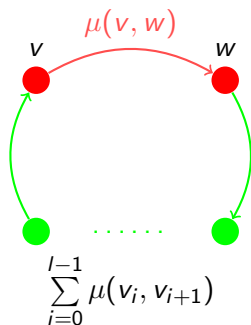
function MAIN( $V_+, V_-, E, \mu$ )
  ...
  while  $r > 0$  do
     $X := X_r - (W_+ \cup W_- \cup W_0)$ 
     $(s, X_r, \dots, X_{r+s-1}) := \text{StrConnComp}(X, E)$ 
     $r := r + s - 1$ 
     $(Y_+, Y_-, Y_0) := \text{SolveMPG}(X_r, V_+, V_-, E, \mu)$ 
     $W_0 := W_0 \cup Y_0$ 
    for  $e \in \{+, -\}$  do
       $F := \{(v, w) \in E \mid v \in Y_e, w \in Y_e \cup W_0\}$ 
       $W_0 := \text{Reach}(W_0, -e, V_+, V_-, F)$ 
      if  $Y_e \cap W_0 = \emptyset$  then
         $W_e := W_e \cup Y_e$ 
      end if
    end for
     $W_+ := \text{Reach}(W_+, +, V_+, V_-, E)$ 
     $W_- := \text{Reach}(W_-, -, V_+, V_-, E)$ 
    while  $r > 0$  and  $X_r \subset W_+ \cup W_- \cup W_0$  do
       $r := r - 1$ 
    end while
  end while
  return  $(W_+, W_-, W_0)$ 
end function

```

- ① Remove obsolete edges
- ② Decompose graph into components
- ③ Solve one component
- ④ Check reachability for vertices outside the winning sets

<sup>3</sup>Akian et al., “Mean payoff games with signature of a potential”.

# Obsolete edges



$$e(v) \sum_{i=0}^{l-1} \mu(v_i, v_{i+1}) < -e(v) \mu(v, w).$$

Figure: Obsolete edges

# Obsolete edges

```

1: function OBSOLETE( $k, V_+, V_-, E, \mu$ )
2:    $F = \emptyset$ 
3:   for  $e \in \{+, -\}$  and  $v \in V_e$  do
4:     for  $i = 0, \dots, k - 1$  do  $\eta_i(v) = 0$ 
5:   end for
6:   for  $w \in (V_+ \cup V_-) - \{v\}$  do  $\eta_0(w) := e \cdot \infty$ 
7:   end for
8:   for  $i = 1, \dots, k - 1$  do
9:     for  $w \in (V_+ \cup V_-) - \{v\}$  do
10:       $\eta_i(w) := \epsilon(w) \cdot \max\{\epsilon(w) \cdot (\mu(w, u) + \eta_{i-1}(u)) \mid (w, u) \in E\}$ 
11:    end for
12:  end for
13:  for  $(v, w) \in E$  do
14:    if  $e \cdot \eta_{k-1}(w) < -e \cdot \mu(v, w)$  then  $F := F \cup \{(v, w)\}$ 
15:    end if
16:  end for
17: end for
18: return  $F$ 
19: end function

```

# Obsolete edges

$$\eta_i(w) := \epsilon(w) \cdot \max\{\epsilon(w) \cdot (\mu(w, u) + \eta_{i-1}(u)) \mid (w, u) \in E\}$$

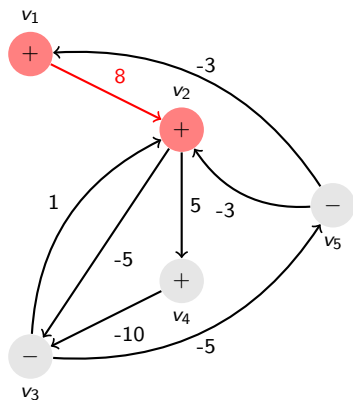


Figure: Graph  $G_3$

$$\eta(v_1) = \begin{matrix} i = 0 & \begin{matrix} v_1 & v_2 & v_3 & v_4 & v_5 \end{matrix} \\ \begin{pmatrix} 0 & \infty & \infty & \infty & \infty \\ 0 & & & & \\ 0 & & & & \\ 0 & & & & \\ 0 & & & & \end{pmatrix} \end{matrix}$$

# Obsolete edges

$$\eta_i(w) := \epsilon(w) \cdot \max\{\epsilon(w) \cdot (\mu(w, u) + \eta_{i-1}(u)) \mid (w, u) \in E\}$$

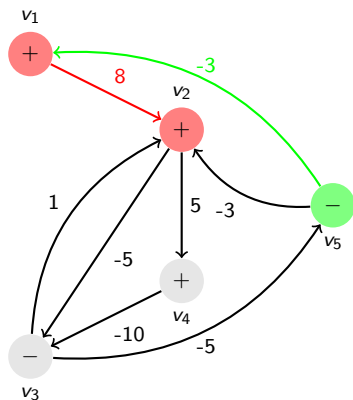


Figure: Graph  $G_3$

$$\eta(v_1) = \begin{matrix} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \begin{matrix} i=0 \\ i=1 \end{matrix} & \begin{pmatrix} 0 & \infty & \infty & \infty & \infty \\ 0 & \infty & \infty & \infty & -3 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \end{matrix}$$



# Obsolete edges

$$\eta_i(w) := \epsilon(w) \cdot \max\{\epsilon(w) \cdot (\mu(w, u) + \eta_{i-1}(u)) \mid (w, u) \in E\}$$

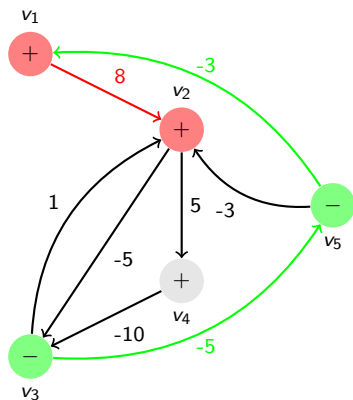


Figure: Graph  $G_3$

$$\eta(v_1) = \begin{matrix} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \begin{matrix} i=0 \\ i=1 \\ i=2 \end{matrix} & \begin{pmatrix} 0 & \infty & \infty & \infty & \infty \\ 0 & \infty & \infty & \infty & -3 \\ 0 & \infty & -8 & \infty & -3 \end{pmatrix} \end{matrix}$$

# Obsolete edges

$$\eta_i(w) := \epsilon(w) \cdot \max\{\epsilon(w) \cdot (\mu(w, u) + \eta_{i-1}(u)) \mid (w, u) \in E\}$$

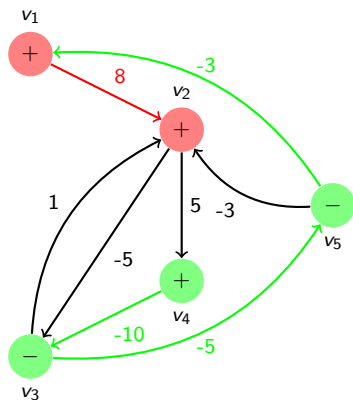


Figure: Graph  $G_3$

$$\eta(v_1) = \begin{matrix} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \begin{matrix} i=0 \\ i=1 \\ i=2 \\ i=3 \end{matrix} & \begin{pmatrix} 0 & \infty & \infty & \infty & \infty \\ 0 & \infty & \infty & \infty & -3 \\ 0 & \infty & -8 & \infty & -3 \\ 0 & \infty & -8 & -18 & -3 \end{pmatrix} \end{matrix}$$

# Obsolete edges

$$\eta_i(w) := \epsilon(w) \cdot \max\{\epsilon(w) \cdot (\mu(w, u) + \eta_{i-1}(u)) \mid (w, u) \in E\}$$

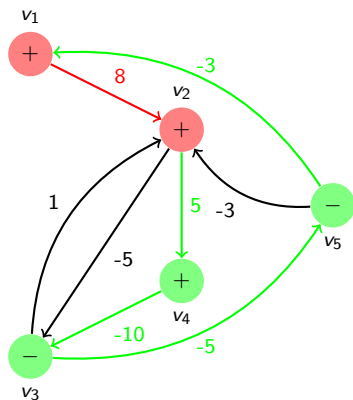
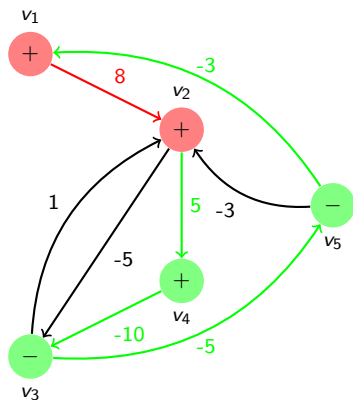


Figure: Graph  $G_3$

$$\eta(v_1) = \begin{matrix} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \begin{matrix} i=0 \\ i=1 \\ i=2 \\ i=3 \\ i=4 \end{matrix} & \begin{pmatrix} 0 & \infty & \infty & \infty & \infty \\ 0 & \infty & \infty & \infty & -3 \\ 0 & \infty & -8 & \infty & -3 \\ 0 & \infty & -8 & -18 & -3 \\ 0 & -13 & -8 & -18 & -3 \end{pmatrix} \end{matrix}$$

# Obsolete edges

**If**  $e \cdot \eta_{k-1}(w) < -e \cdot \mu(v, w)$  **then**  $F := F \cup \{(v, w)\}$



$$\eta(v_1) = \begin{matrix} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \begin{matrix} i=0 \\ i=1 \\ i=2 \\ i=3 \\ i=4 \end{matrix} & \begin{pmatrix} 0 & \infty & \infty & \infty & \infty \\ 0 & \infty & \infty & \infty & -3 \\ 0 & \infty & -8 & \infty & -3 \\ 0 & \infty & -8 & -18 & -3 \\ 0 & -13 & -8 & -18 & -3 \end{pmatrix} \end{matrix}$$

$$1 \cdot \eta_4(v_2) < (-1) \cdot \mu(v_1, v_2)$$

$$1 \cdot (-13) < (-1) \cdot 8$$

Figure: Graph  $G_3$

# Obsolete edges

$$F = \{(v_1, v_2), (v_2, v_3), (v_2, v_4), (v_4, v_3)\}$$

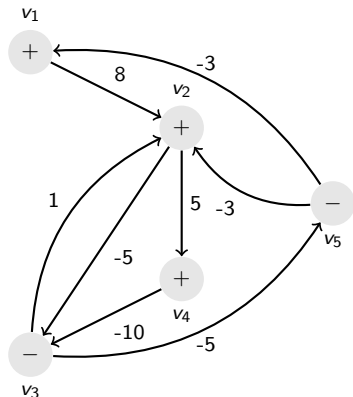


Figure: Graph  $G_3$

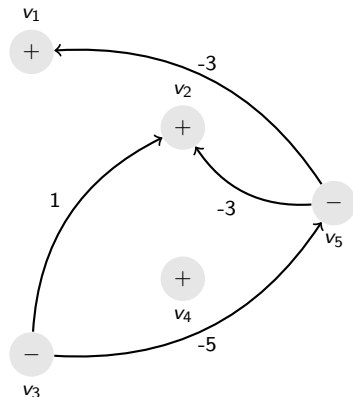


Figure: Graph  $G'_3$  after removing obsolete edges

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$

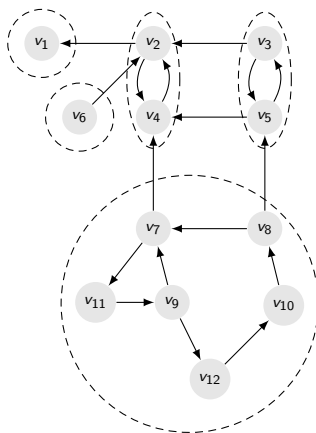
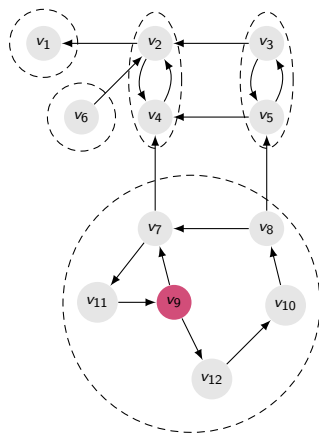


Figure:  $G_4$  decomposed into it's strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9]$$

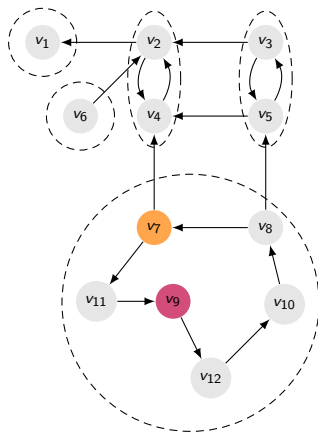
$$m = [1]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7]$$

$$m = [1, 2]$$

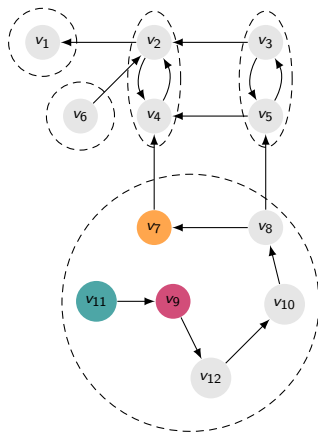
Figure:  $G_4$  decomposed into its strongly connected components



# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}]$$

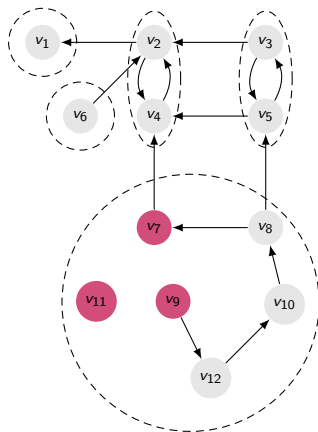
$$m = [1, 2, 3]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}]$$

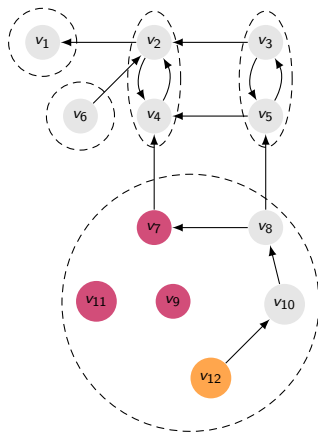
$$m = [1, 1, 1]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}]$$

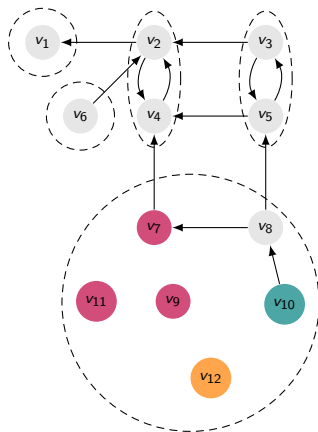
$$m = [1, 1, 1, 2]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}]$$

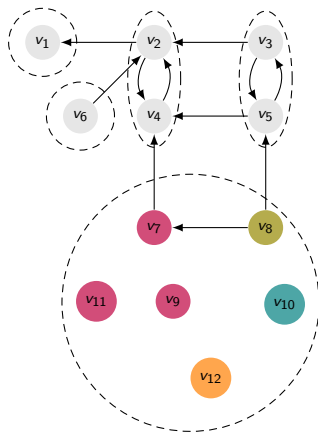
$$m = [1, 1, 1, 2, 3]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8]$$

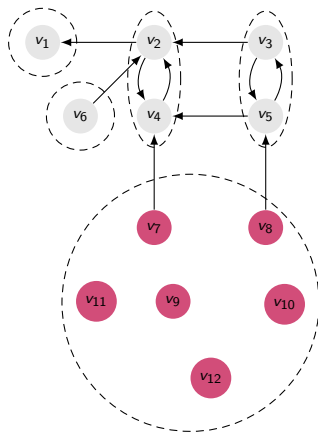
$$m = [1, 1, 1, 2, 3, 4]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8]$$

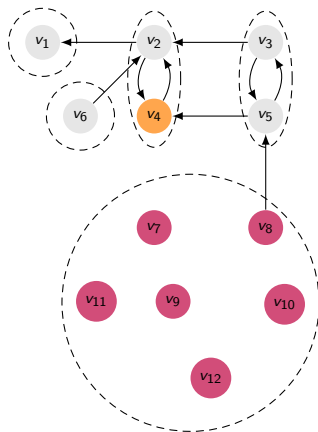
$$m = [1, 1, 1, 1, 1, 1]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_4]$$

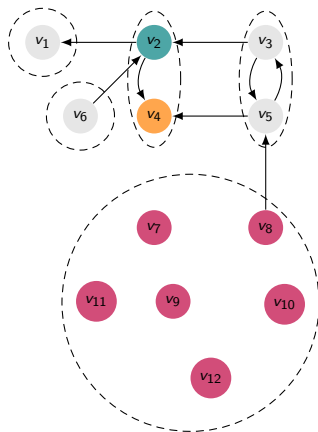
$$m = [1, 1, 1, 1, 1, 1, 2]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_4, v_2]$$

$$m = [1, 1, 1, 1, 1, 1, 2, 3]$$

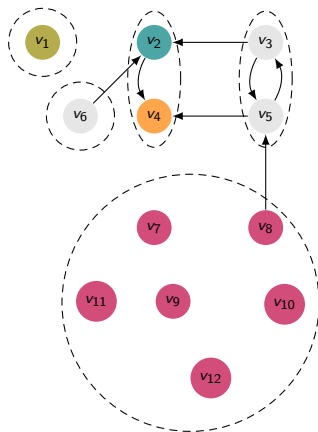
Figure:  $G_4$  decomposed into its strongly connected components



# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_4, v_2, v_1]$$

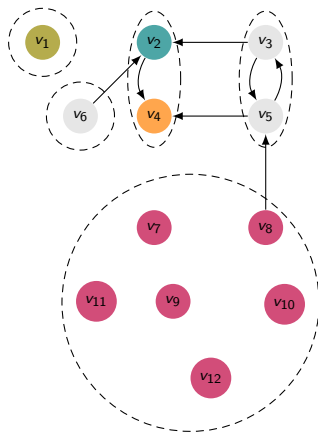
$$m = [1, 1, 1, 1, 1, 1, 2, 3, 4]$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_4, v_2]$

$m = [1, 1, 1, 1, 1, 1, 2, 3]$

No outgoing edge  $\Rightarrow$  First component found

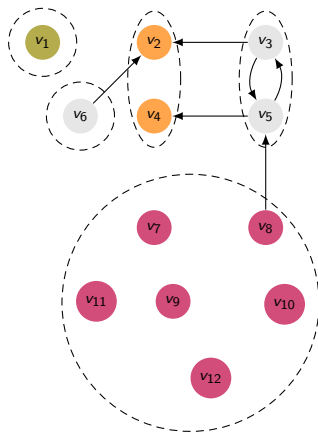
$Y_1 = \{v_1\}$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_4, v_2]$$

$$m = [1, 1, 1, 1, 1, 1, 2, 2]$$

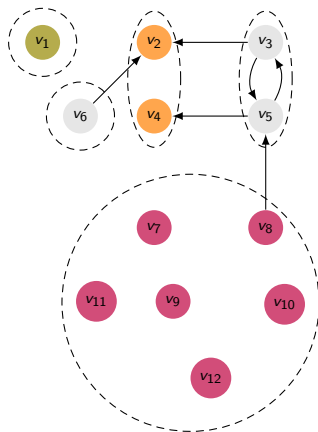
$$Y_1 = \{v_1\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8]$$

$$m = [1, 1, 1, 1, 1, 1]$$

No outgoing edge for component  
 $\{v_2, v_4\} \Rightarrow$  Second component found

$$Y_1 = \{v_1\}$$

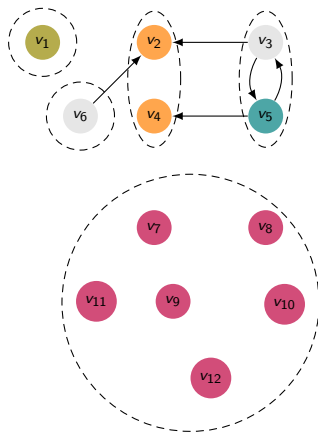
$$Y_2 = \{v_2, v_4\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_5]$$

$$m = [1, 1, 1, 1, 1, 1, 2]$$

$$Y_1 = \{v_1\}$$

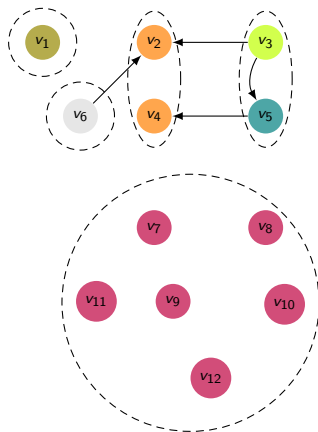
$$Y_2 = \{v_2, v_4\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_5, v_3]$$

$$m = [1, 1, 1, 1, 1, 1, 2, 3]$$

$$Y_1 = \{v_1\}$$

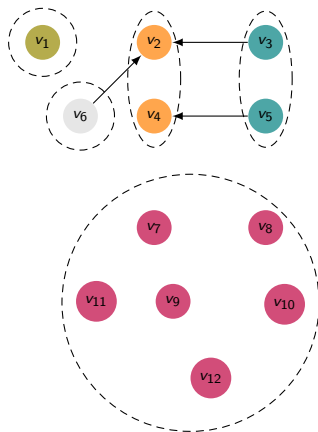
$$Y_2 = \{v_2, v_4\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8, v_5, v_3]$$

$$m = [1, 1, 1, 1, 1, 1, 2, 2]$$

$$Y_1 = \{v_1\}$$

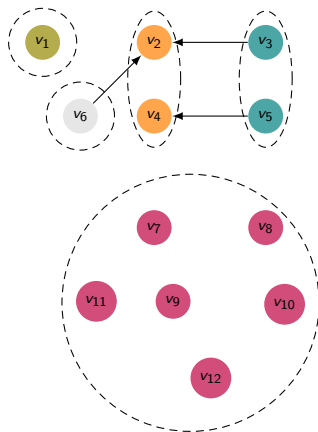
$$Y_2 = \{v_2, v_4\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$v = [v_9, v_7, v_{11}, v_{12}, v_{10}, v_8]$

$m = [1, 1, 1, 1, 1, 1]$

Edges lead to already allocated nodes  
 $\Rightarrow$  Third component found

$Y_1 = \{v_1\}$

$Y_2 = \{v_2, v_4\}$

$Y_3 = \{v_3, v_5\}$

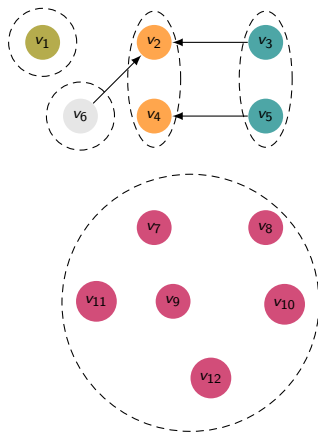
Figure:  $G_4$  decomposed into its strongly connected components



# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$v = []$

$m = []$

Edges lead to already allocated nodes  
 $\Rightarrow$  Fourth component found

$Y_1 = \{v_1\}$

$Y_2 = \{v_2, v_4\}$

$Y_3 = \{v_3, v_5\}$

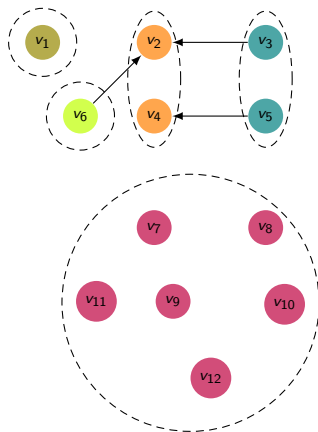
$Y_4 = \{v_9, v_7, v_{11}, v_{12}, v_{10}, v_8\}$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



$$v = [v_6]$$

$$m = [1]$$

$$Y_1 = \{v_1\}$$

$$Y_2 = \{v_2, v_4\}$$

$$Y_3 = \{v_3, v_5\}$$

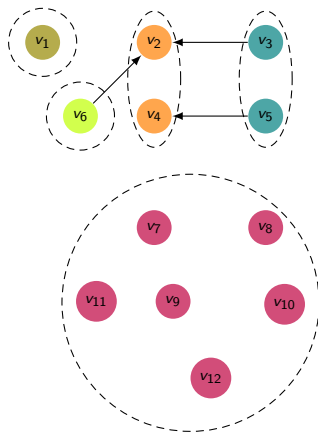
$$Y_4 = \{v_6, v_7, v_8, v_9, v_{10}, v_{11}, v_{12}\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Strongly connected components

**Input:**  $X, E$

**Output:**  $s, Y_s, \dots, Y_1$



No other vertices left  
 $\Rightarrow$  Fifth component found

$$Y_1 = \{v_1\}$$

$$Y_2 = \{v_2, v_4\}$$

$$Y_3 = \{v_3, v_5\}$$

$$Y_4 = \{v_9, v_7, v_{11}, v_{12}, v_{10}, v_8\}$$

$$Y_5 = \{v_6\}$$

Figure:  $G_4$  decomposed into its strongly connected components

# Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

Two ways of ending a game:

- **Finite** game: ends in a sink
- **Infinite** game: ends in a loop

# Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

- **Finite** game: ends in a sink

**Function** SolveMPG( $X, V_+, V_-, E, \mu$ ) **is**

$Y_+ = Y_- = Y_0 := \emptyset$

$C := \{(v, w) \in E \mid v, w \in X\}$

**if**  $C = \emptyset$  **and**  $X \subset V_+$  **then**  $Y_- := X$

**if**  $C = \emptyset$  **and**  $X \subset V_-$  **then**  $Y_+ := X$

Figure: pseudo code: finite game

# Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

- **Infinite** game: ends in a loop

# Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

- **Infinite** game: ends in a loop
  - Optimal positional strategy
  - Characteristic value

## Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

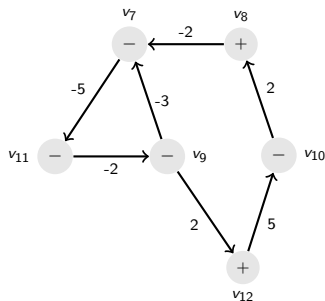


Figure: Subgraph of graph  $G_4$



## Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

Characteristic value  $\chi(v) < 0 \quad \forall v \in \{v_7, \dots, v_{12}\}$

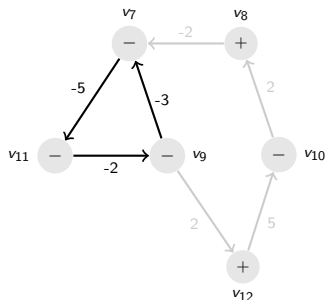


Figure: Subgraph of graph  $G_4$

## Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

Characteristic value  $\chi(v) = 0 \quad \forall v \in \{v_7, \dots, v_{12}\}$

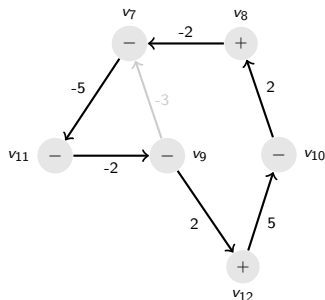


Figure: Subgraph of graph  $G_4$

## Solving subgraphs

**Input:**  $X, V_+, V_-, E, \mu$

**Output:**  $Y_+, Y_-, Y_0$

$$Y_- = \{v_7, v_8, v_9, v_{10}, v_{11}, v_{12}\}, Y_+ = \emptyset, Y_0 = \emptyset$$

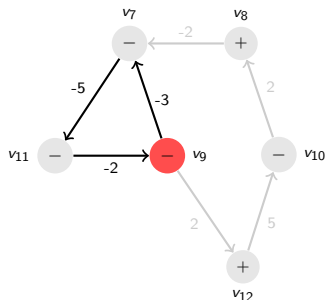


Figure: Subgraph of graph  $G_4$

# Reachability

```
function REACH( $X, e, V_+, V_-, E$ )  
2:   for  $(v, w) \in E$  with  $v \notin X$  and  $w \in X$  do  
      if  $v \in V_e$  or  $u \in X$  for all  $(v, u) \in E$  then  
4:        $X := X \cup \{v\}$   
      end if  
6:   end for  
      return  $X$   
8: end function
```

# Reachability

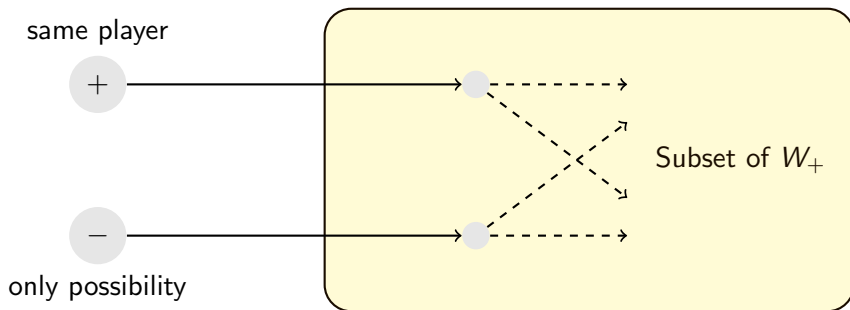


Figure: Finding additional winning vertices of *Max* via reachability

# Reachability

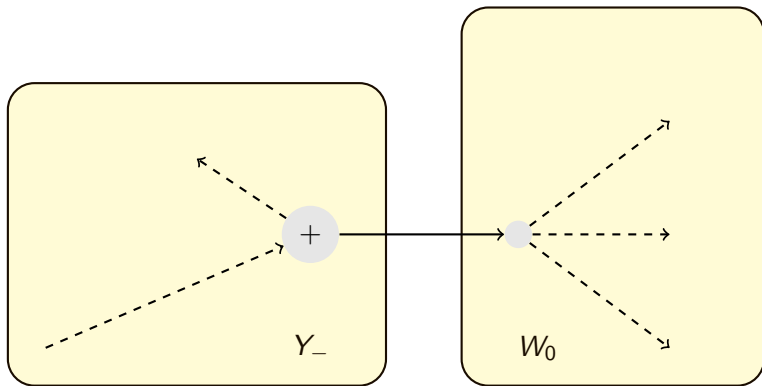


Figure: Additional draw positions for *Max*

# Example

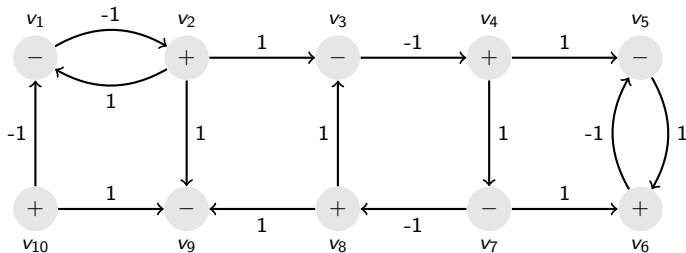


Figure: Graph  $G_5$

# Example

- $(s, X_r, \dots, X_{r+s-1}) := \text{StrConnComp}(X, E)$

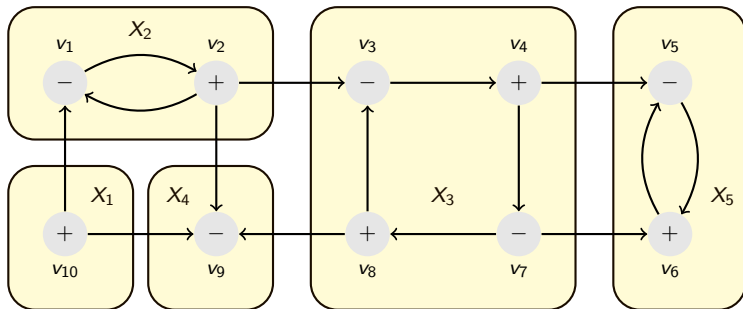
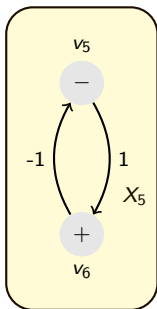


Figure: Strongly connected components



# Example

- $(Y_+, Y_-, Y_0) := \text{SolveMPG}(X_r, V_+, V_-, E, \mu)$



$$Y_+ = Y_- = \emptyset, \quad Y_0 = \{v_5, v_6\}$$

Figure: Subgraph of  $G_5$

# Example

- $W_0 := W_0 \cup Y_0 = \{v_5, v_6\}$

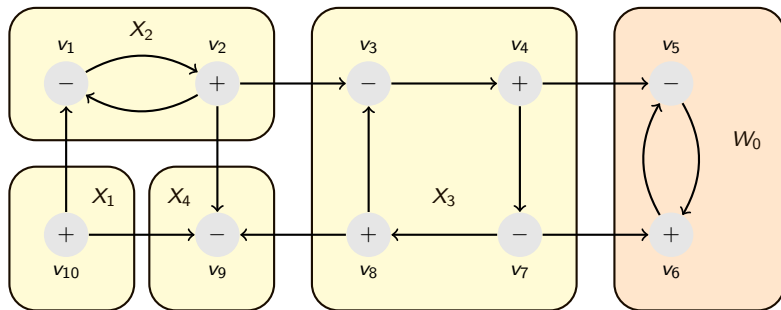


Figure:  $v_5$  and  $v_6$  assigned to  $W_0$

- $(Y_+, Y_-, Y_0) := \text{SolveMPG}(X_r, V_+, V_-, E, \mu)$

# Example

- $W_+ := \text{Reach}(W_+, +, V_+, V_-, E)$

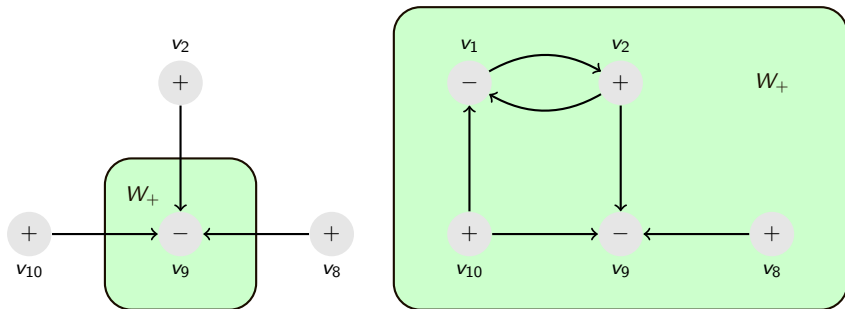


Figure: Reachability game

# Example

- $(s, X_r, \dots, X_{r+s-1}) = (3, X_3, X_4, X_5) := \text{StrConnComp}(X, E)$

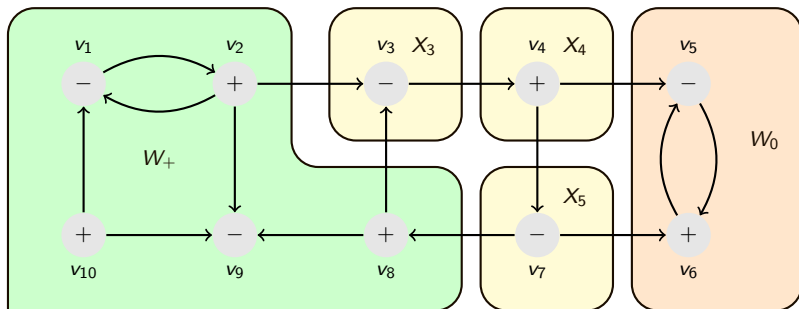


Figure: Third run of *StrConnComp*

# Example

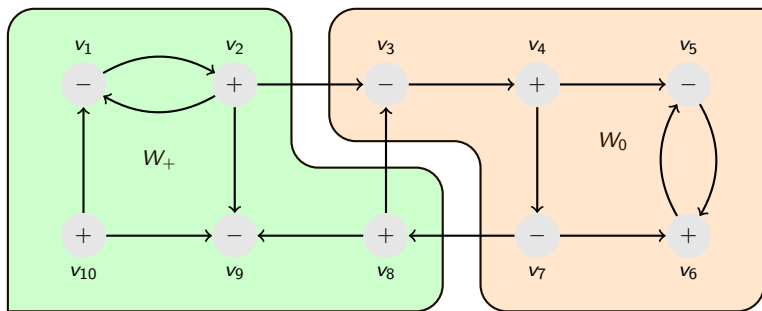


Figure: Resulting winning sets  $W_+$  and  $W_0$

- 1 Mean Payoff Games
  - Introduction
  - Central Idea
- 2 Algorithm
  - Implementation of the Functions
  - Example
- 3 Testing Results
  - Initialization
  - Testing Results

# Initialization

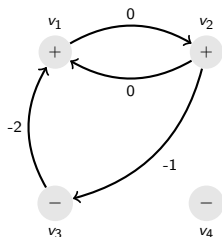


Figure: Graph  $G_1$

$$E = \begin{matrix} & v_1 & v_2 & v_3 & v_4 \\ \begin{matrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{matrix} & \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \end{matrix}$$

$$\mu = \begin{matrix} & v_1 & v_2 & v_3 & v_4 \\ \begin{matrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{matrix} & \begin{pmatrix} -\infty & 0 & -\infty & -\infty \\ 0 & -\infty & -1 & -\infty \\ -2 & \infty & \infty & \infty \\ \infty & \infty & \infty & \infty \end{pmatrix} \end{matrix}$$

$$V_+ = \begin{matrix} & v_1 & v_2 & v_3 & v_4 \\ & (1 & 1 & 0 & 0) \end{matrix}$$

$$V_- = \begin{matrix} & v_1 & v_2 & v_3 & v_4 \\ & (0 & 0 & 1 & 1) \end{matrix}$$

# Graph generator

- Program to create a random graph<sup>4</sup>
- The number of vertices and *degrees* can be specified or arise randomly
- The graph is stored in a text file
- Option to check and create a signature of a potential

---

<sup>4</sup>Manish Bhojasia. access on 27.05.22. URL: <https://www.sanfoundry.com/>



# Experiment 1

- Simultaneously increasing number of nodes and max. degree

#Vertices	max degree	Time[s]
10	10	0.37
50	50	0.51
100	100	2.00
200	200	43.11
500	500	2047.86

Table: Simultaneously increasing vertices and degrees

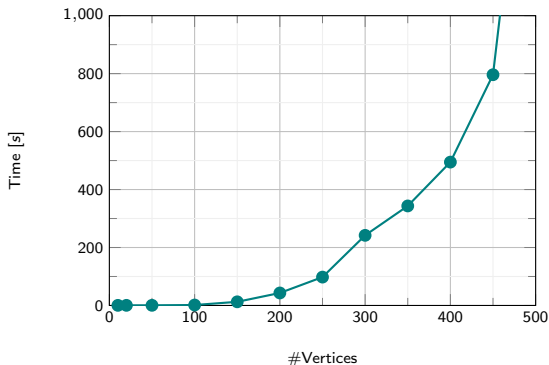


Figure: Simultaneously increasing vertices and degrees

# The degree

- Decreasing degree while number of vertices is fixed to 100

max degree	#Components
100	2
50	2
30	3
20	6
10	17
5	46
2	77
1	83
0	100

Table: #Components in relation to degree in a graph with 100 vertices

## Experiment 2

- Repeat first experiment with a lower maximal degree of 5

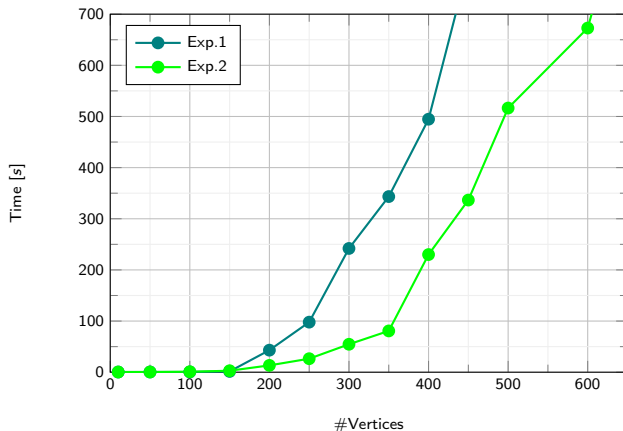


Figure: Comparison arbitrary degree to constant maximal degree 5

# Conclusion

- The function *Obsolete* that should normally improve the algorithm claims most of the time
- Already a slightly larger degree leads to a graph with very few components which contradicts the idea of the algorithm
- Observation: the slightly larger degree leads to the result that all nodes belong to the same winning set



Akian, Mariane et al. "Mean payoff games with signature of a potential". 2021.



Bhojasia, Manish. access on 27.05.22. URL:  
<https://www.sanfoundry.com/>.



Ehrenfeucht, A. and J. Mycielski. "Positional strategies for mean payoff games". In: *International Journal of Game Theory* 8.2 (1979), pp. 109–113. ISSN: 1432-1270. DOI: 10.1007/BF01768705.



Zwick, Uri and Mike Paterson. "The complexity of mean payoff games on graphs". In: *Theoretical Computer Science* 158.1 (1996), pp. 343–359. ISSN: 0304-3975. DOI:  
[https://doi.org/10.1016/0304-3975\(95\)00188-3](https://doi.org/10.1016/0304-3975(95)00188-3). URL:  
<https://www.sciencedirect.com/science/article/pii/0304397595001883>.