

# Parity Games

Sven Schewe

University of Liverpool

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# Beautiful games you cannot stop playing

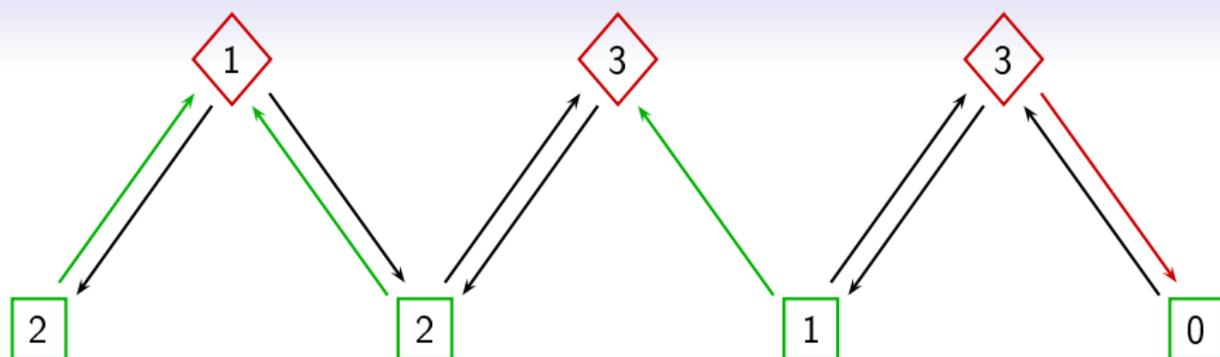
- 1 Parity Games with Few Colours
- 2 Parity Games with Many Colours
- 3 Parity Games with Few Colours
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- 6 Parity Games with Bounded Treewidth
- 7 Strategy Improvement Algorithms

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Parity Game  $\mathcal{P} = \langle V_0, V_1, E, \alpha \rangle$

- $V_0$ , and  $V_1$  are disjoint finite sets of game positions
- $E \subseteq V_0 \cup V_1 \times V_0 \cup V_1$  is a set of edges, and
- $\alpha : V_0 \cup V_1 \rightarrow \mathbb{N}$  is a colouring function

Played by placing a pebble on the arena

- on  $V_0$  player 0 chooses a successor, on  $V_1$  player 1
- $\Rightarrow$  infinite play, highest colour occurring infinite often
- even  $\rightsquigarrow$  player 0 wins, odd  $\rightsquigarrow$  player 1 wins

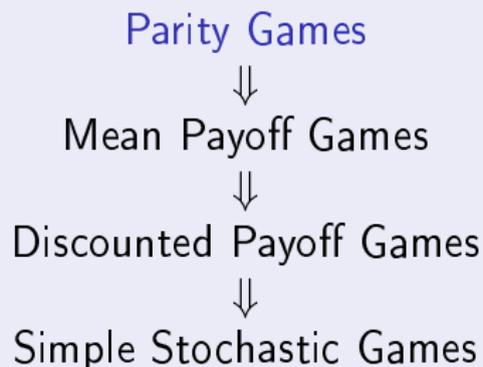
# Applications

- (non)emptiness game for parity tree automata
- acceptance game for parity tree automata
- satisfiability checking for CTL\*, ATL\*,  $\mu$ -calculus, AT $\mu$ C ...
- open synthesis for LTL, CTL\*, ATL\*,  $\mu$ -calculus, AT $\mu$ C ...
  
- $\mu$ -calculus model checking & extensions  
(e.g., graded  $\mu$ -calculus, alternating-time  $\mu$ -calculus)
- CTL\* model checking (three colours), ATL\* model checking
- module checking

# Simple & Symmetric

## Simple Reduction

[Zwick+Paterson 96]



## Symmetric Problem

Until recently, only a single deterministic symmetric algorithm  
Fixed Point, [Zwick+Paterson 96]

# Obvious Facts and Open Questions

## Obvious Facts

- symmetric

⇒ in class  $\cap$  co-class

- **single** fixed point of DPG can be guessed

⇒ in UP  $\cap$  co-UP

[Jurdziński 00]

## Less Obvious Facts

- PLS

[Beckmann and Moller 08]

- $n^{O(\sqrt{n})}$

[Jurdziński, Zwick, and Paterson 08]

- PPAD

[Etessami and Yannakakis 10]

# Obvious Facts and Open Questions

## Obvious Facts

- symmetric
- ⇒ in class  $\cap$  co-class
- **single** fixed point of DPG can be guessed
- ⇒ in UP  $\cap$  co-UP

[Jurdziński 00]

## Open Problems

- **P?**
- RP / ZPP?
- pay-off games:  $2^{O(\sqrt{n})}$ ?,  $2^{\sigma(n)}$ ?

# Overview

- Reachability Games
- Büchi Games
- Parity Games
  - McNaughton
  - Jurdziński, Paterson, and Zwick
  - Browne & al. / Jurdziński
  - their synthesis
- bounded tree-width & Co
- strategy improvement

few colours

# Part I

## Reachability & Büchi Games

## Solving Reachability Games



Algorithm – for  $\mathcal{R} = \langle V_0, V_1, E, F \rangle$

- start with the final states  $F$
- set  $W_{\diamond}$  to  $\diamond$ -attractor( $F$ )
- set  $W_{\square}$  to  $V \setminus W_{\diamond}$

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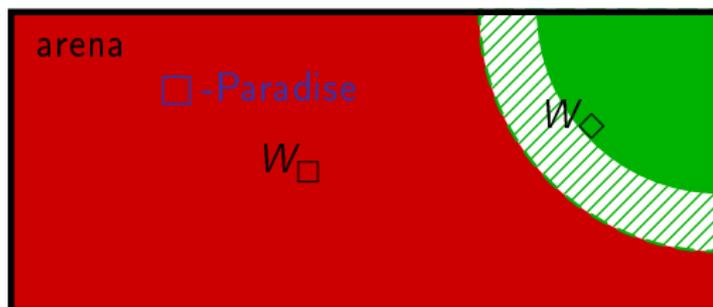
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# Traps and Paradises



## Traps and Paradises

- A  $\diamond$ -trap is a set of states where  $\diamond$  cannot get out.  
E.g.:  $W_{\square}$
- Remark:  $W_{\diamond} = W_{\diamond}^{\infty}$  is usually no  $\square$ -trap.
- A  $\square$ -paradise is a  $\diamond$ -trap such that  $\square$  can win without leaving it

Example:  $W_{\square}$

## Solving Büchi Games



Algorithm – for  $\mathcal{B} = \langle V_0, V_1, E, F \rangle$

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- set  $A$  to  $\diamond$ -attractor( $F$ )
- $U_{\square} = V \setminus A$  is a  $\square$ -paradise (strategy: stay there)
- $V_{\square} = \square$ -attractor( $U_{\square}$ ) is a  $\square$ -paradise (go to  $U_{\square}$ , stay)
- $W_{\diamond}$  for  $\mathcal{B}$  is  $W_{\diamond}$  for  $\mathcal{B} \setminus V_{\square}$
- solve  $\mathcal{B} \setminus V_{\square}$

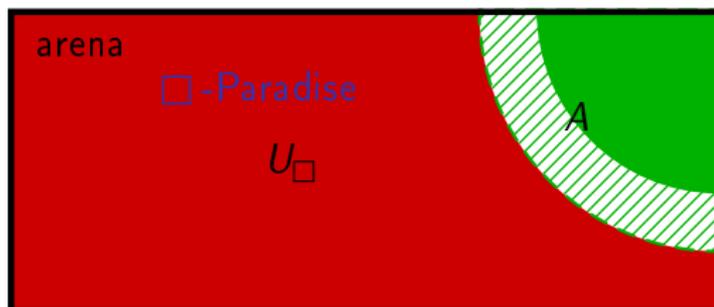
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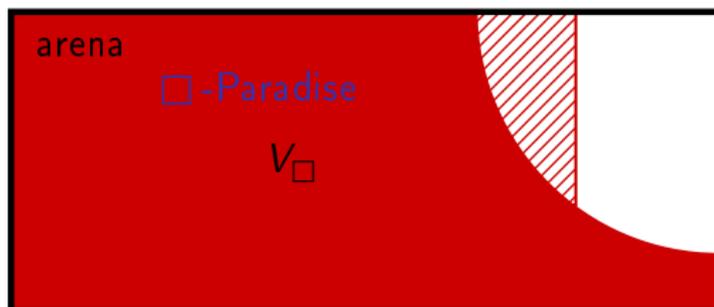
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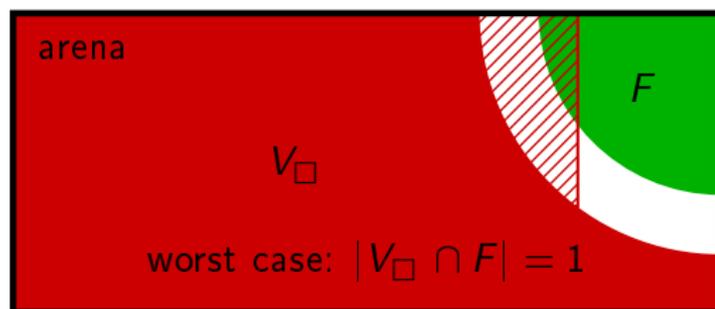
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## Solving Büchi Games



## Remark

- 'outdated' approach
- $O(n^2)$

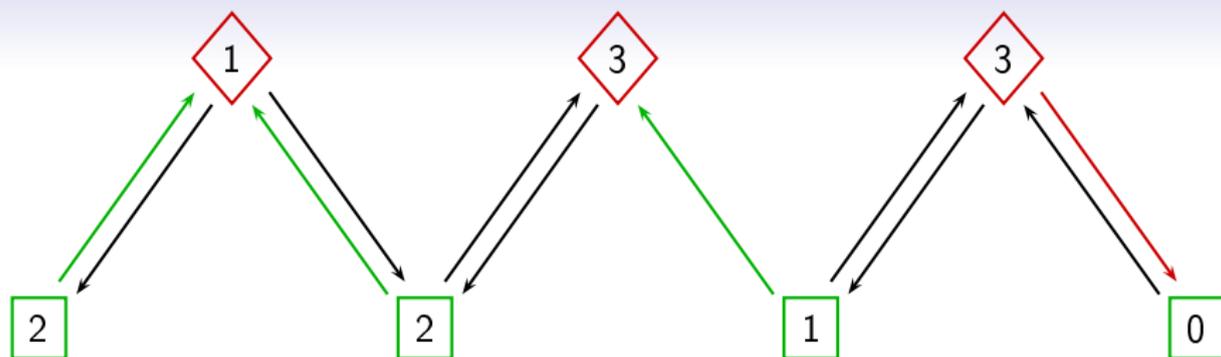
[Chatterjee and Henzinger 12]

## Part II

# Parity Games

# Overview

# colours	3	4	5	6	7	8
McNaughton	$O(mn^2)$	$O(mn^3)$	$O(mn^4)$	$O(mn^5)$	$O(mn^6)$	$O(mn^7)$
Browne & al.	$O(mn^3)$	$O(mn^3)$	$O(mn^4)$	$O(mn^4)$	$O(mn^5)$	$O(mn^5)$
Jurdziński	$O(mn^2)$	$O(mn^2)$	$O(mn^3)$	$O(mn^3)$	$O(mn^4)$	$O(mn^4)$
w.o. strategy / [GW15]	$O(mn)$		$O(mn^2)$		$O(mn^3)$	
<b>Big Steps [S07]</b>	$O(mn)$	$O(mn^{1\frac{1}{2}})$	$O(mn^2)$	$O(mn^{2\frac{1}{3}})$	$O(mn^{2\frac{3}{4}})$	$O(mn^{3\frac{1}{16}})$
[CHL15]	$O(n^{2.5})$	$O(n^3)$	$O(n^{3\frac{1}{3}})$	$O(n^{3\frac{3}{4}})$	$O(n^{4\frac{1}{16}})$	$O(n^{4\frac{9}{20}})$



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- set  $c$  to the maximal colour,  $\sigma$  to  $c$  modulo 2, and  $\bar{\sigma}$  to  $1 - \sigma$
- set  $A$  to  $\sigma$ -attractor( $\alpha^{-1}(c)$ )
- set  $(U_0, U_1)$  to  $\text{McNaughton}(P \setminus A)$
- set  $W_{\bar{\sigma}}$  to  $\bar{\sigma}$ -attractor( $U_{\bar{\sigma}}$ ), and set  $W_{\sigma}$  to  $\emptyset$
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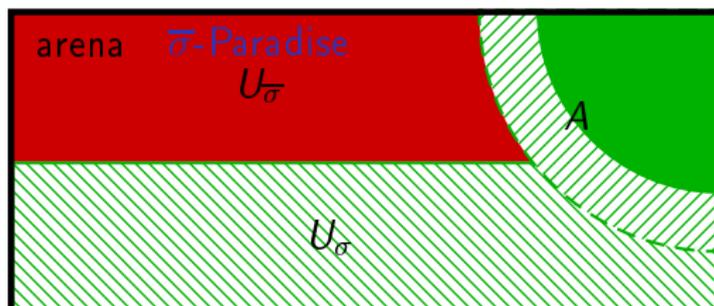
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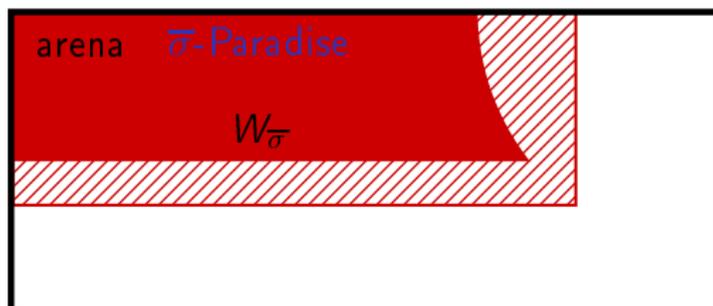
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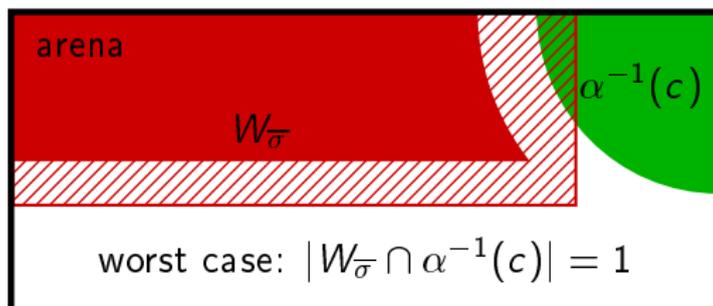
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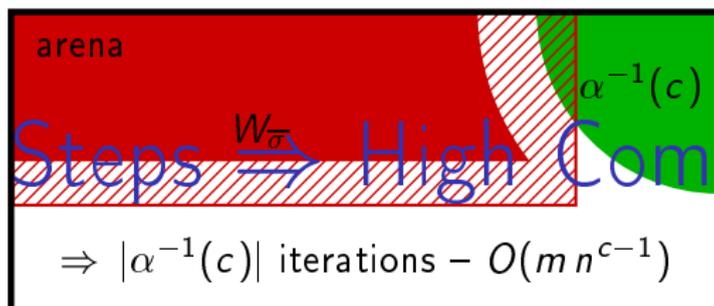
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## McNaughton's Algorithm—Weakness

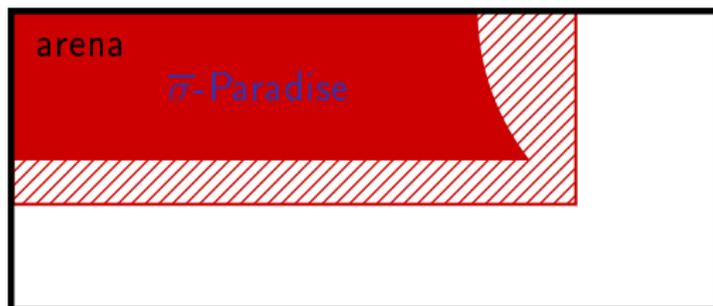


Small Steps  $\Rightarrow$  High Complexity

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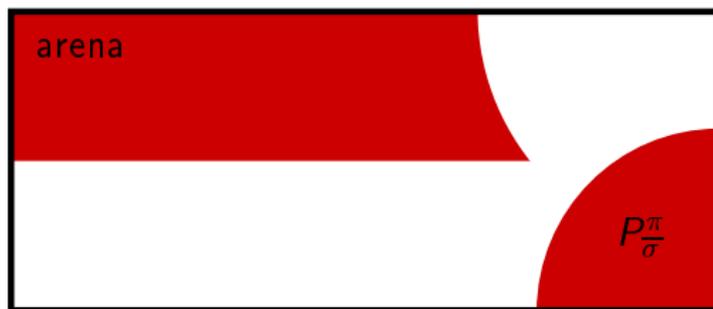
# $\sigma$ -Paradise



## Definition – $\sigma$ -Paradise

- Subset  $P_\sigma$  of the positions, s.t. player  $\sigma$  has a strategy to
  - stay in  $P_\sigma$  ( $\bar{\sigma}$ -trap)
  - that is winning for all states in  $P_\sigma$ .
- $\sigma$ -Paradises are closed under
  - union, and
  - $\sigma$ -attractor.

# $\sigma/\pi$ -Paradise



## Definition – $\sigma/\pi$ -Paradise

- Paradise  $P_{\sigma}^{\pi}$  that contains all  $\sigma$ -paradisess of size  $\leq \pi$ .
- $\sigma/\pi$ -Paradisess are closed under
  - union with any  $\sigma$ -paradisess, and
  - $\sigma$ -attractor.

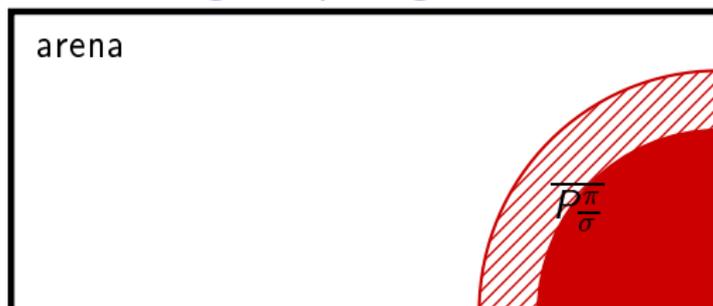
## Big-Step Algorithm



BigStep Algorithm – for  $\mathcal{P} = \langle V_0, V_1, E, \alpha \rangle$

- set  $c$  to the maximal color,  $\sigma$  to  $c$  modulo 2, and  $\bar{\sigma}$  to  $1 - \sigma$
- compute  $\bar{\sigma}/\pi$ -paradise  $P_{\bar{\sigma}}^{\pi}$ , and set  $\overline{P_{\bar{\sigma}}^{\pi}}$  to  $\bar{\sigma}$ -attractor( $P_{\bar{\sigma}}^{\pi}$ )
- set  $\mathcal{P}'$  to  $\mathcal{P} \setminus \overline{P_{\bar{\sigma}}^{\pi}}$
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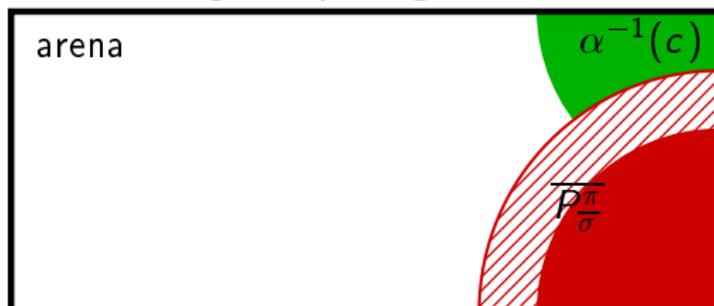
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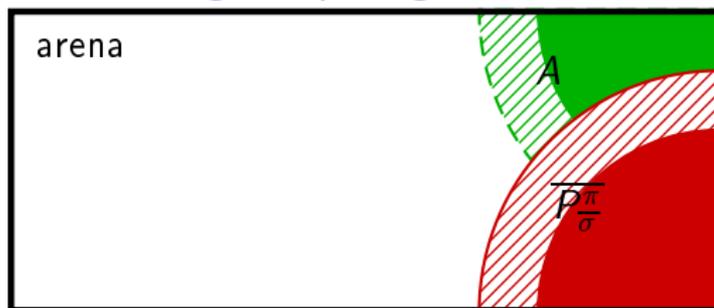
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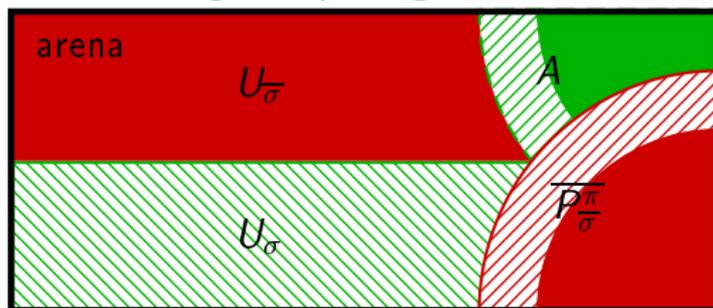
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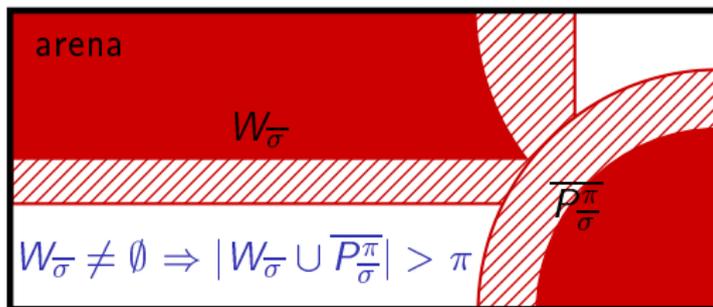
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- set  $W_{\bar{\sigma}}$  to  $\bar{\sigma}$ -attractor( $U_{\bar{\sigma}} \cup \overline{P_{\bar{\sigma}}^{\pi}}$ ), and set  $W_{\sigma}$  to  $\emptyset$
- set  $(U_0, U_1)$  to BigStep( $\mathcal{P} \setminus W_{\bar{\sigma}}$ ), return  $(W_0 \dot{\cup} U_0, W_1 \dot{\cup} U_1)$

## Jurdziński, Paterson, and Zwick

- invented this approach
- used it to establish a deterministic  $n^{O(\sqrt{n})}$  bound

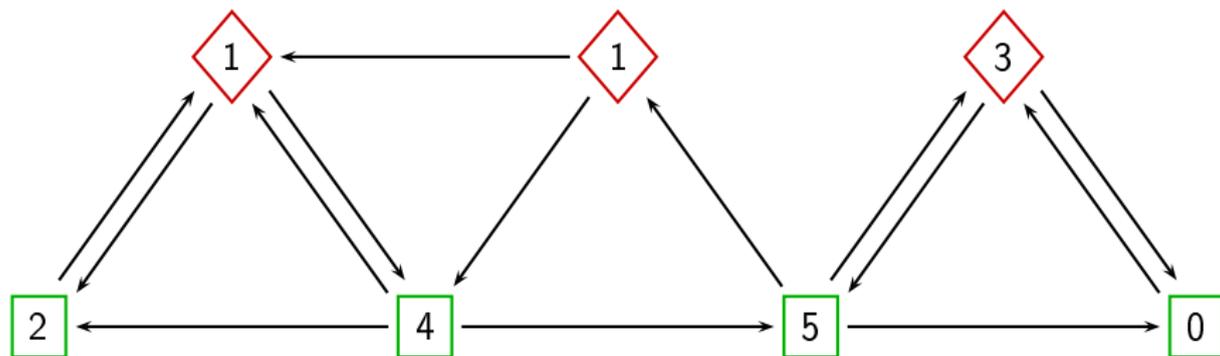
### Brute Force

(roughly)

- try all sets of size up to  $\pi \in O(\sqrt{n})$
- there are some  $n^{O(\sqrt{n})}$  many
- each level has up to  $O(\sqrt{n})$  many calls
- call tree of size  $n^{O(\sqrt{n})}$

drawback:  $c$  is, in fact, usually tiny compared to  $\sqrt{n}$

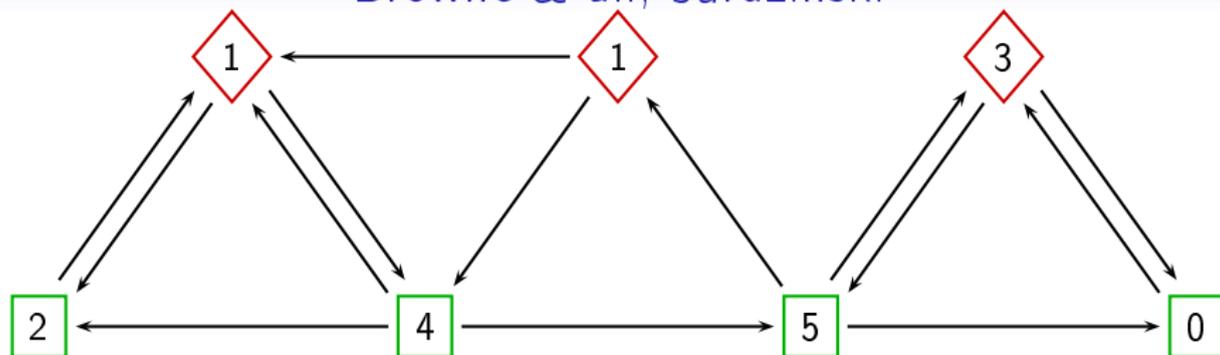
## Browne &amp; al., Jurdziński



If you follow a winning strategy of **even** on  $W_0$ , then ...

- player **odd** cannot force  $> |\alpha^{-1}(c)|$  occurrences of any **odd colour**  $c$  without a **higher even** colour in between
- player **even** can force  $> |\alpha^{-1}(c)|$  occurrences of some (not a particular!) **even colour**  $c$  without a **higher odd** colour in between

## Browne &amp; al., Jurdziński



Rules: Jurdziński: backwards, order on counter vector

- we start at some **initial positions** with counters for, say, the **odd colours** only, initially **set to 0**
- each player chooses how to continue on her vertices
- if we pass an **odd colour**  $c$ , the counter is increased
- if we pass an **even colour**  $c$ , all counters for **smaller** colours are **re-set**
- player **odd** wins if a counter exceeds  $|\alpha^{-1}(c)|$

# Big Steps – What if $c$ is Small?

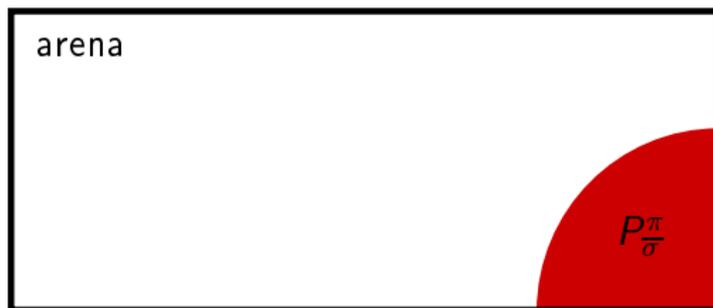
— the common case —

Stop counting at  $\pi$

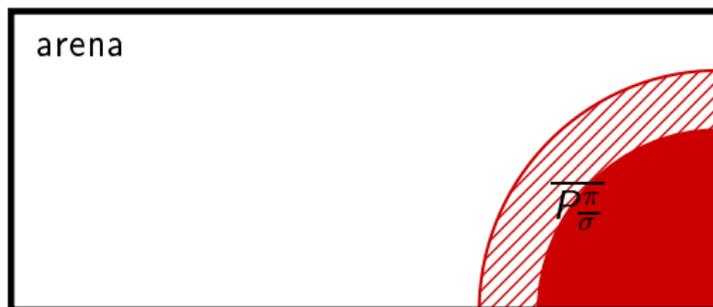
(simple!)

- $\lceil 0.5c \rceil$  many counters
- their sum bounded by  $\pi$
- $\leq \binom{\pi + \lceil 0.5c \rceil}{\pi} \approx \frac{\pi^{\lceil 0.5c \rceil}}{\pi}$  values
- covers all  $\sigma$ -paradises  $P_{\sigma}$  with  $|P_{\sigma}| \leq \pi$
- Complexity:  $O(c m \pi^{\lceil 0.5c \rceil})$

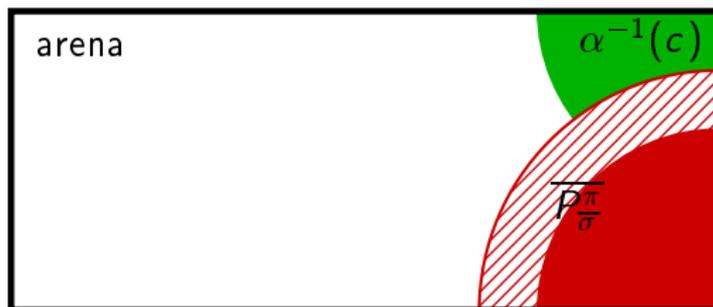
# Big-Step Algorithm



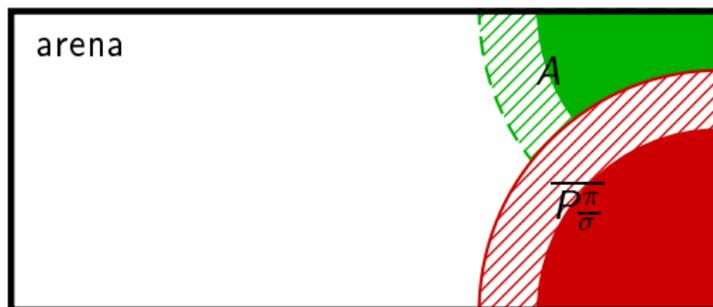
# Big-Step Algorithm



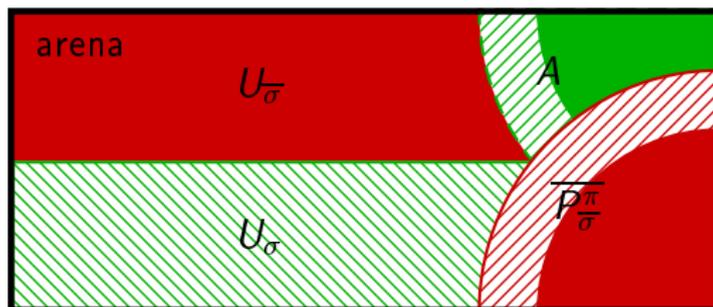
# Big-Step Algorithm



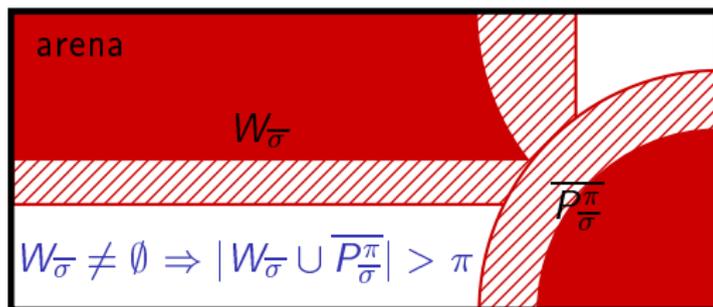
# Big-Step Algorithm



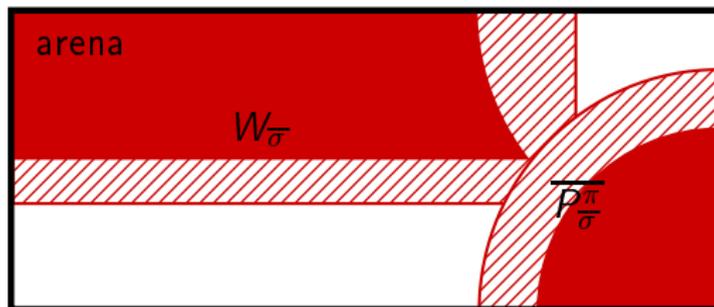
# Big-Step Algorithm



# Big-Step Algorithm



## Solving Parity Games in Big Steps – Complexity



number of colours	3	4	5	6	7	8
paradise construction	-	$O(mn)$	$O(mn^{1\frac{1}{2}})$	$O(mn^2)$	$O(mn^{2\frac{1}{3}})$	$O(mn^{2\frac{3}{4}})$
chosen parameter $\pi_c(n)$	-	$n^{\frac{1}{2}}$	$n^{\frac{1}{2}}$	$n^{\frac{2}{3}}$	$n^{\frac{7}{12}}$	$n^{\frac{11}{16}}$
number of iterations $\frac{n}{\pi_c(n)}$	-	$n^{\frac{1}{2}}$	$n^{\frac{1}{2}}$	$n^{\frac{1}{3}}$	$n^{\frac{5}{12}}$	$n^{\frac{5}{16}}$
solving complexity	$O(mn)$	$O(mn^{1\frac{1}{2}})$	$O(mn^2)$	$O(mn^{2\frac{1}{3}})$	$O(mn^{2\frac{3}{4}})$	$O(mn^{3\frac{1}{16}})$

## State of the Art

# colours	3	4	5	6	7	8
McNaughton	$O(mn^2)$	$O(mn^3)$	$O(mn^4)$	$O(mn^5)$	$O(mn^6)$	$O(mn^7)$
Browne & al.	$O(mn^3)$	$O(mn^3)$	$O(mn^4)$	$O(mn^4)$	$O(mn^5)$	$O(mn^5)$
Jurdziński	$O(mn^2)$	$O(mn^2)$	$O(mn^3)$	$O(mn^3)$	$O(mn^4)$	$O(mn^4)$
w.o. strategy / [GW15]	$O(mn)$		$O(mn^2)$		$O(mn^3)$	
<b>Big Steps [S07]</b>	$O(mn)$	$O(mn^{1\frac{1}{2}})$	$O(mn^2)$	$O(mn^{2\frac{1}{3}})$	$O(mn^{2\frac{3}{4}})$	$O(mn^{3\frac{1}{16}})$
[CHL15]	$O(n^{2.5})$	$O(n^3)$	$O(n^{3\frac{1}{3}})$	$O(n^{3\frac{3}{4}})$	$O(n^{4\frac{1}{16}})$	$O(n^{4\frac{9}{20}})$

- Significantly improved complexity bound
  - from  $O(c m (\frac{n}{0.5c})^{\lceil 0.5c \rceil})$  to  $O(m (\frac{\kappa n}{c})^{\gamma(c)})$  for
 
$$\gamma(c) = \frac{1}{3}c + \frac{1}{2} - \frac{1}{3c} - \frac{1}{\lceil \frac{c}{2} \rceil \lfloor \frac{c}{2} \rfloor}$$
 if  $c$  is even, and
 
$$\gamma(c) = \frac{1}{3}c + \frac{1}{2} - \frac{1}{\lceil \frac{c}{2} \rceil \lfloor \frac{c}{2} \rfloor}$$
 if  $c$  is odd
- Second improvement that reduces the growth in # colours

# Part III

## Bounded Treewidth & Co

## Other Parameter

Parity games are in P for other parameters than # colours

- tree-width [Obdržálek 03]
- DAG-width [Berwanger, Dawar, Hunter, and Kreutzer 06]
- clique-width [Obdržálek 07]

Hope

Can this be a foundation for a tractable algorithm?

## A 'Positive' Result

### Fearnley and Schewe 2013

- $NC^2$  for bounded tree-width  $k$
- + improved bound  $O(nc^{2(k+1)^2}) \rightsquigarrow O((nk^2 k!(c+1)^{3k+1}))$
- + fixed parameter tractable for bounded DAG-width

### Improved by Ganardi 2015

- LogCFL for bounded tree-width
- LogCFL for bounded cleaque-width
- LogDCFL for tree-width 2

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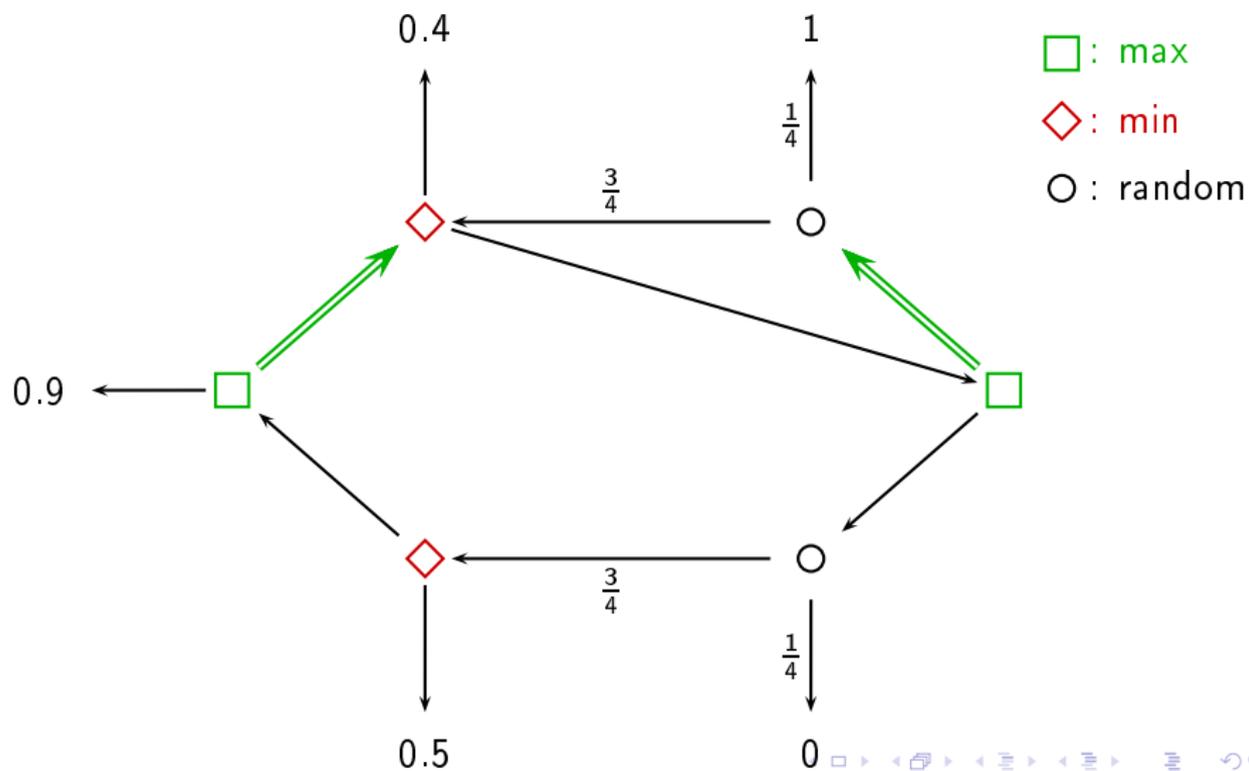
- LogCFL for bounded tree-width
- LogCFL for bounded cleaque-width
- LogDCFL for tree-width 2

# Part IV

## Strategy Improvement

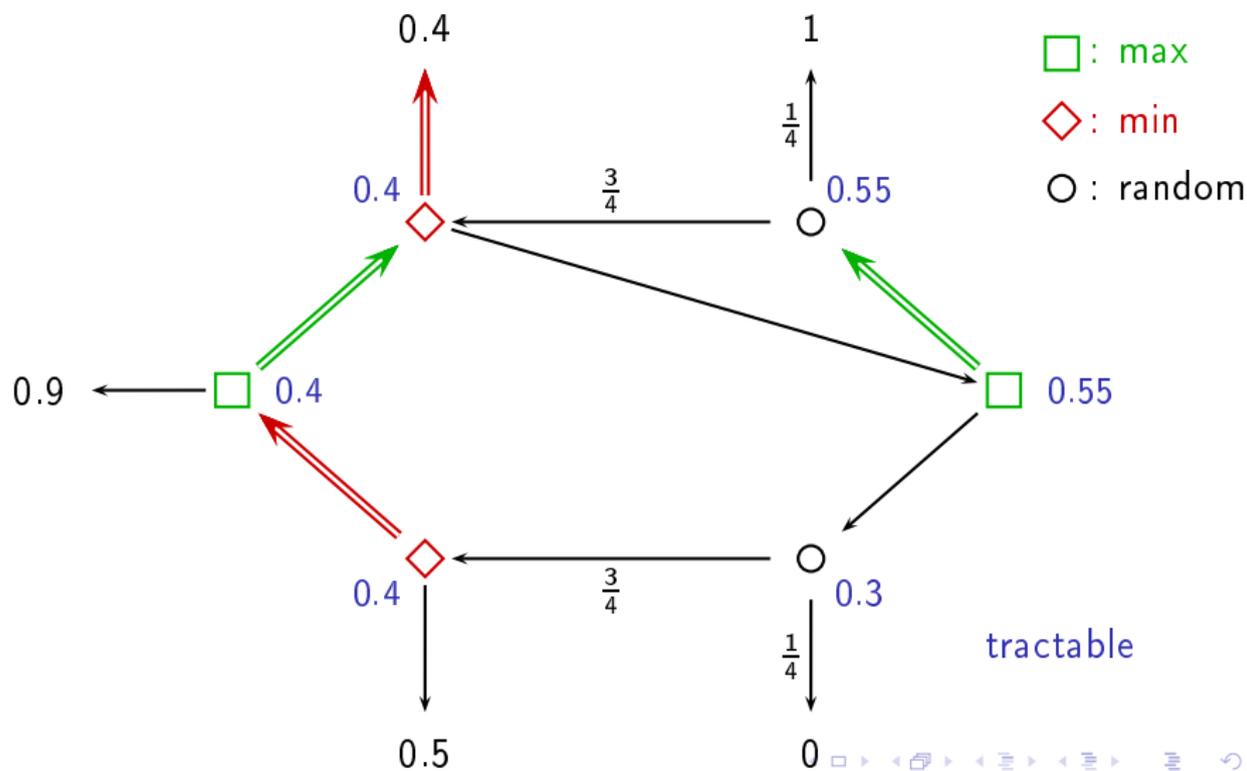
# Classic Strategy Improvement

## fix strategy



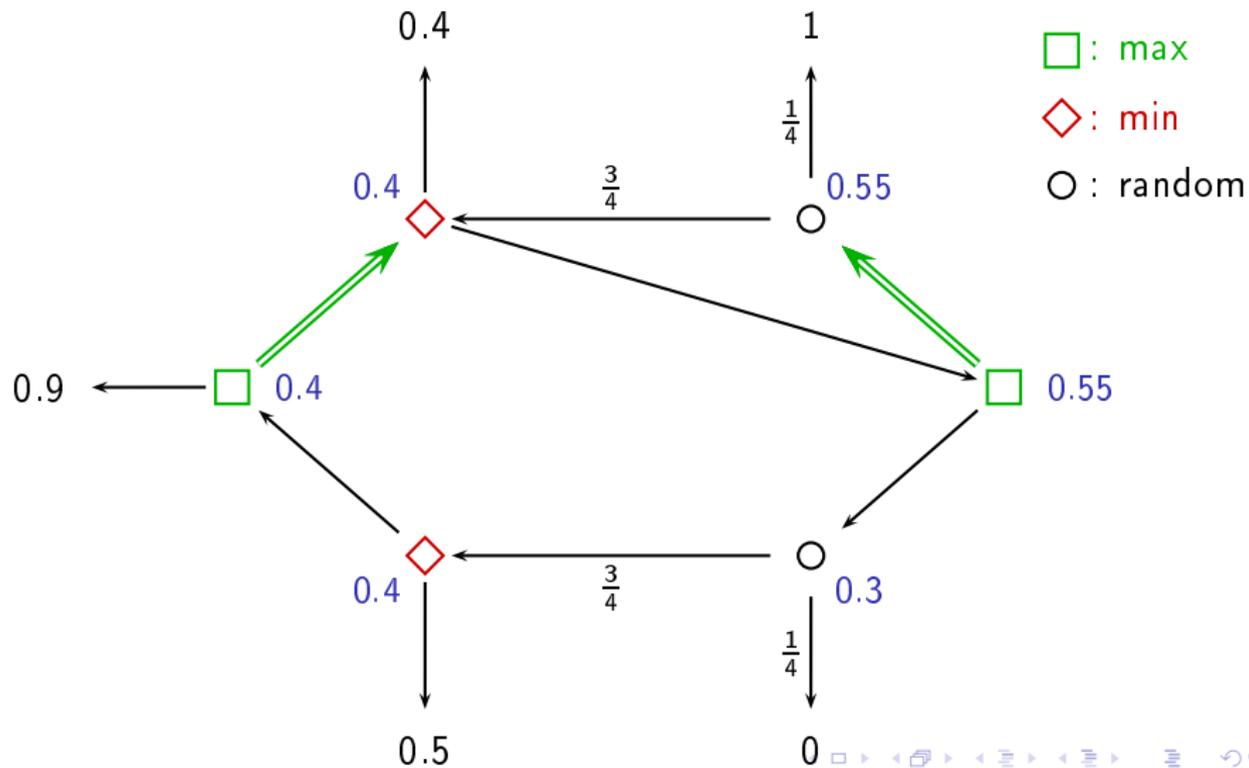
# Classic Strategy Improvement

find best response and evaluate



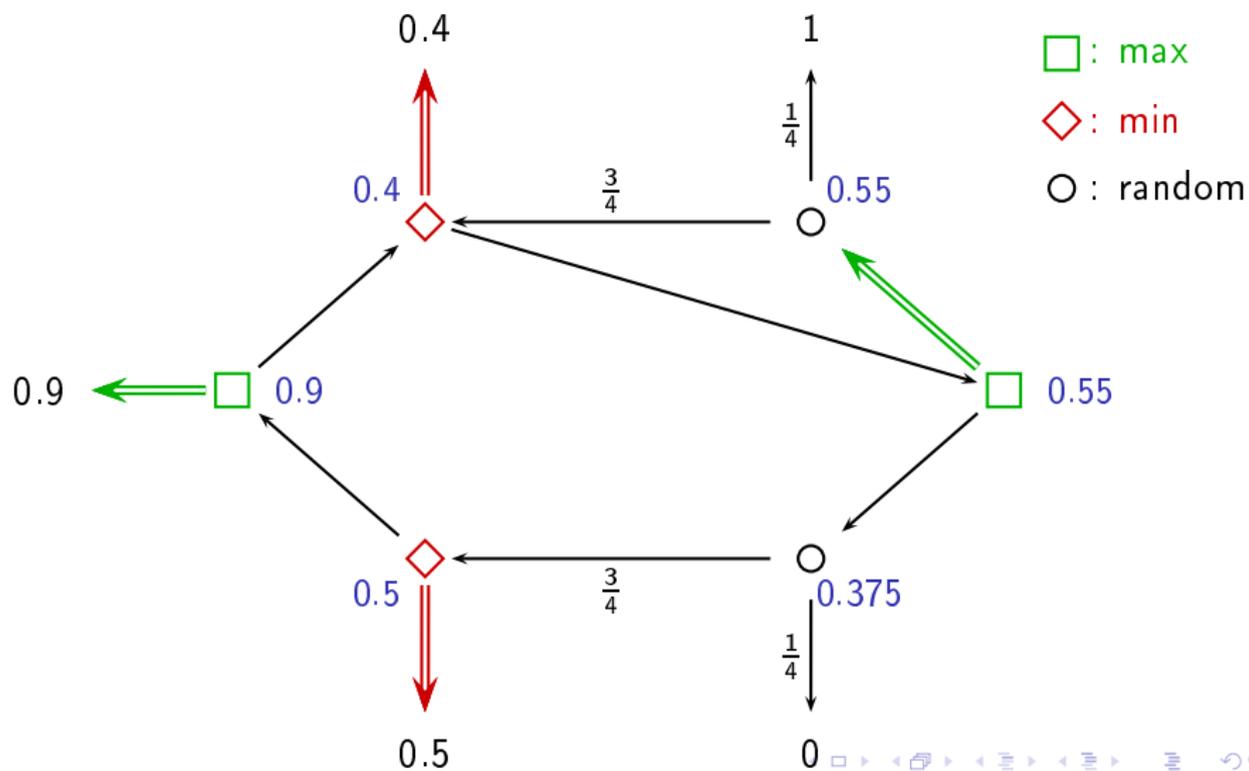
# Classic Strategy Improvement

apply local improvements



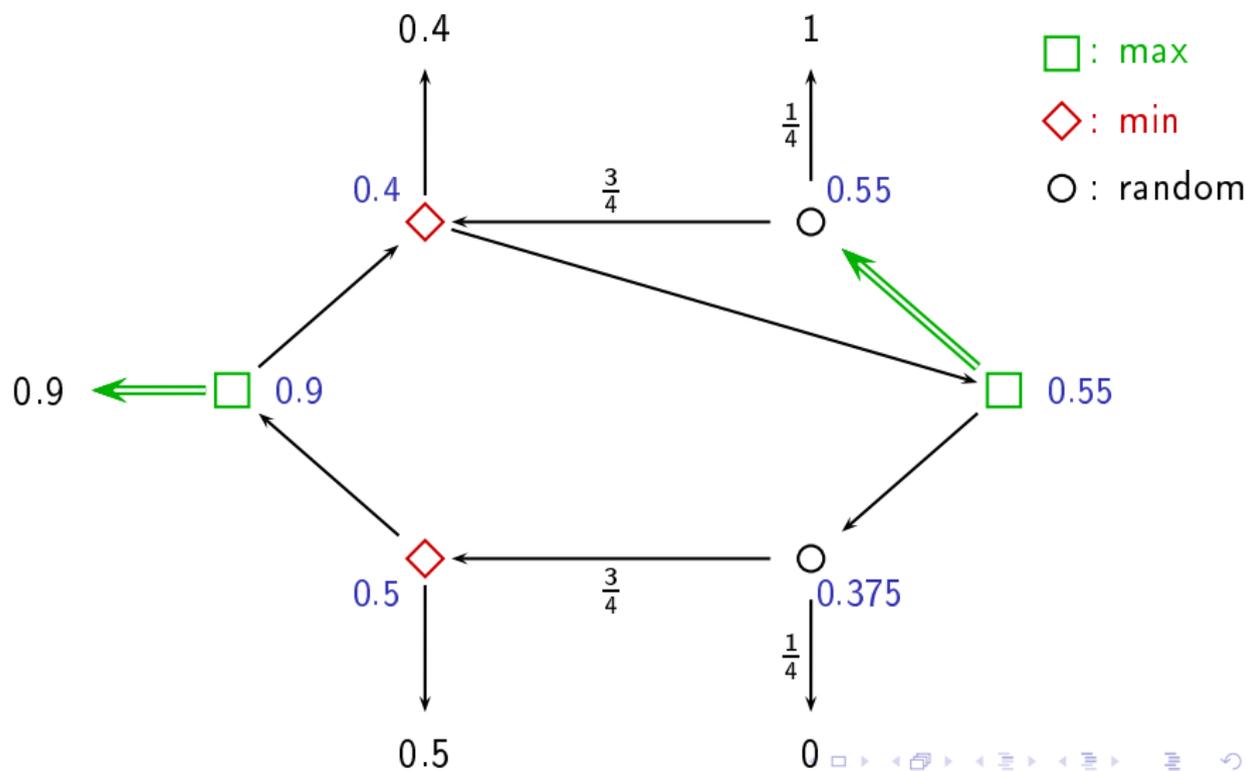
# Classic Strategy Improvement

find best response & evaluate



# Classic Strategy Improvement

no local improvent: done



## CSI – failed hope

- was long hoped to be tractable
- many update policies
- ∇ exponential lower bounds
  - use static update policy
- ∃ PSPACE powerful

[Friedmann 11, ...]

[Fearnley+Savani 15]

# SYMMETRY

## Symmetry and Complexity

[Jurdziński 98]

- 1 guess valuation
  - 2 verify
- ⇒ one value: UP  
symmetry:  $UP \cap CoUP$

## Iterated Fixed Point [Emerson+Lei 86]

parity games

- similar treatment
- best performing algorithm

## Optimal Strategy Improvement

[Schewe 08]

parity games, MPG mean partitions

- some symmetry
- fab performance

# Why not?

## Naive symmetric strategy improvement

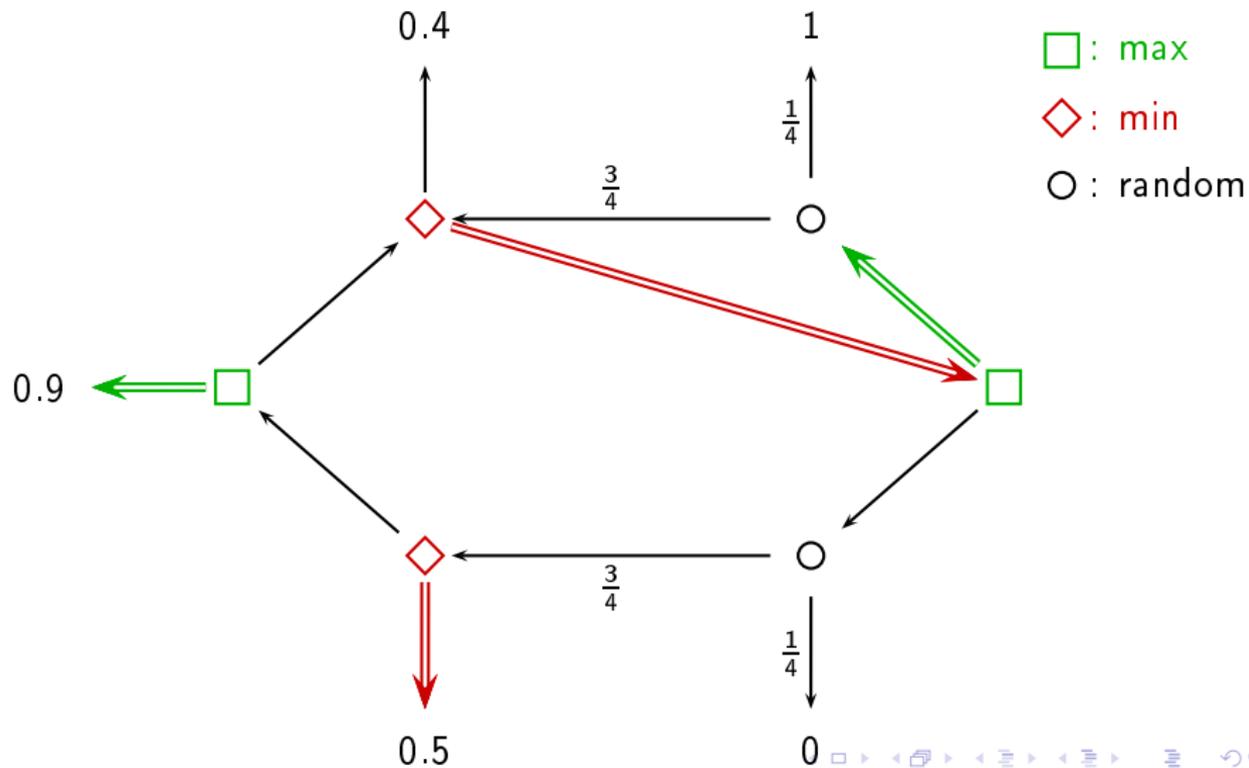
**Question:** Why has SSI not been thoroughly studied?

**Answer:** Anne Condon has proved it wrong [Condon 93]

- 1 Concurrent Switch
- 2 Alternating Best Response

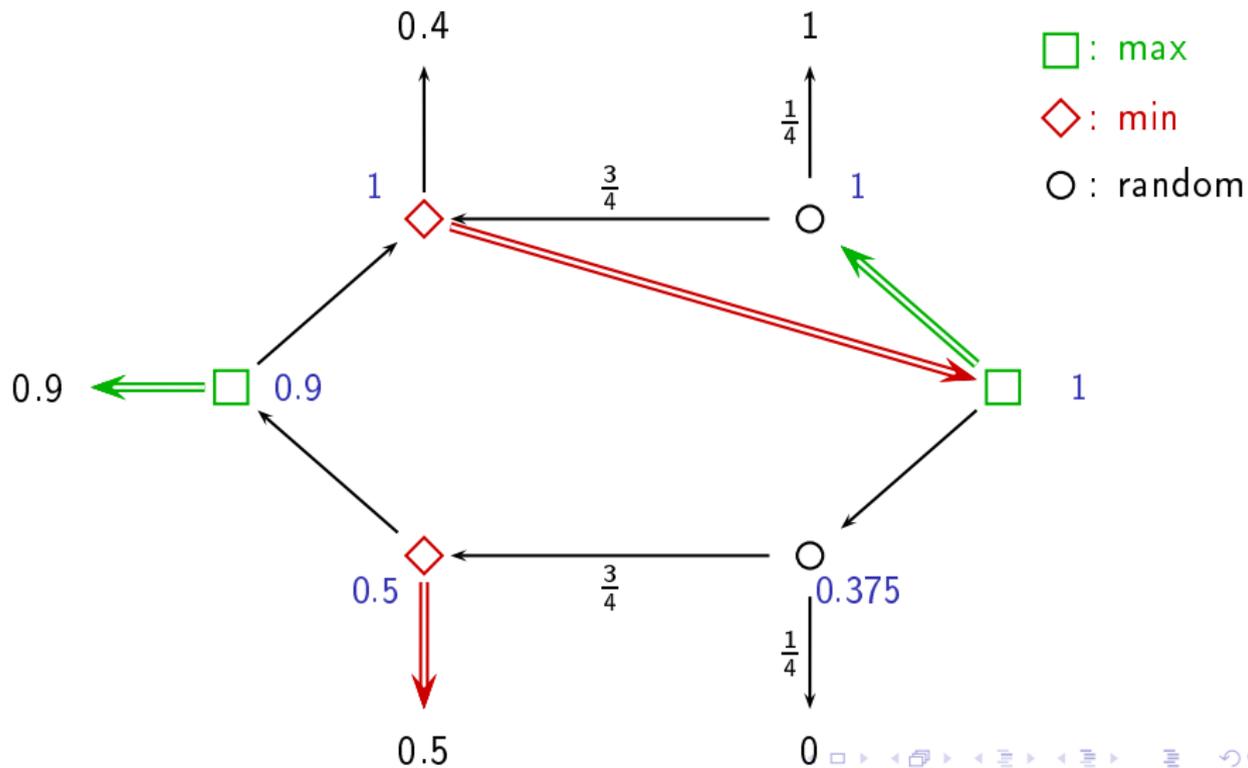
# Concurrent Switch

starting strategies



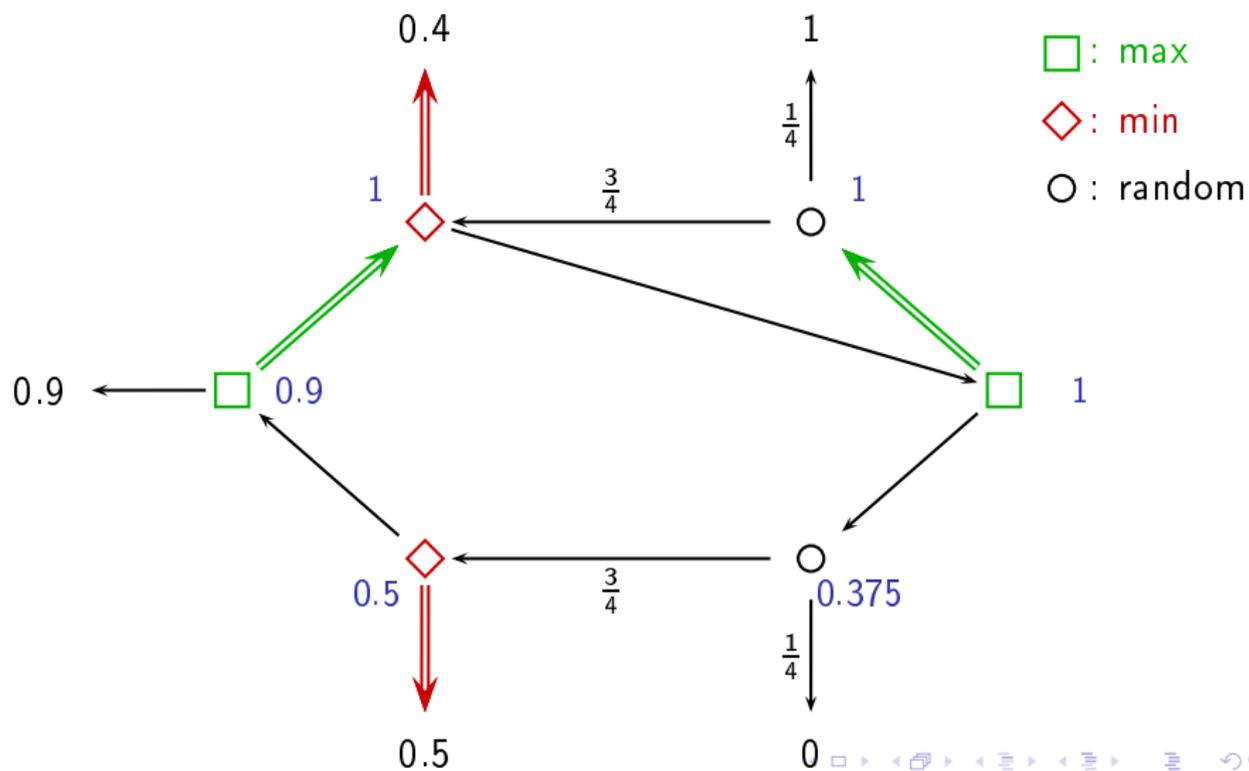
# Concurrent Switch

evaluate



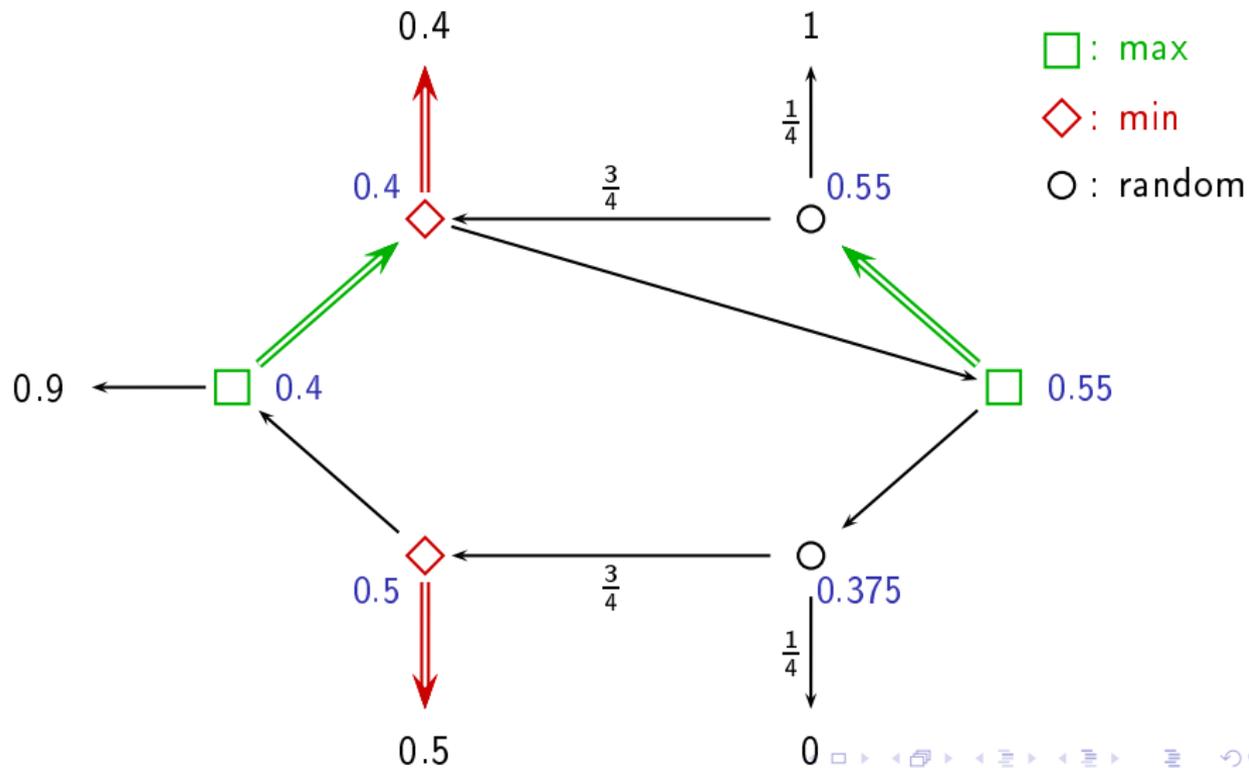
# Concurrent Switch

update strategies



# Concurrent Switch

update evaluation

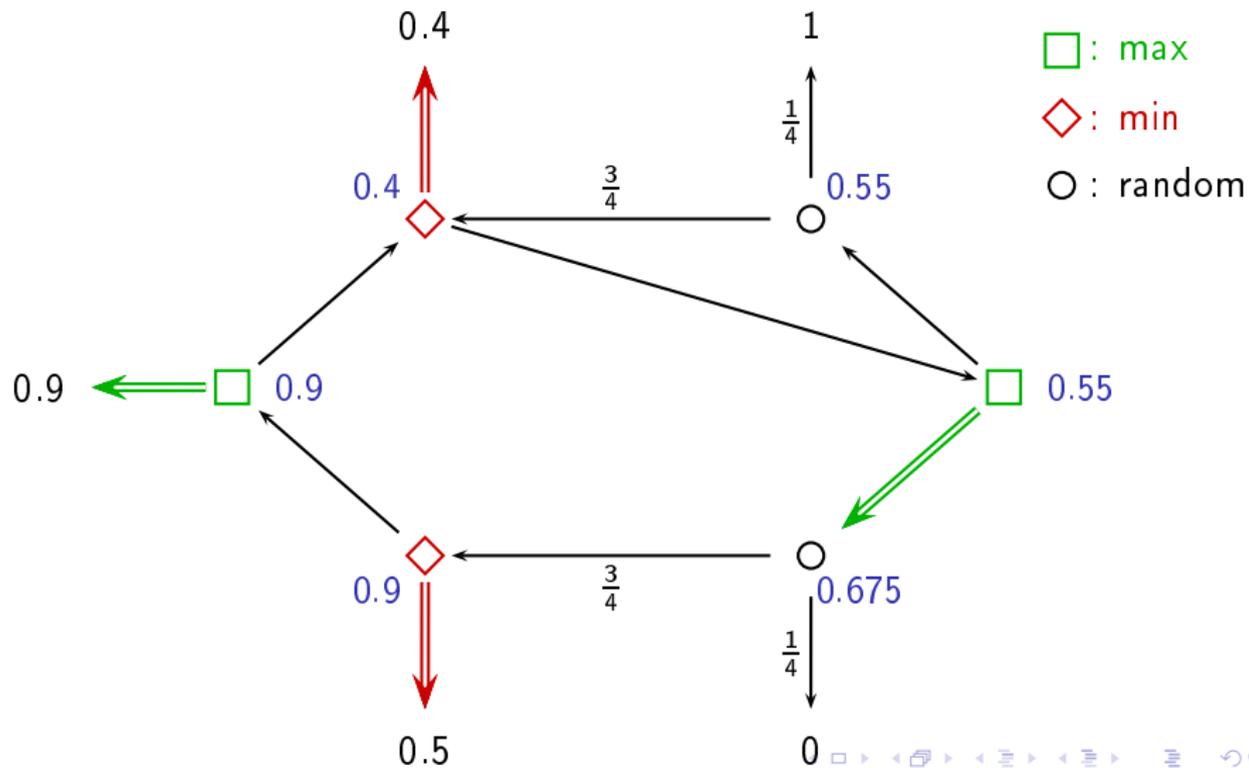






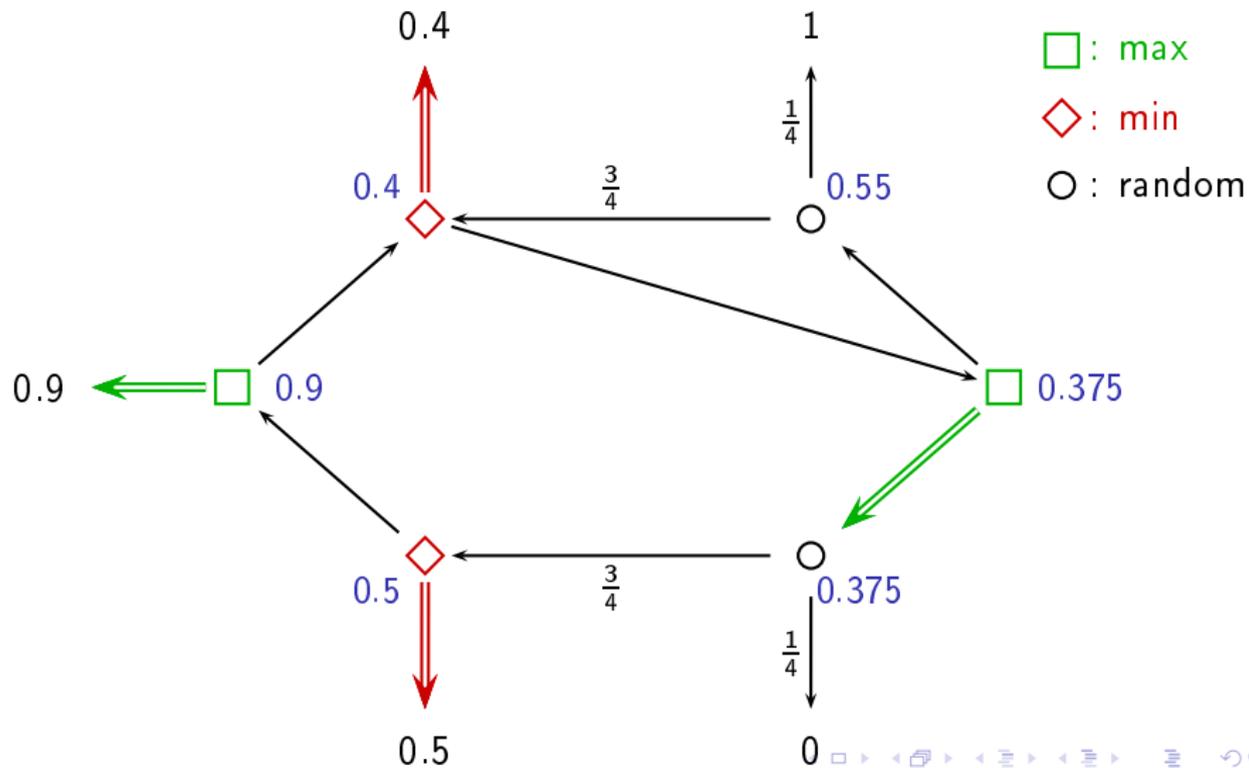
# Concurrent Switch

update strategy



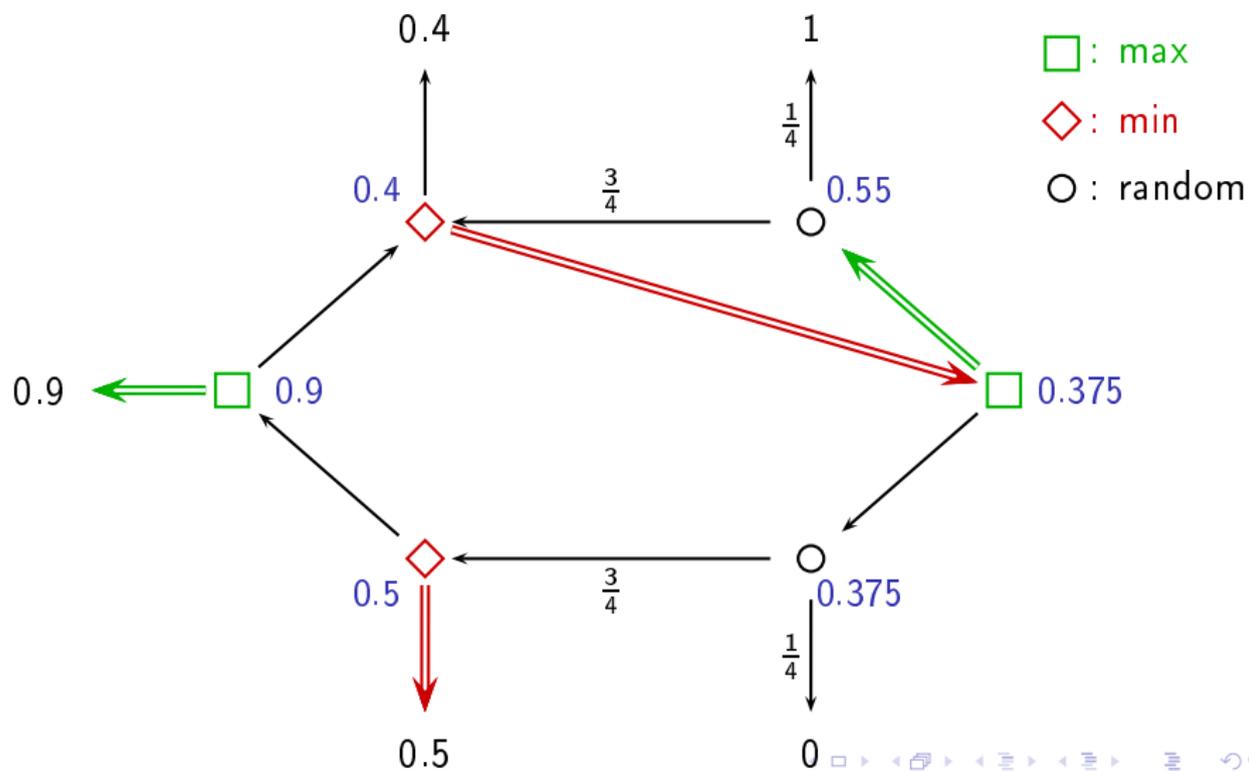
# Concurrent Switch

update evaluation



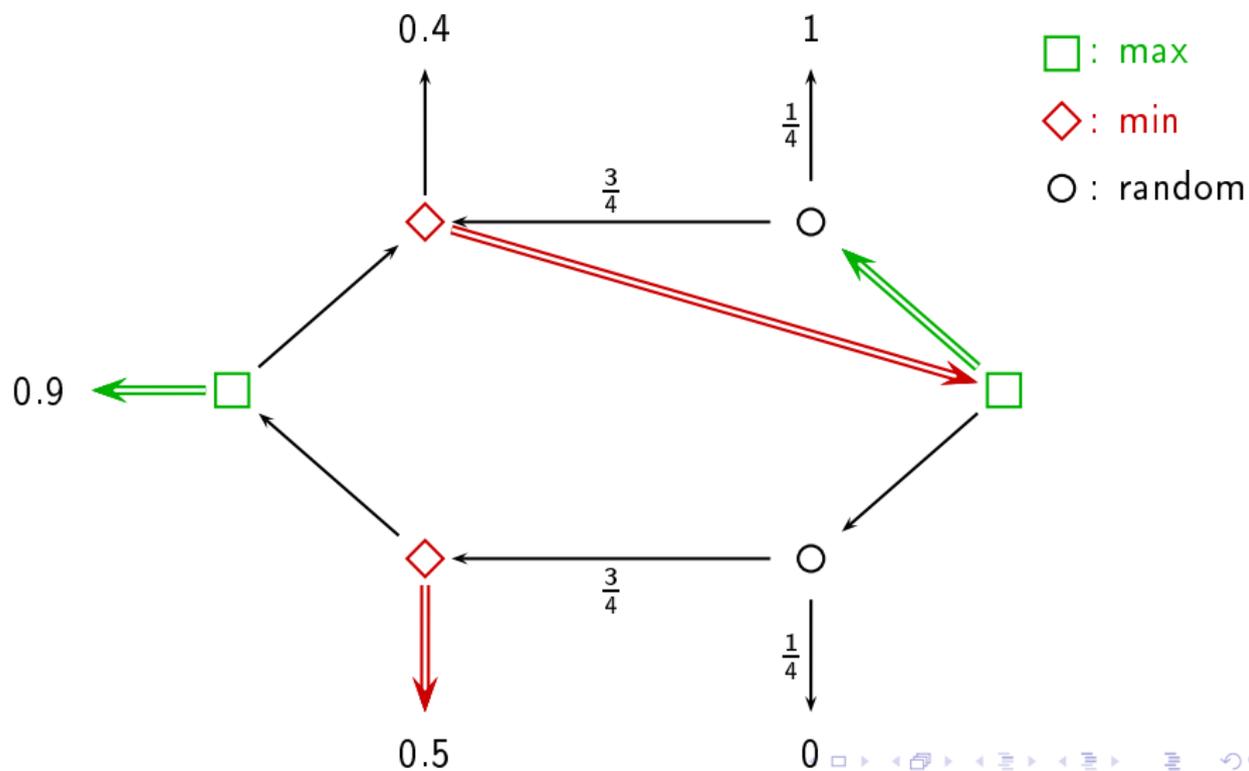
# Concurrent Switch

update strategy (cycle)



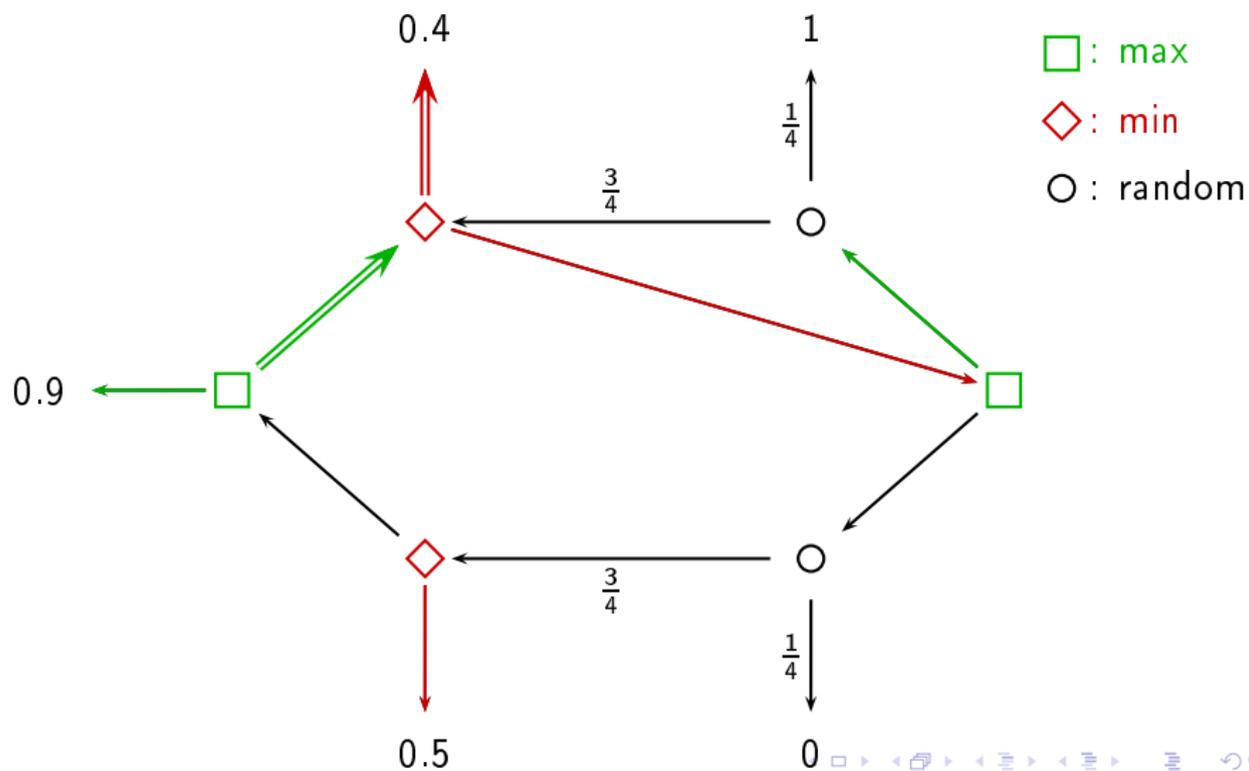
# Symmetric Strategy Improvement

starting strategies



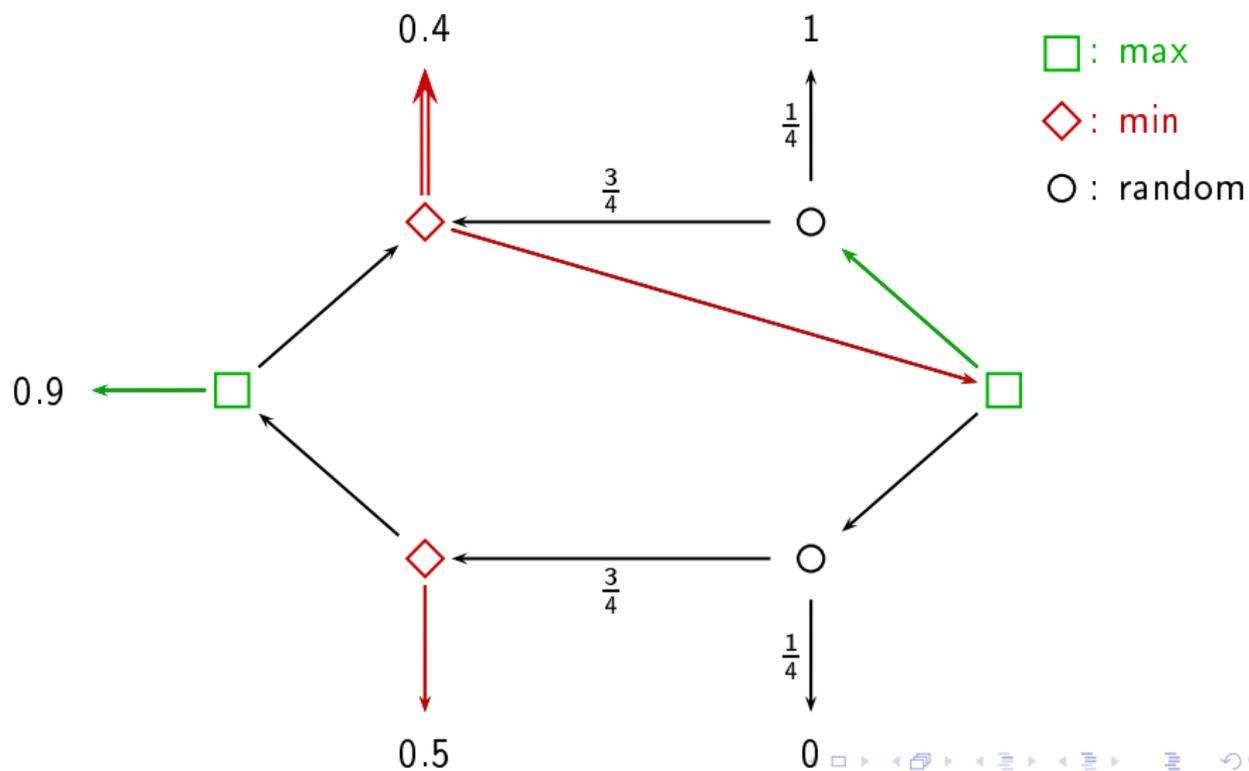
# Symmetric Strategy Improvement

evaluate – best response



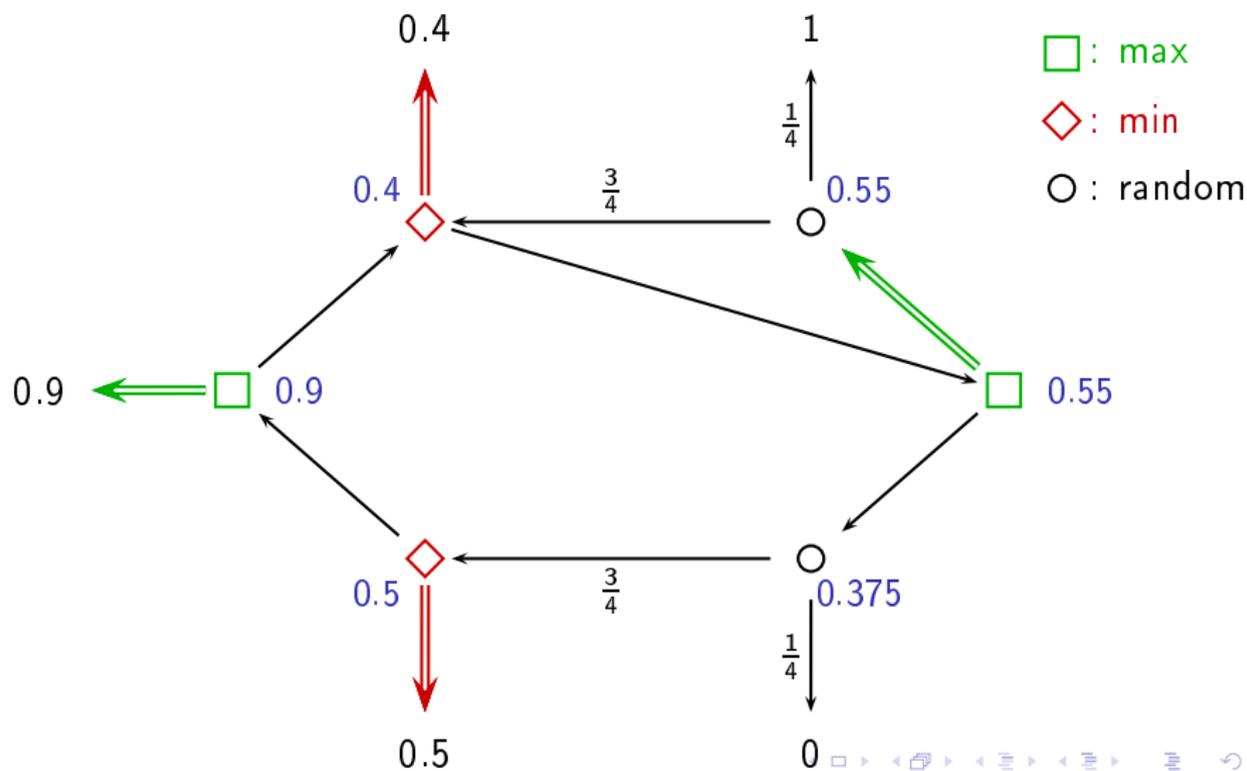
# Symmetric Strategy Improvement

best response & improvement



# Symmetric Strategy Improvement

update (done)



# Symmetric Strategy Improvement

## Can SSI help overcome problems of CSI?

**Question:** How about single player examples? [Fearnley 10]

**Answer:** Easy (but no surprise there)

**Question:** How about Friedmann's traps? [Friedmann 11,...]

**Answer:** Yes but this doesn't imply there are no traps

**Question:** Less iterations on random games?

**Answer:** Yes but probably not half

**Question:** Is SSI polynomial?

**Answer:** Look at the weather! Isn't it lovely?

## Friedmann's Traps

Switch Rule	1	2	3	4	5	6	7	8	9	10
Cunningham	2	6	9	12	15	18	21	24	27	30
CunninghamSubexp	1	1	1	1	1	1	1	1	1	1
FearnleySubexp	4	7	11	13	17	21	25	29	33	37
FriedmannSubexp	4	9	13	15	19	23	27	31	35	39
RandomEdgeExpTest	1	2	2	2	2	2	2	2	2	2
RandomFacetSubexp	1	2	7	9	11	13	15	17	19	21
SwitchAllBestExp	4	5	8	11	12	13	15	17	18	19
SwitchAllBestSubExp	5	7	9	11	13	15	17	19	21	23
SwitchAllSubExp	3	5	7	9	10	11	12	13	14	15
SwitchAllExp	3	4	6	8	10	11	12	14	16	18
ZadehExp	-	6	10	14	18	21	25	28	32	35
ZadehSubexp	5	9	13	16	20	23	27	30	34	37

# Parity Games

with few colours

# colours	3	4	5	6	7	8
McNaughton	$O(mn^2)$	$O(mn^3)$	$O(mn^4)$	$O(mn^5)$	$O(mn^6)$	$O(mn^7)$
Browne & al.	$O(mn^3)$	$O(mn^3)$	$O(mn^4)$	$O(mn^4)$	$O(mn^5)$	$O(mn^5)$
Jurdziński	$O(mn^2)$	$O(mn^2)$	$O(mn^3)$	$O(mn^3)$	$O(mn^4)$	$O(mn^4)$
w.o. strategy / [GW15]	$O(mn)$		$O(mn^2)$		$O(mn^3)$	
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# Parity Games

## Further complexity results

- $NP \cap CoNP$  [McNaughton 93]
- $UP \cap CoUP$  [Zwick and Paterson '96, Jurziński 98]
- PLS [Beckmann and Moller 08]
- PPAD [Etessami and Yannakakis 10]
- $n^{O(\sqrt{n})}$  [Jurziński, Zwick, and Paterson 08]
- in LogCFL for bounded tree- and clique-width [Ganardi 15]
- fixed parameter tractable for bounded DAG-width

# Parity & Pay-Off Games

## Strategy Improvement

- deterministic update [Puri 95, Vöge and Jurdziński 00]
- randomised updates [Ludwig 95, Björklund and Vorobyov 07]
- one-step optimal updates [S 08]
- they are all expensive [Friedmann 09, FHZ 11a]
- symmetric strategy improvement [STV 15]

# Parity Games

... are simply **beautiful!**