



# Goal

- Introduce games and simulations thereon.
  - Relevant simulation is PA-I-simulation.
- Introduce General Coarsest Partition Problem.
- Link between GCPP and PA-I-simulation.
- Solve the GCPP with polynomial algorithm.
  - Obtain largest probabilistic alternating simulation relation.

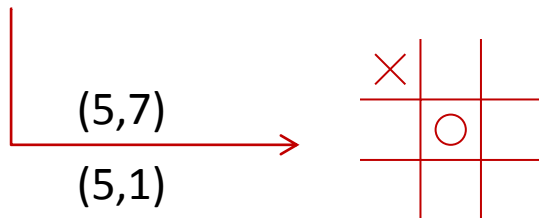
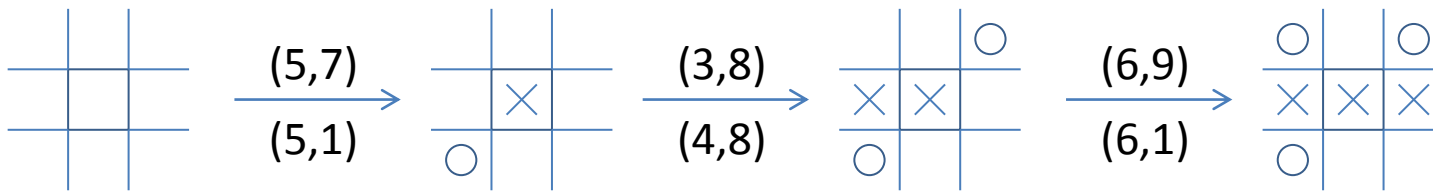
# Example of a game

- Probabilistic simultaneous tic-tac-toe:
  - Similar to tic-tac-toe:
    - Board consists of nine squares.
    - At most one player can mark a square.
    - Three in a row wins (vertical, horizontal or diagonal).
  - Turns are simultaneous.
  - A player picks a favorite and an alternative square:
    - If the favorites differ, both players mark their favorite.
    - If the favorites are the same, each player has a 50% chance of marking their favorite, the other player marks his alternative.
  - If there is only one square left, and there is no winner, then the game is over, and it's a draw.

# Example run

- Player I
- × Player II

1	2	3
4	5	6
7	8	9



# Probabilistic game structure

- A probabilistic game structure consists of:
  - A finite set of states  $S$ , with an initial state  $s_0$ .
  - A finite set of joint actions  $Act = Act_I \times Act_{II}$ ,
    - $Act_I$  is the set of actions for player I.
  - A labeling function  $L : S \rightarrow 2^{Prop}$
  - A transition function  $\delta : S \times Act \rightarrow \mathcal{D}(S)$

# Example (ctd.)

- A state is a board with equal number of O's and X's, and squares marked at most once.
  - Initial state is the empty board.
- An action for one player is any pair  $(n, m)$ , with  $1 \leq n, m \leq 9$  and  $n \neq m$ .
- Predicate  $P_1$  holds iff there is a line of three O's, and  $P_2$  holds iff there is a line of three X's.
- The pair  $s, ((n, m), (n', m'))$  maps to  $U(s)$  if  $n, m, n'$  or  $m'$  is marked in  $s$  or if  $P_1$  or  $P_2$  holds. Otherwise:
  - The pair  $s, ((n, m), (n', m'))$  maps to  $U(s[\text{O on } n; \text{X on } n'])$  if  $n \neq n'$ .
  - The pair  $s, ((n, m), (n, m'))$  maps to  $U(s[\text{O on } n; \text{X on } m'], s[\text{O on } m; \text{X on } n])$ .

# Mixed actions

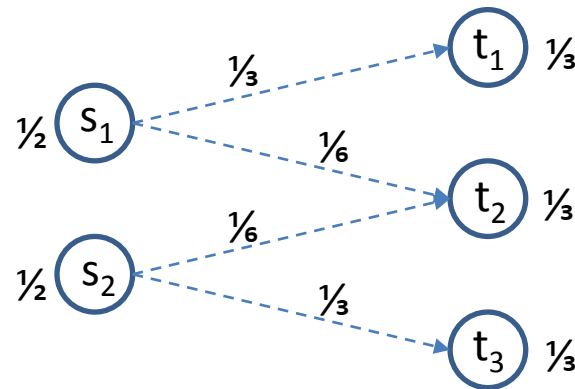
- A mixed action of player I is a distribution over  $Act_I$ , denoted with  $\pi \in \Pi_I$ .

$$\bar{\delta}(s, (\pi_1, \pi_2))(t) = \sum_{a_1 \in Act_I, a_2 \in Act_{II}} \pi_1(a_1) \cdot \pi_2(a_2) \cdot \delta(s, (a_1, a_2))(t)$$

# Simulation

- If  $A$  simulates  $B$ , then every move by  $B$  can be matched by  $A$ .

- Probabilistic:
  - From  $R$  to  $\bar{R}$



- Alternating:

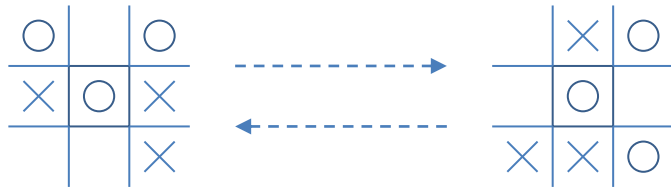
- If  $A$  I-simulates  $B$ , then every move of I in  $B$  can be matched by a move of I in  $A$ , such that for every counter-move of II in  $A$  there is a counter-move of II in  $B$ .



# Probabilistic alternating I-simulation

- A PA-I-simulation is a relation  $\sqsubseteq \subseteq S \times S$  satisfying: if  $s \sqsubseteq t$ , then:
  - $L(s) = L(t)$
  - For all  $\pi_1 \in \Pi_I$ , there exists  $\pi_1' \in \Pi_I$ , such that for all  $\pi_2' \in \Pi_{II}$ , there exists  $\pi_2 \in \Pi_{II}$ , such that  $\delta(s, (\pi_1, \pi_2)) \bar{\sqsubseteq} \delta(t, (\pi_1', \pi_2'))$
- If  $s \sqsubseteq t$  and  $t \sqsubseteq s$ , then  $s$  and  $t$  are PA-I-simulation equivalent.

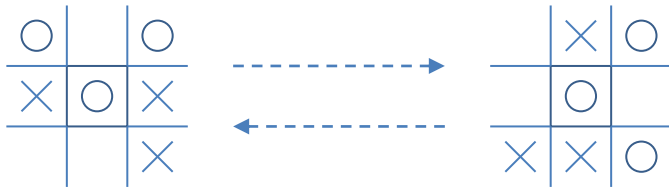
# Examples of PA-I-simulations



PA-I-simulation equivalent!



# Examples of PA-I-simulations



PA-I-simulation equivalent!



4,8 | ?,?  
↓  
Tie

All ties, wins and losses are PA-I-simulation equivalent.

# General Coarsest Partition Problem

- We want to partition a set of states into blocks, where all states in a block are PA-I-simulation equivalent.
- We want to define a partial order,  $\preceq$ , such that if  $P \preceq Q$ , then every state in  $Q$  simulates every state in  $P$ .
- We want to find the coarsest such partition and partial order (partition pair).

# From GCPP to PA-I-simulation

- We write  $[s]_{\Sigma}$  for the block containing  $s$ .
- Define a relation  $\Xi_{(\Sigma, \preceq)} \subseteq S \times S$  as  $s \Xi_{(\Sigma, \preceq)} t$  iff  $[s]_{\Sigma} \preceq [t]_{\Sigma}$
- If  $(\Sigma, \preceq)$  is the coarsest partition pair, then  $s \Xi_{(\Sigma, \preceq)} t$  is the largest PA-I-simulation.
  - There is a unique coarsest partition pair.

# Sketch of approach

- Start with a partition pair that is too coarse.
  - If  $s \sqsubseteq t$  then  $[s]_{\Sigma} \preceq [t]_{\Sigma}$
- Keep refining blocks in the partition pair, until nothing changes.
  - Make sure not to refine too much!
- Return the partition pair.

# Main procedure

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**Algorithm 2** Computing the Generalised Coarsest Partition Pair

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INPUT: a probabilistic game structure  $\mathcal{G} = \langle S, s_0, \mathcal{L}, Act, \delta \rangle$

OUTPUT: a partition pair  $(\Sigma, \preceq)$  on  $S$

function GCPP ( $\mathcal{G}$ )

$\Sigma := \{\{t \mid \mathcal{L}(t) = \mathcal{L}(s)\} \mid s \in S\}; \preceq := \{(B, B) \mid B \in \Sigma\}$

$\Sigma' := \emptyset; \preceq' := \emptyset$

while  $\Sigma \neq \Sigma' \vee \preceq \neq \preceq'$  do

$\Sigma' := \Sigma; \preceq' := \preceq$

for each  $B \in \Sigma$  do

$(\Sigma_B, \preceq_B) := \text{Split}((\Sigma', \preceq'), B)$

$\Sigma := \Sigma \setminus \{B\} \cup \Sigma_B$

$\preceq := \preceq \cup \preceq_B$

$\cup \{(B', X) \mid X \in \Sigma : B' \in \Sigma_B : (B, X) \in \preceq\}$

$\cup \{(X, B') \mid X \in \Sigma : B' \in \Sigma_B : (X, B) \in \preceq\}$

$\setminus \{(B, X), (X, B) \mid X \in \Sigma : (X, B), (B, X) \in \preceq\}$

endfor

endwhile

return  $(\Sigma, \preceq)$

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# Sketch of refining blocks

- Input is a partition pair  $(\Sigma, \preceq)$  and a block  $B \in E$ .
- Start with a partition pair  $(\Sigma_B, \preceq_B)$  that is too fine.
- Keep merging blocks until  $(\Sigma_B, \preceq_B)$  stops changing:
  - For any pair of blocks:
    - If a state  $s$  in  $A$  can simulate a state  $t$  in  $B$  under  $(\Sigma, \preceq)$ , and vice versa, then  $A$  and  $B$  should be the same block.
    - If a state  $s$  in  $A$  can simulate a state  $t$  in  $B$  under  $(\Sigma, \preceq)$ , but not vice versa, then  $A \preceq B$  should hold (and vice versa).
- Return  $(\Sigma_B, \preceq_B)$



# Refinement procedure

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**Algorithm 1** Refining a block to make it stable on a partition pair

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INPUT: a partition pair  $(\Sigma, \preceq)$ , a block  $B \in \Sigma$

OUTPUT: a partition pair  $(\Sigma_B, \preceq_B)$  on  $B$

function Split  $((\Sigma, \preceq), B)$

$\Sigma_B := \{\{s\} \mid s \in B\}$ ;  $\preceq_B := \{(s, s) \mid s \in B\}$ ;  $\Sigma' := \emptyset$ ;  $\preceq' := \emptyset$

while  $\Sigma_B \neq \Sigma' \vee \preceq_B \neq \preceq'$  do

$\Sigma' := \Sigma_B$ ;  $\preceq' := \preceq_B$

for each *distinct*  $B_1, B_2 \in \Sigma_B$  do

*pick any*  $s_1 \in B_1$  and  $s_2 \in B_2$

if  $(\text{CanSim}((\Sigma, \preceq), s_1, s_2) \wedge \text{CanSim}((\Sigma, \preceq), s_2, s_1))$  then

$\Sigma_B := \Sigma_B \setminus \{B_1, B_2\} \cup \{B_1 \cup B_2\}$

$\preceq_B := \preceq_B \cup \{(X, B_1 \cup B_2) \mid X \in \Sigma : (X, B_1) \in \preceq_B \vee (X, B_2) \in \preceq_B\}$

$\cup \{(B_1 \cup B_2, X) \mid X \in \Sigma : (B_1, X) \in \preceq_B \vee (B_2, X) \in \preceq_B\}$

$\setminus \{(B_i, X), (X, B_i) \mid X \in \Sigma : (B_i, X), (X, B_i) \in \preceq_B \wedge i \in \{1, 2\}\}$

else if  $(\text{CanSim}((\Sigma, \preceq), s_1, s_2))$  then

$\preceq_B := \preceq_B \cup \{(B_2, B_1)\}$

else if  $(\text{CanSim}((\Sigma, \preceq), s_2, s_1))$  then

$\preceq_B := \preceq_B \cup \{(B_1, B_2)\}$

endfor

endwhile

return  $(\Sigma_B, \preceq_B)$

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# Correctness

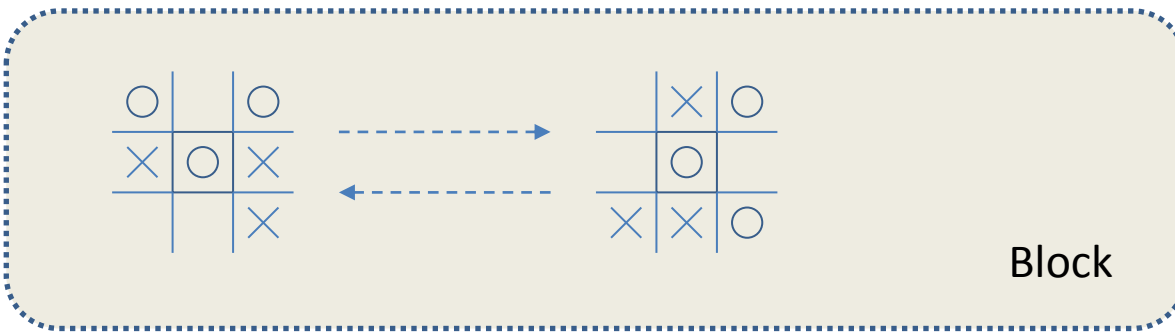
- Define an operator  $\rho$  on partition pairs, by letting  $\rho((\Sigma, \preceq))$  be the coarsest partition pair  $(\Sigma', \preceq')$ , finer than  $(\Sigma, \preceq)$  that is stable on  $(\Sigma, \preceq)$ .
  - If  $(\Sigma_1, \preceq_1)$  is finer than  $(\Sigma_2, \preceq_2)$ , then  $(\Sigma_1, \preceq_1)$  is stable on  $(\Sigma_1, \preceq_1)$ , when for all  $P, Q \in \Sigma_1$  with  $P \preceq_1 Q$  and  $s \in P$  and  $t \in Q$ ,  $s \sqsubseteq_{(\Sigma_2, \preceq_2)} t$ .
- Running Split on each block in  $\Sigma$ , for  $(\Sigma, \preceq)$  is equal to  $\rho((\Sigma, \preceq))$ .
- Theorem 1: If  $(\Sigma, \preceq)$  is the fixed point of  $\rho((\Sigma_0, Id))$  then  $\sqsubseteq_{(\Sigma, \preceq)}$  is the largest PA-I-simulation.

# Polynomial run-time

- CanSim can be reduced to  $|Act_I|$  applications of CanFollow.
- CanFollow can be reduced to a linear program (Lemma 9).
- The foreach in split is trivially polynomial.
- For the while in split, note that there are at most  $O(\text{number of blocks})$  iterations.
- The same argument applies to the main GCPP algorithm.

# Example (ctd.)

- State space reduction:



- Winning strategy:



# Conclusion

- Probabilistic game structures are a relevant (interesting) class of structures.
- PA-I-simulations offer reductions that help answer (some) questions about a PGS.
- There is a polynomial algorithm that finds the strongest PA-I-simulation.

# Questions?