

Research Project Proposal: Abstractions in Extensive-Form Games

Jacopo Pio Gargano
jacopopio.gargano@mail.polimi.it
CSE Track



POLITECNICO
MILANO 1863



HP-SR
in Information Technology

Recreational games



Chess

Recreational games



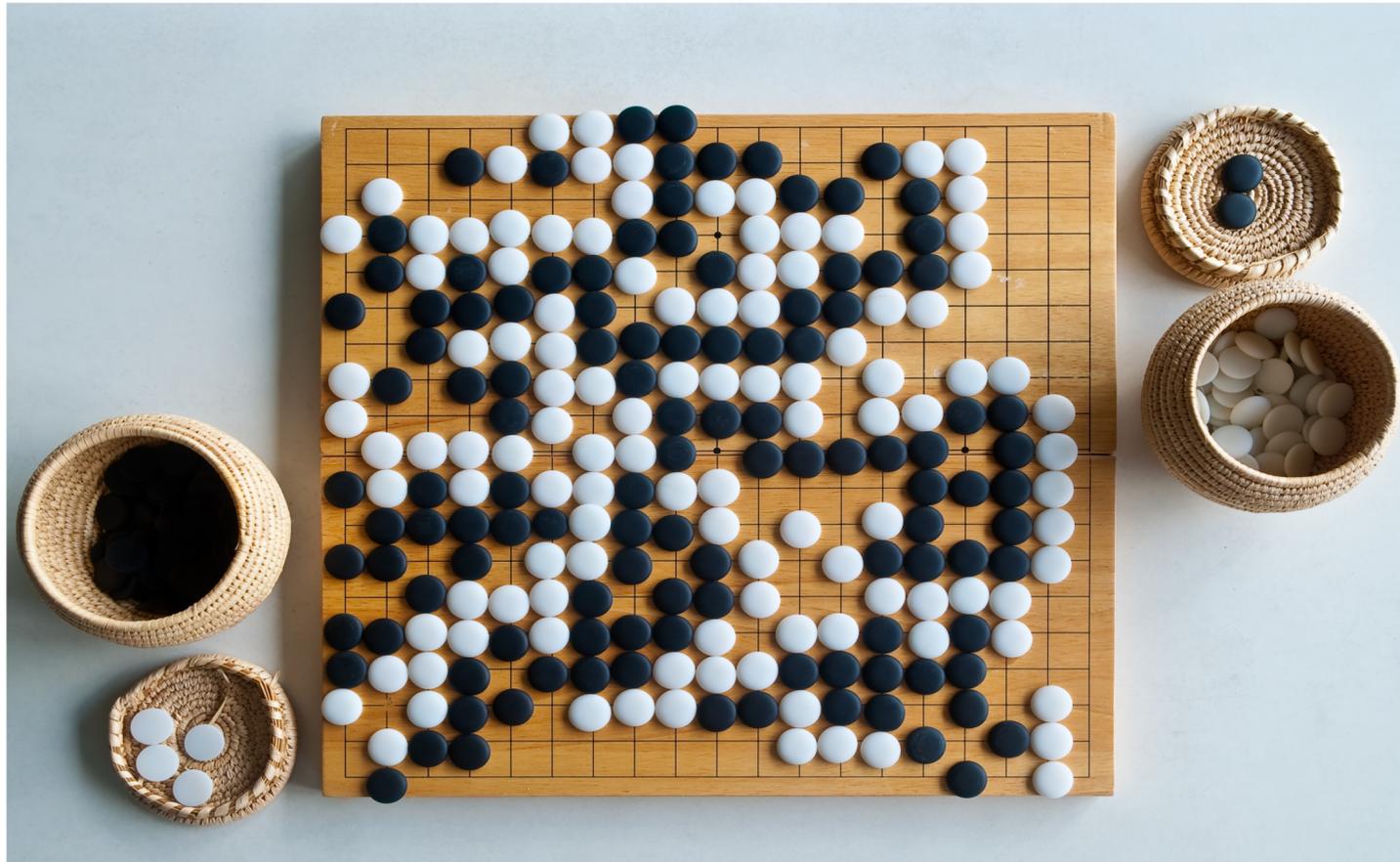
Chess



1996

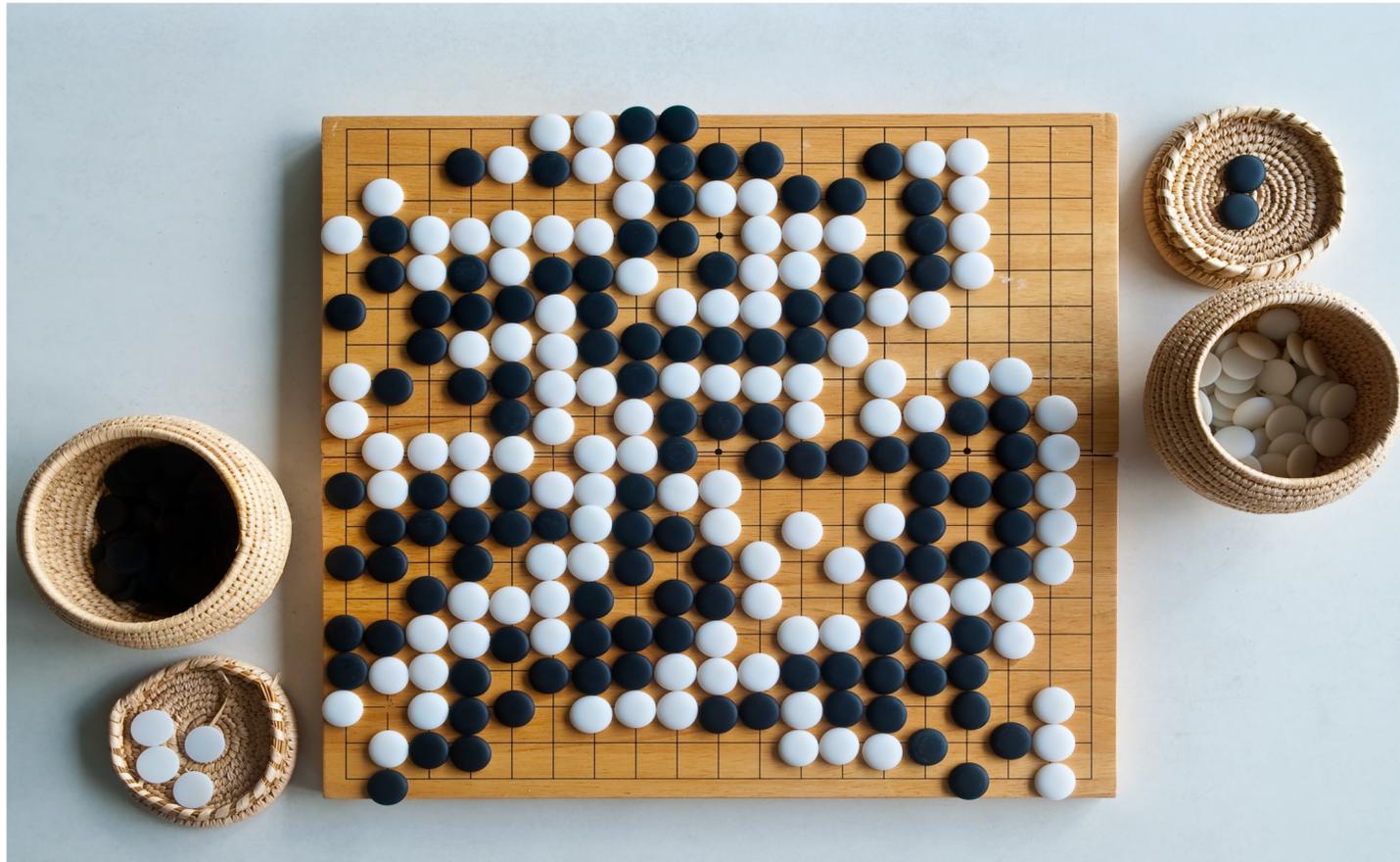
MiniMax with alpha-beta pruning search

Recreational games



Go

Recreational games



Go



2015



Monte Carlo tree search
Deep neural networks
Reinforcement Learning - self-play

Physical security



ARMOR - LAX airport



PROTECT - New York City port

Physical security



ARMOR - LAX airport



PROTECT - New York City port

Physical security



Wildlife poaching



Forest protection

Physical security



Wildlife poaching



Forest protection

Car racing



Defender



Attacker



Military



Attacker



Defender

Military



Attacker



Defender

Dogfighting

Game theory

- Theoretical framework for strategic interaction
- Mathematical models and algorithms (Algorithmic Game Theory)
- Conflict and cooperation
- Intelligent rational decision-makers
- Decisions influencing agents' welfare

Game theory

A *game* is a process consisting in:

- a set of players
- an initial situation
- rules that players must follow
- all possible final situations - outcomes
- the preferences of all the players - utilities

Game tree representation



Player 1



Player 2

Players

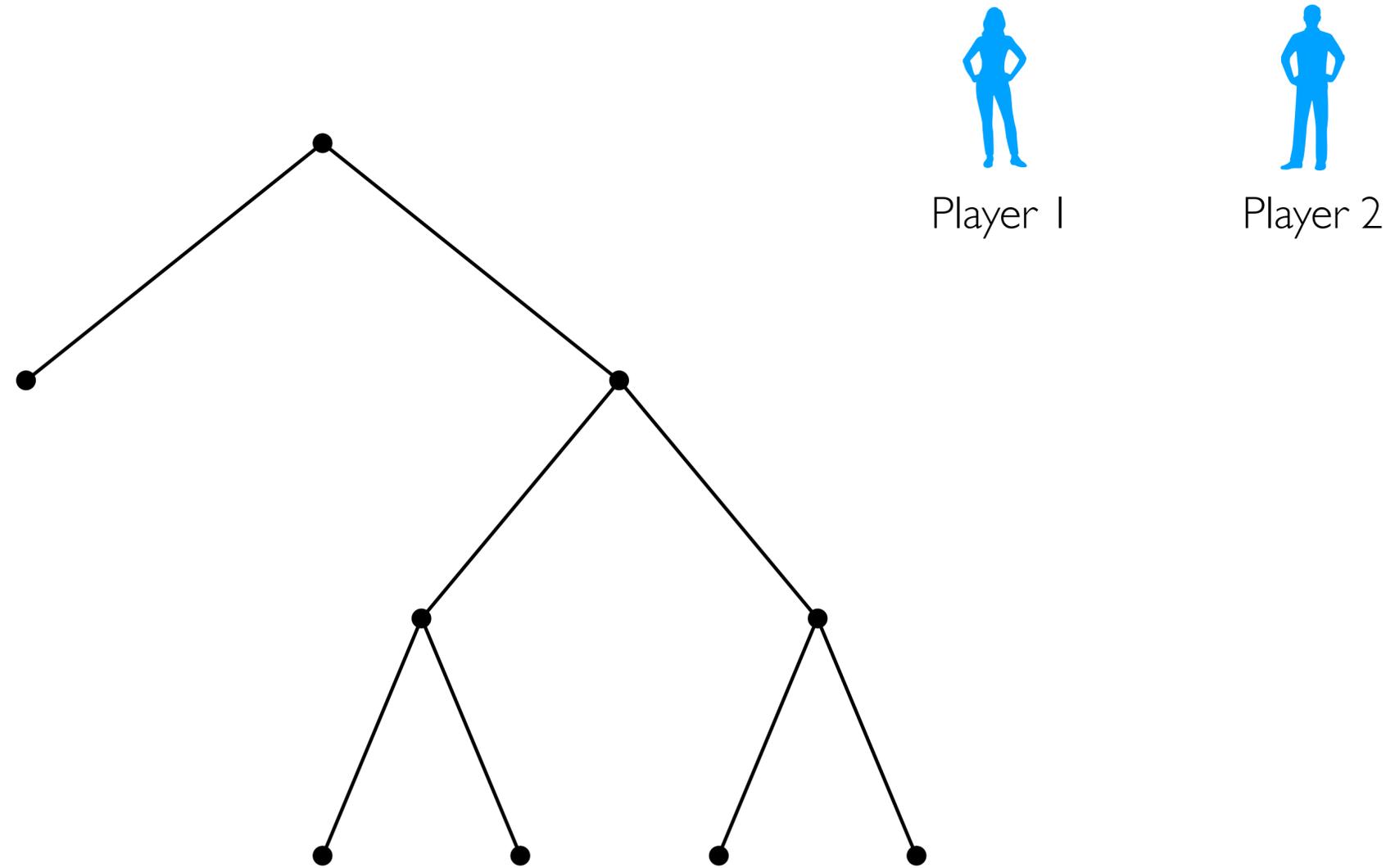
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

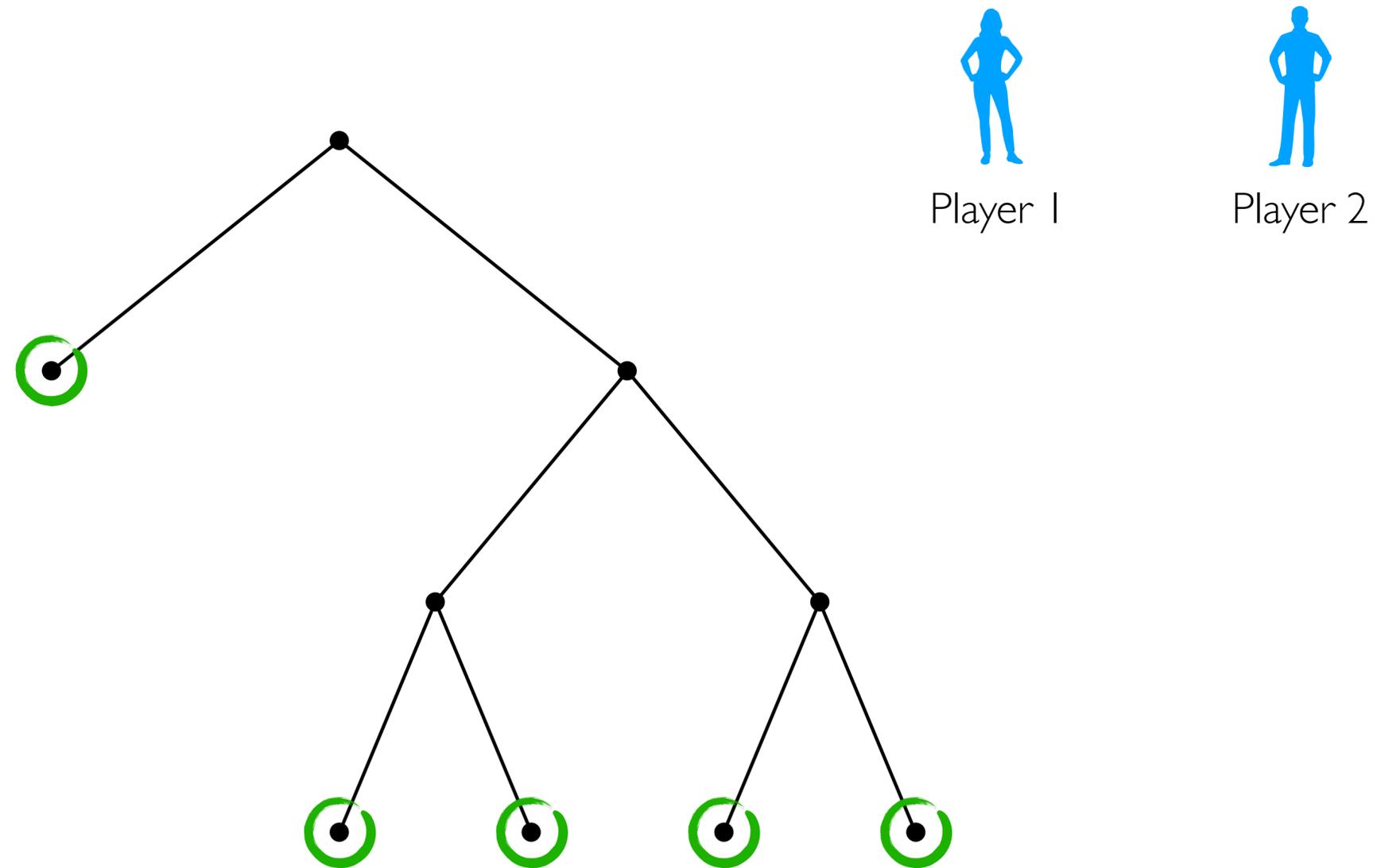
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

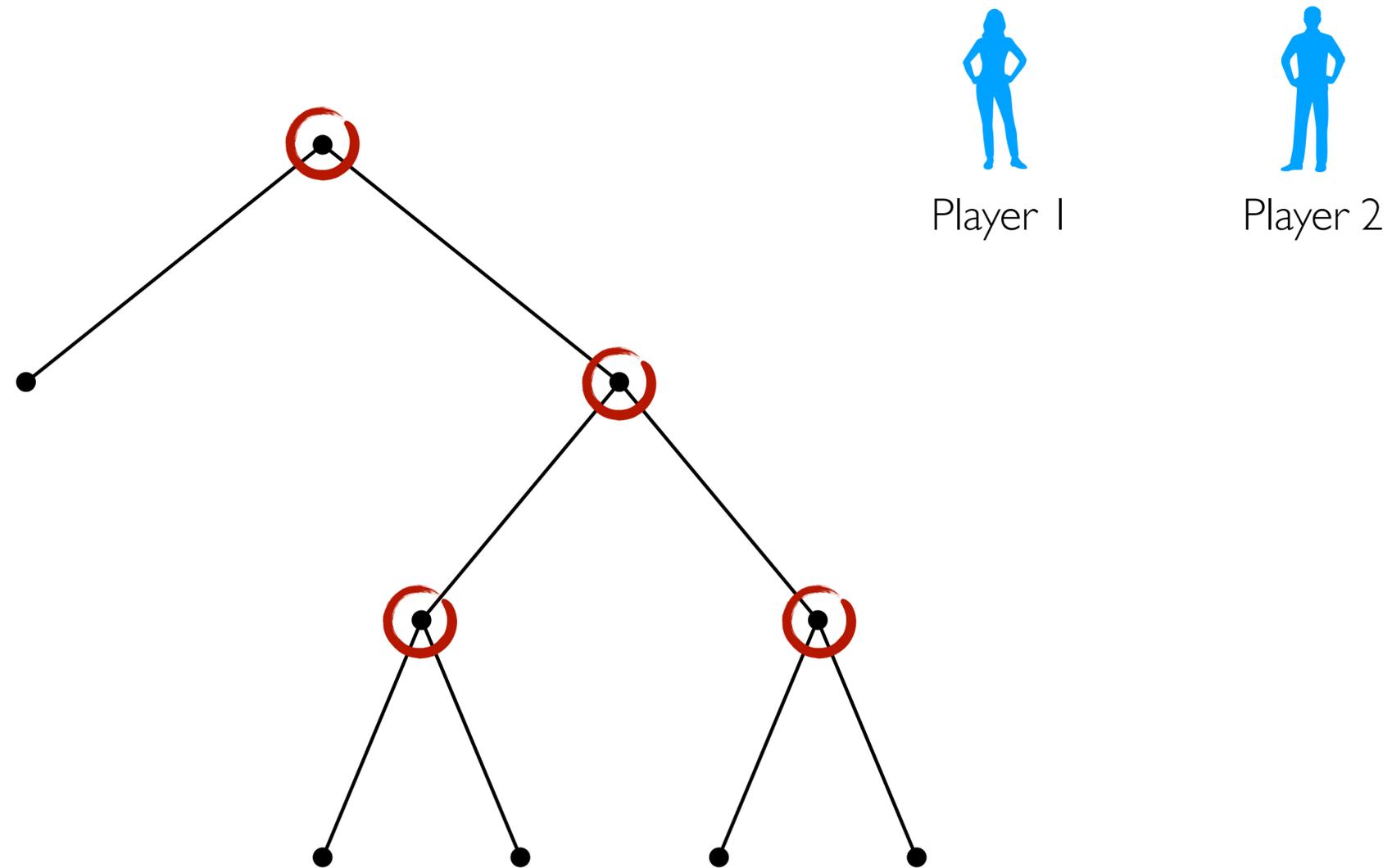
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

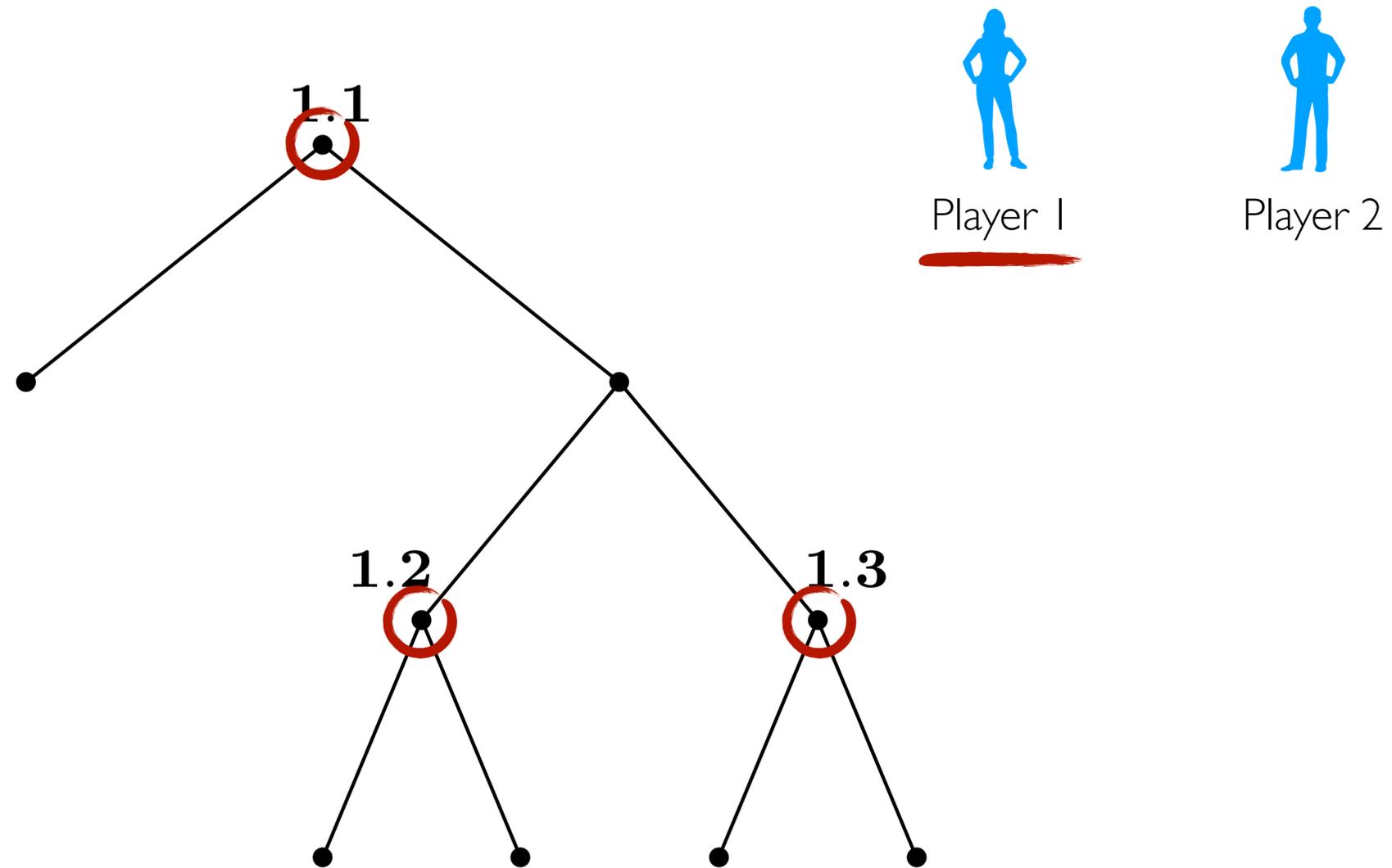
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

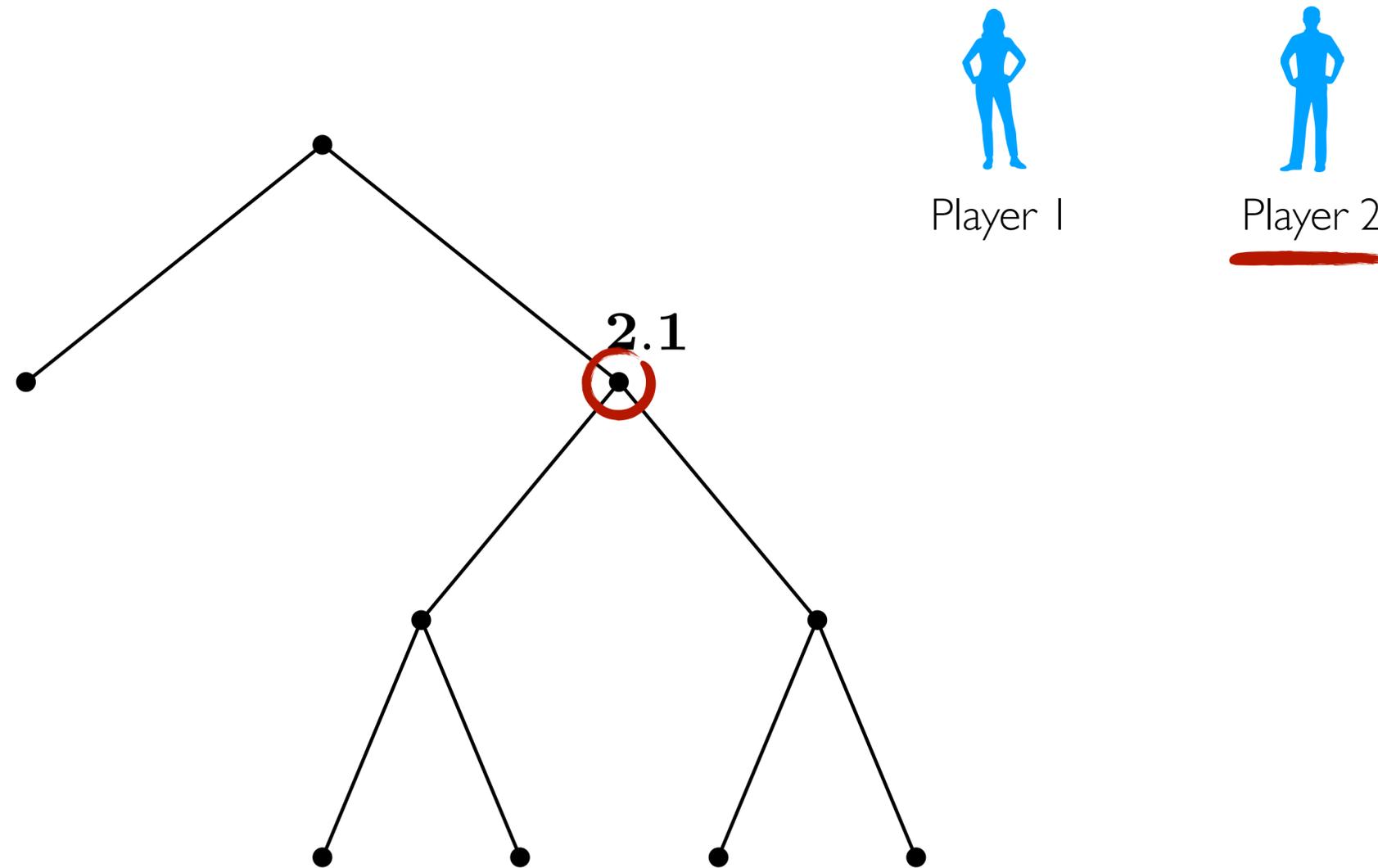
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

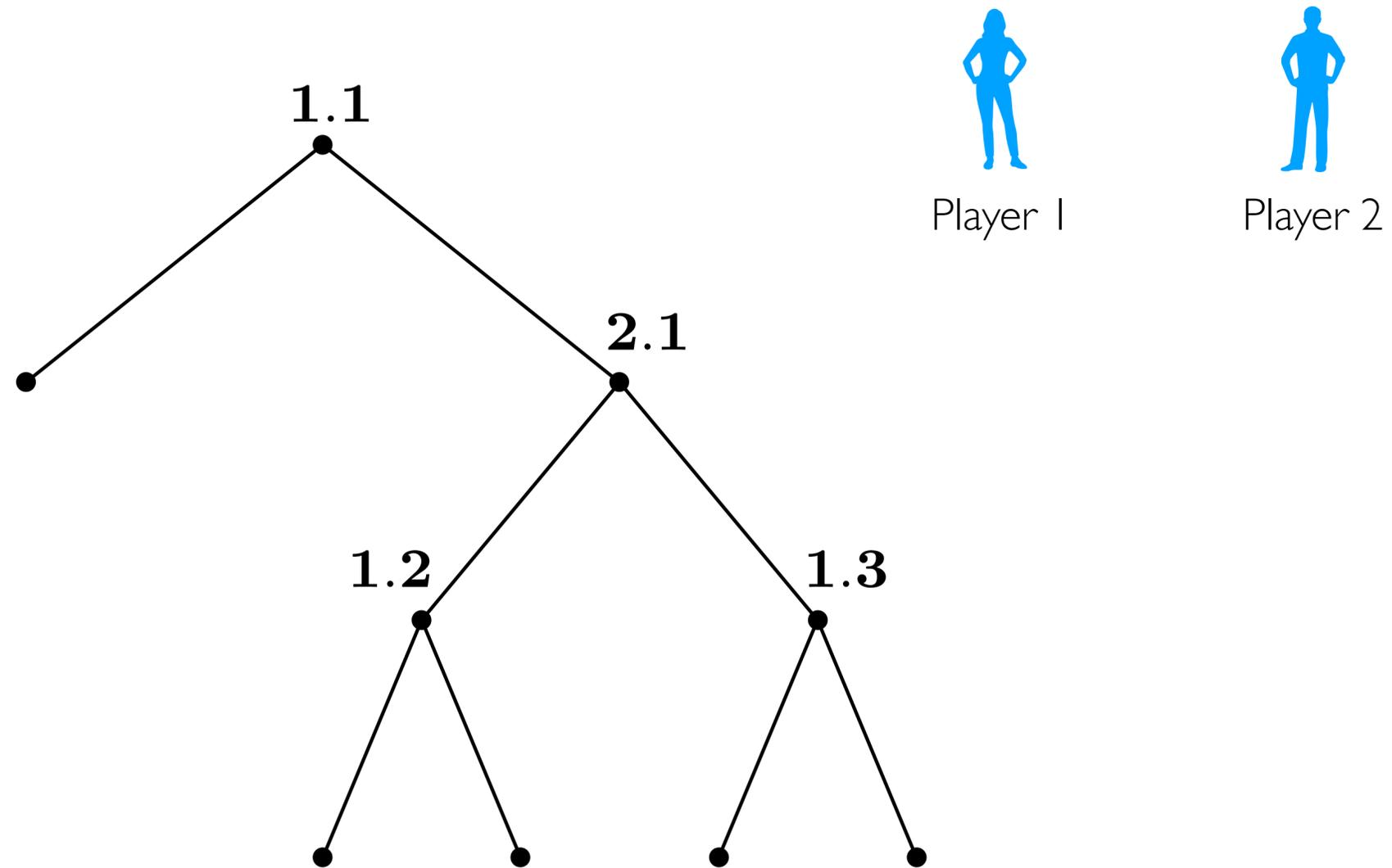
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

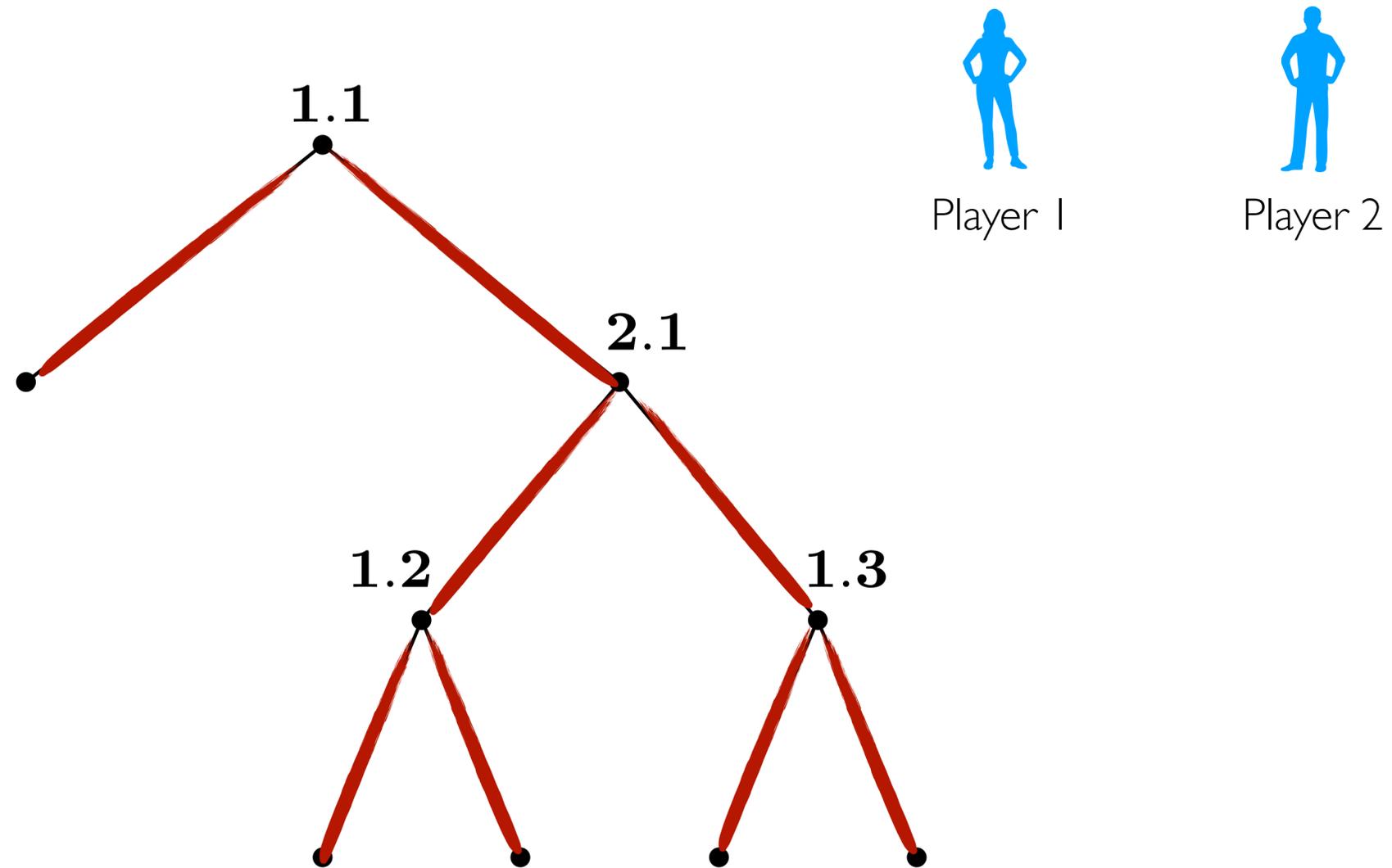
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

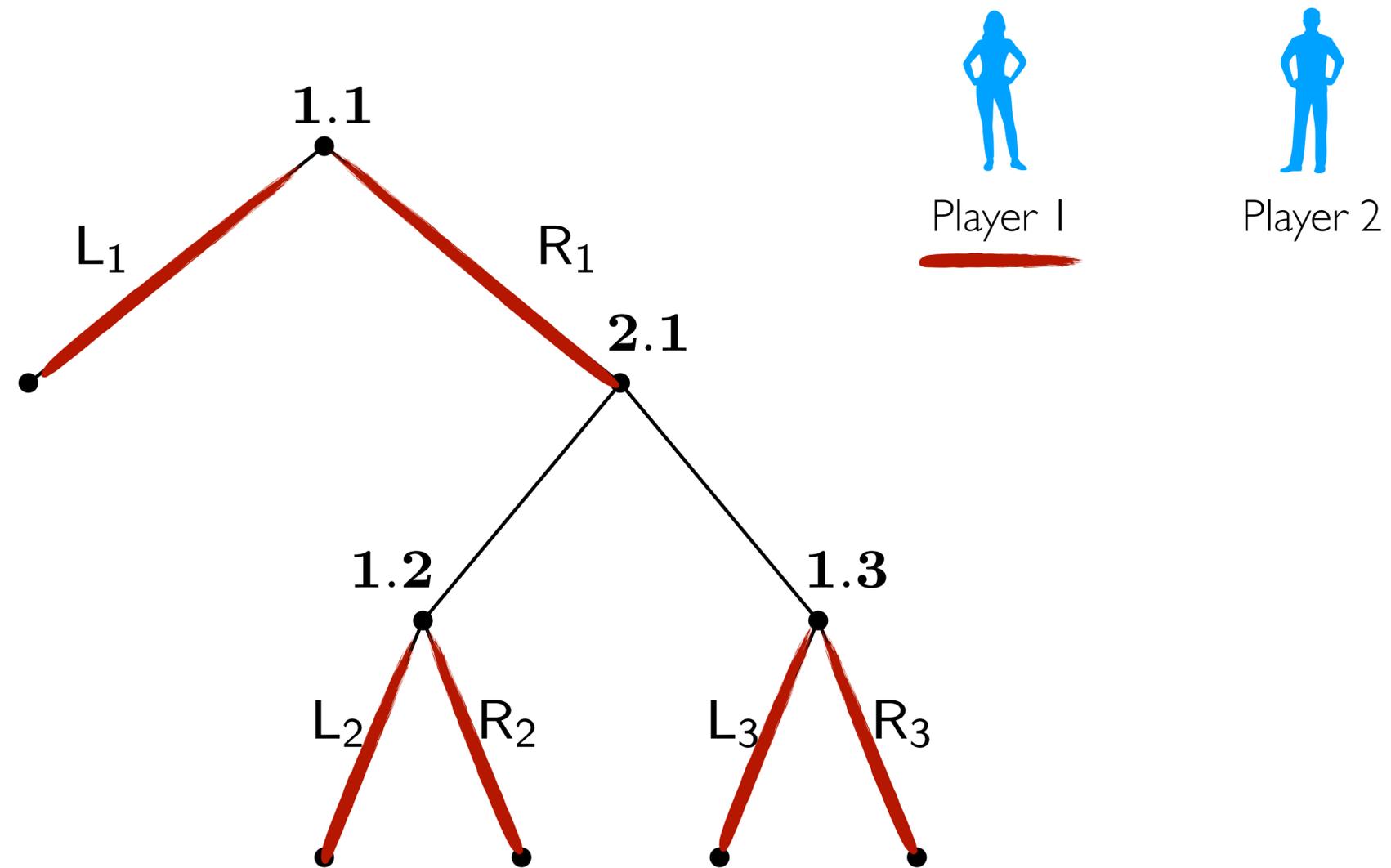
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

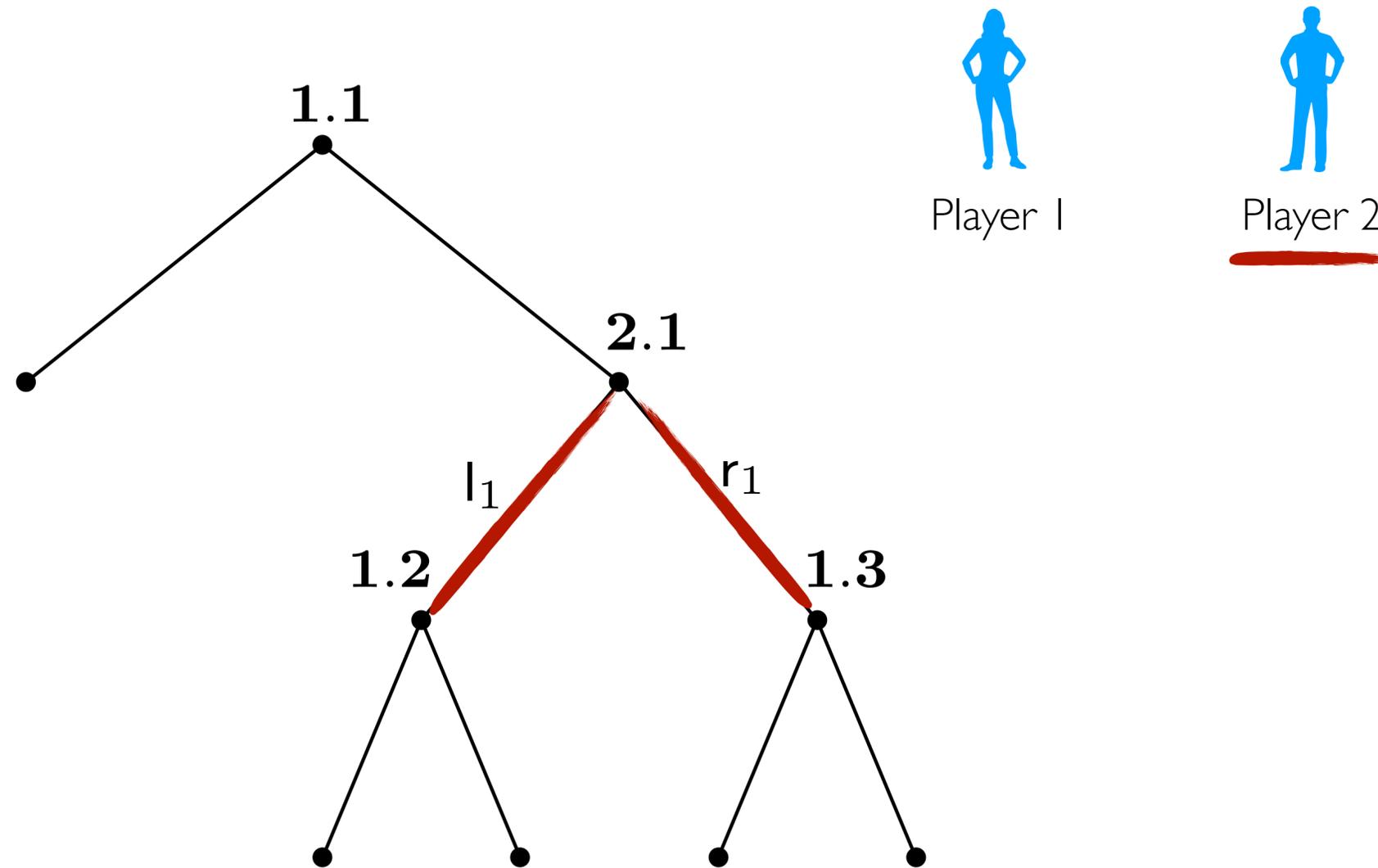
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

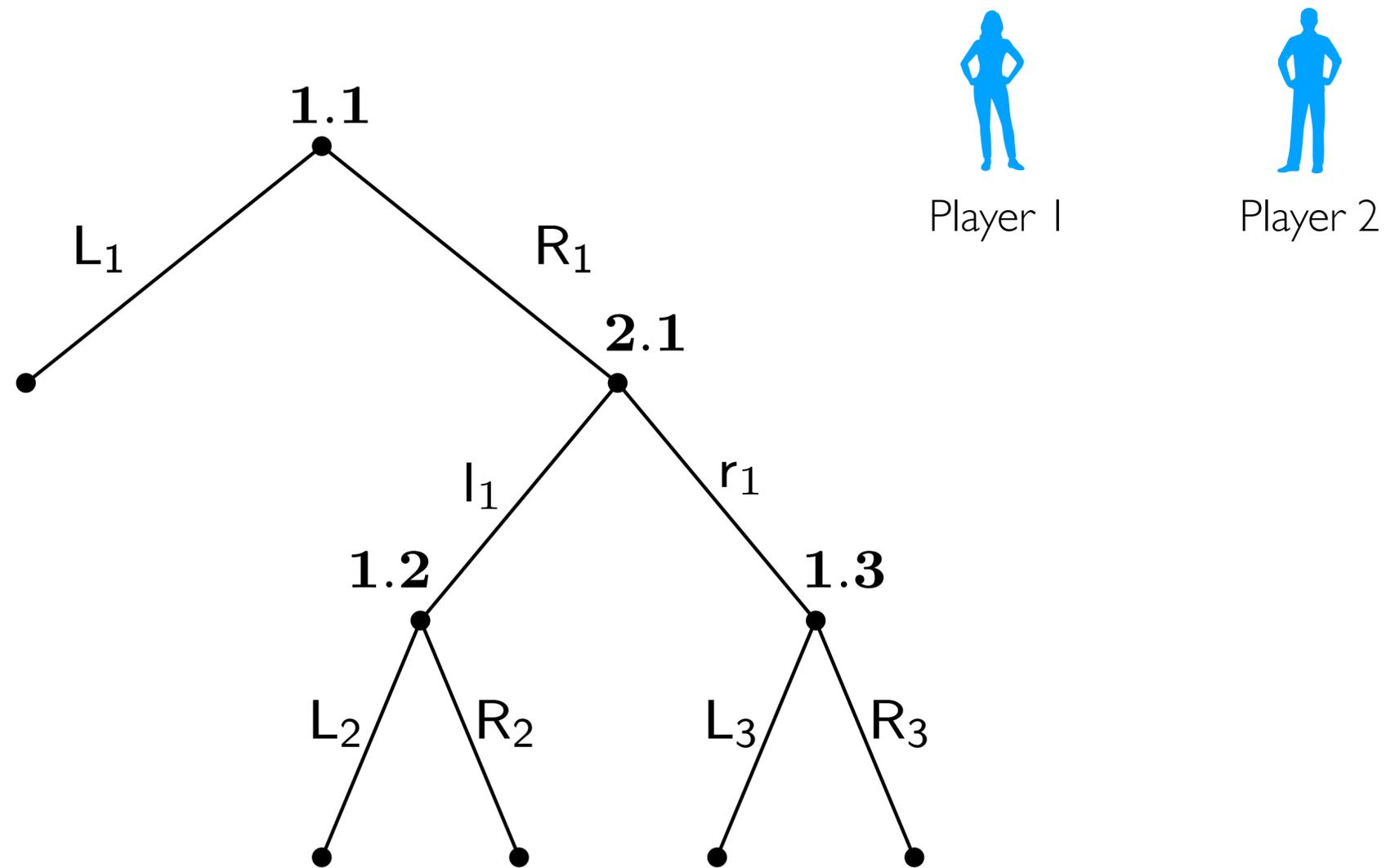
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

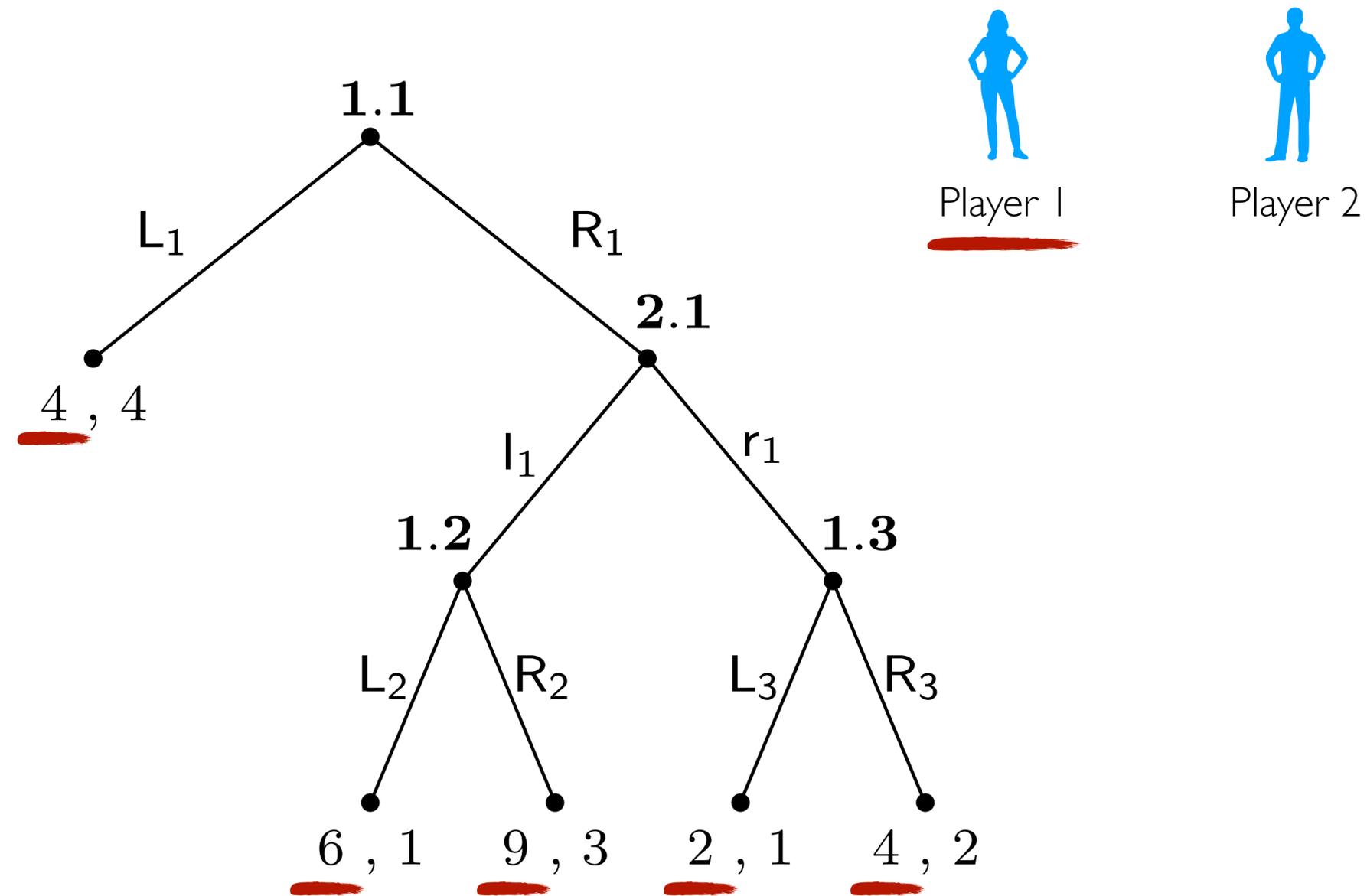
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

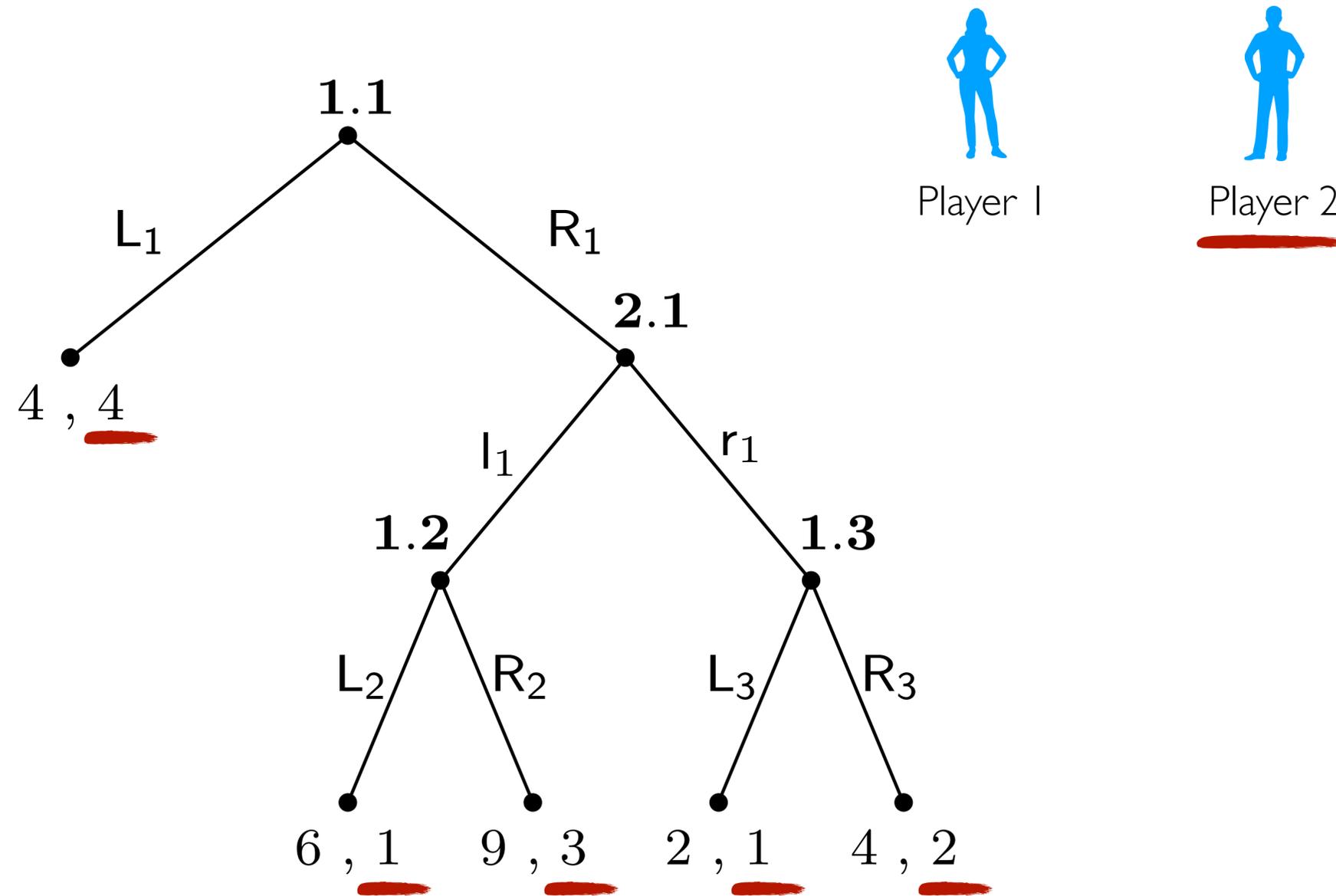
Terminal nodes

Decision nodes

Actions

Utilities

Game tree representation



Players

Terminal nodes

Decision nodes

Actions

Utilities

Formal model

$(N, A, V, T, \iota, \rho, \chi, U)$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of players $\{1, 2\}$
(Nature can be a player)



Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Set of **decision nodes** $\{1.1, 1.2, 1.3, 2.1\}$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Set of **decision nodes** $\{1.1, 1.2, 1.3, 2.1\}$

Set of **terminal nodes** $\{t_1, t_2, t_3, t_4, t_5\}$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Set of **decision nodes** $\{1.1, 1.2, 1.3, 2.1\}$

Set of **terminal nodes** $\{t_1, t_2, t_3, t_4, t_5\}$

Player function: $V \rightarrow N$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Set of **decision nodes** $\{1.1, 1.2, 1.3, 2.1\}$

Set of **terminal nodes** $\{t_1, t_2, t_3, t_4, t_5\}$

Next node function: $V \times A \longrightarrow V \cup T$

Player function: $V \longrightarrow N$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Set of **decision nodes** $\{1.1, 1.2, 1.3, 2.1\}$

Set of **terminal nodes** $\{t_1, t_2, t_3, t_4, t_5\}$

Action function: $V \longrightarrow 2^A$

Next node function: $V \times A \longrightarrow V \cup T$

Player function: $V \longrightarrow N$

Formal model

$$(N, A, V, T, \iota, \rho, \chi, U)$$

Set of **players** $\{1, 2\}$
(Nature can be a player)

Set of **actions** $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$

Set of **decision nodes** $\{1.1, 1.2, 1.3, 2.1\}$

Set of **terminal nodes** $\{t_1, t_2, t_3, t_4, t_5\}$

Set of **utility functions**: $\{U_1, U_2\}$

Action function: $V \longrightarrow 2^A$

Next node function: $V \times A \longrightarrow V \cup T$

Player function: $V \longrightarrow N$

Perfect vs imperfect information



Perfect information

Perfect vs imperfect information



Perfect information



Imperfect information

Perfect vs imperfect information



Perfect information



Imperfect information

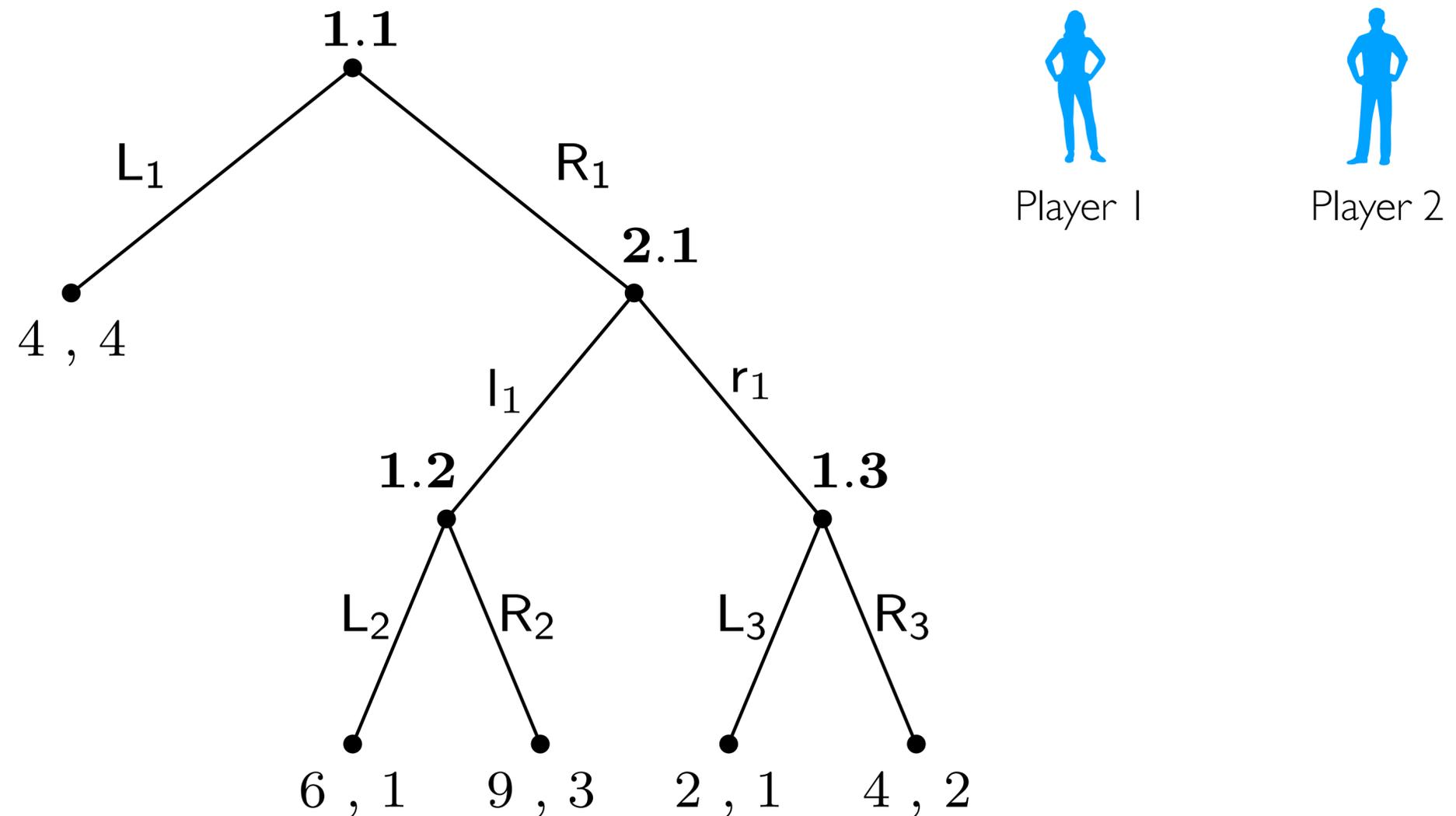
Information sets

Strategies

A function associating to each information set a probability distribution over the available actions

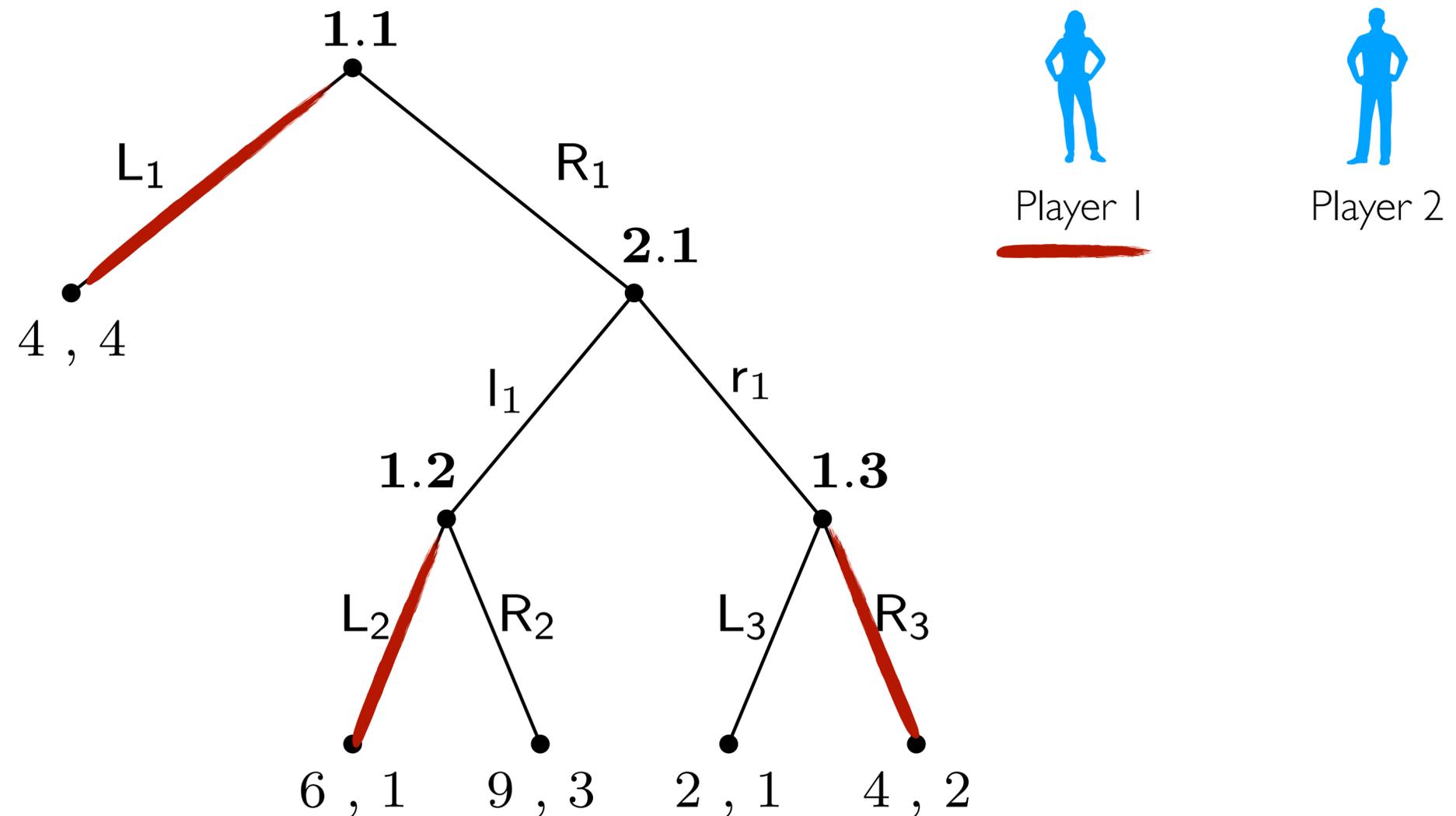
Strategies

A function associating to each information set a probability distribution over the available actions



Strategies

A function associating to each information set a probability distribution over the available actions



Nash equilibrium

A joint combination of strategies, stable w.r.t. unilateral deviations of a single player

Nash equilibrium

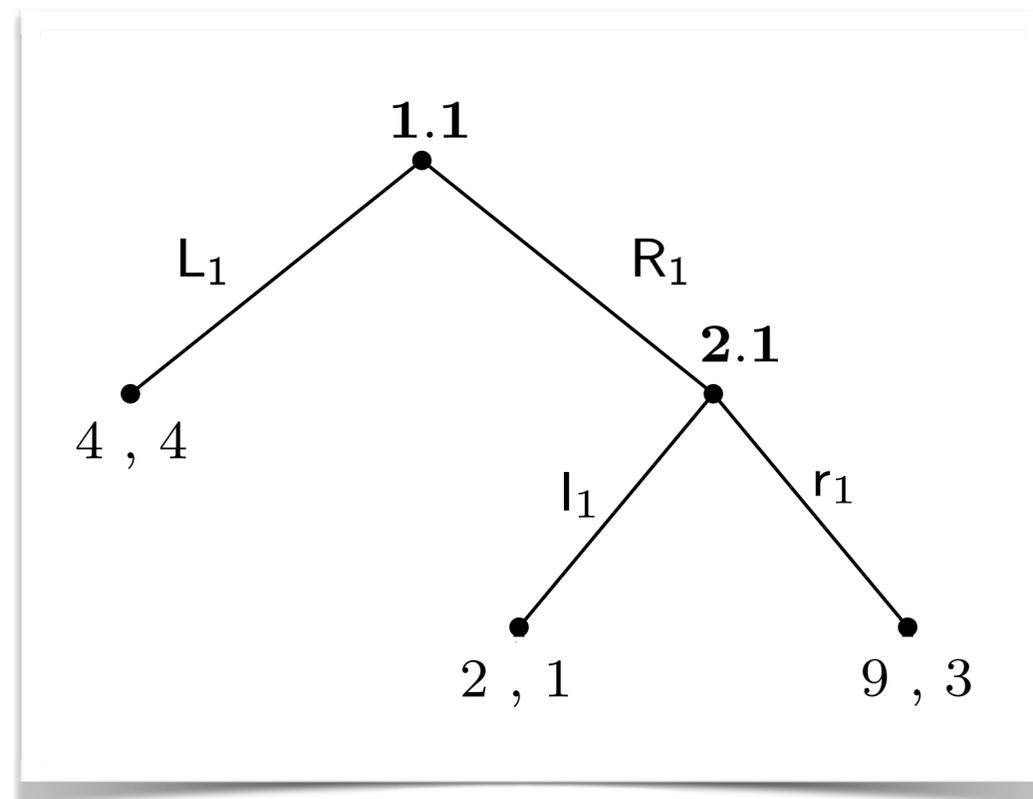
A joint combination of strategies, stable w.r.t. unilateral deviations of a single player



Player 1

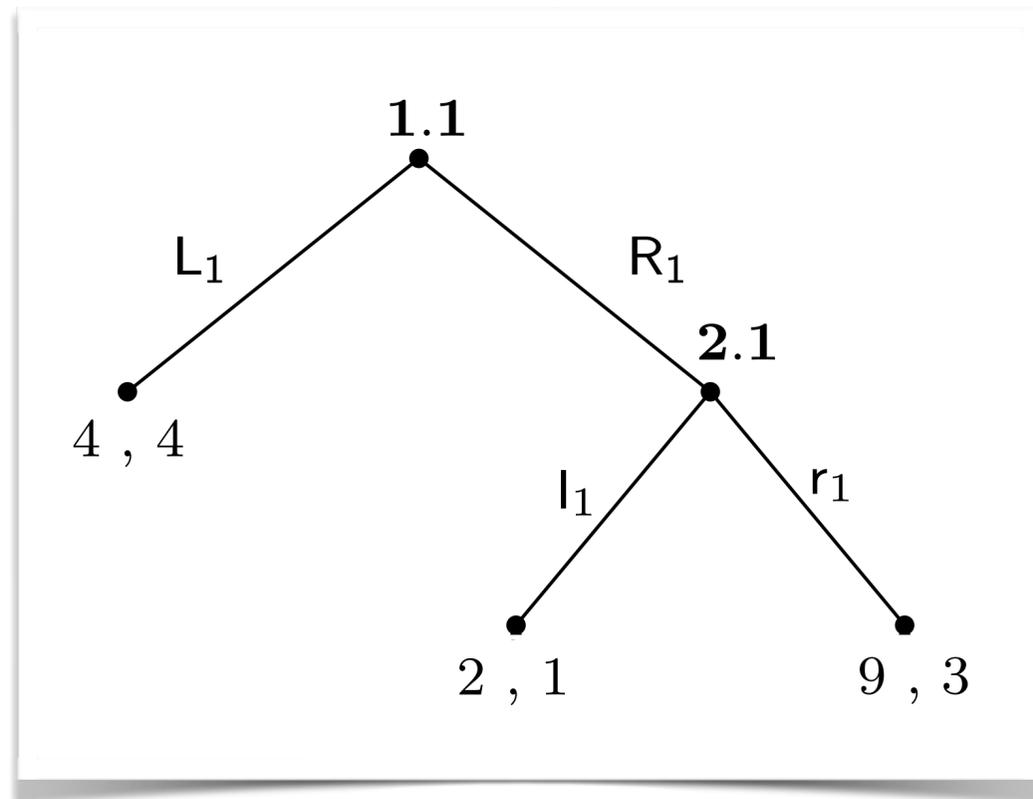


Player 2



Nash equilibrium

A joint combination of strategies, stable w.r.t. unilateral deviations of a single player

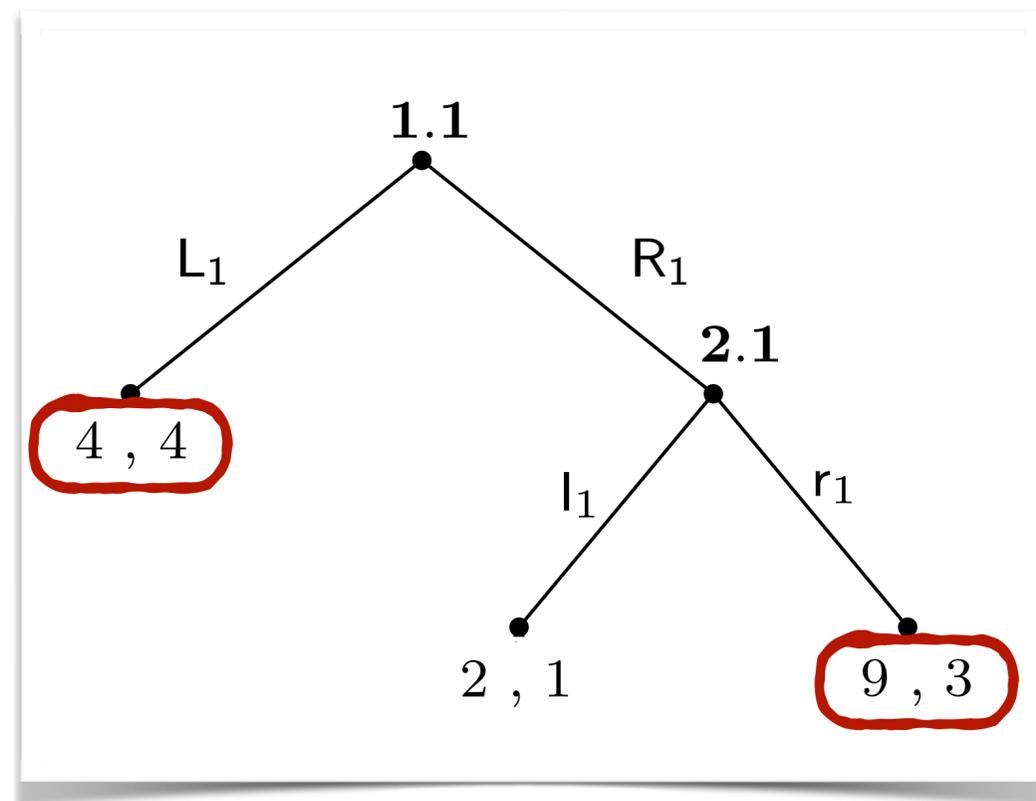


A normal form game matrix. The rows are labeled L₁ and R₁, and the columns are labeled l₁ and r₁. The payoffs are (Player 1, Player 2).

	Player 2	
	l ₁	r ₁
Player 1		
L ₁	4, 4	4, 4
R ₁	2, 1	9, 3

Nash equilibrium

A joint combination of strategies, stable w.r.t. unilateral deviations of a single player



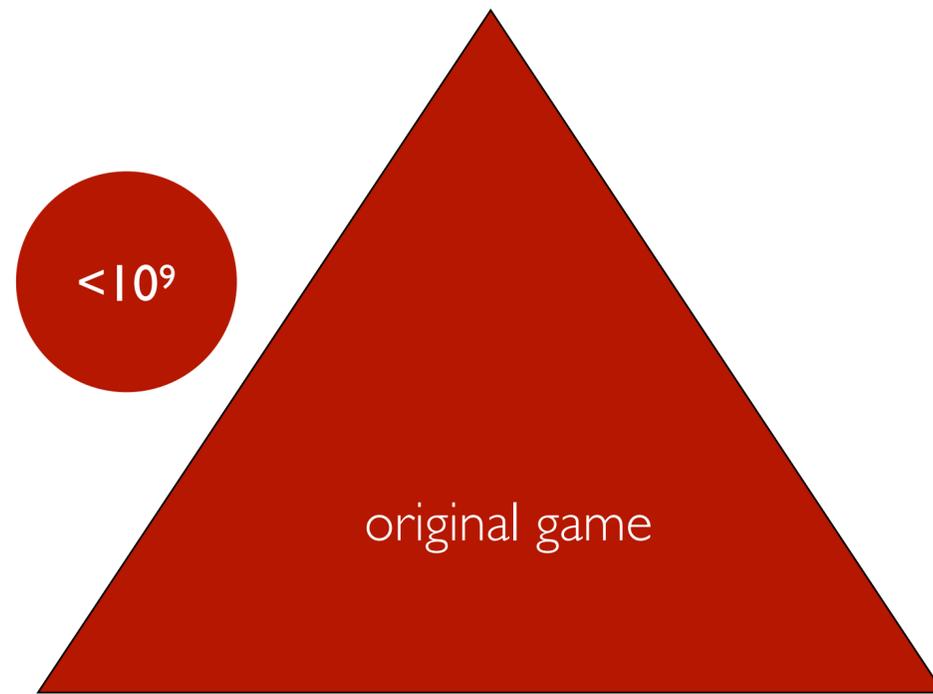
A normal form payoff matrix for the game. The rows represent Player 1's strategies (L_1 , R_1) and the columns represent Player 2's strategies (l_1 , r_1). The payoffs are shown in the cells, with the Nash equilibrium outcomes (4, 4) and (9, 3) highlighted in red.

	Player 2		
	l_1	r_1	
Player 1	L_1	$4, 4$	$4, 4$
R_1	$2, 1$	$9, 3$	

Nash equilibrium

- Mixed strategy NE
- Every n -player finite game has at least one Nash Equilibrium profile in mixed strategies
- epsilon-Nash Equilibrium

State-of-the-art game solving



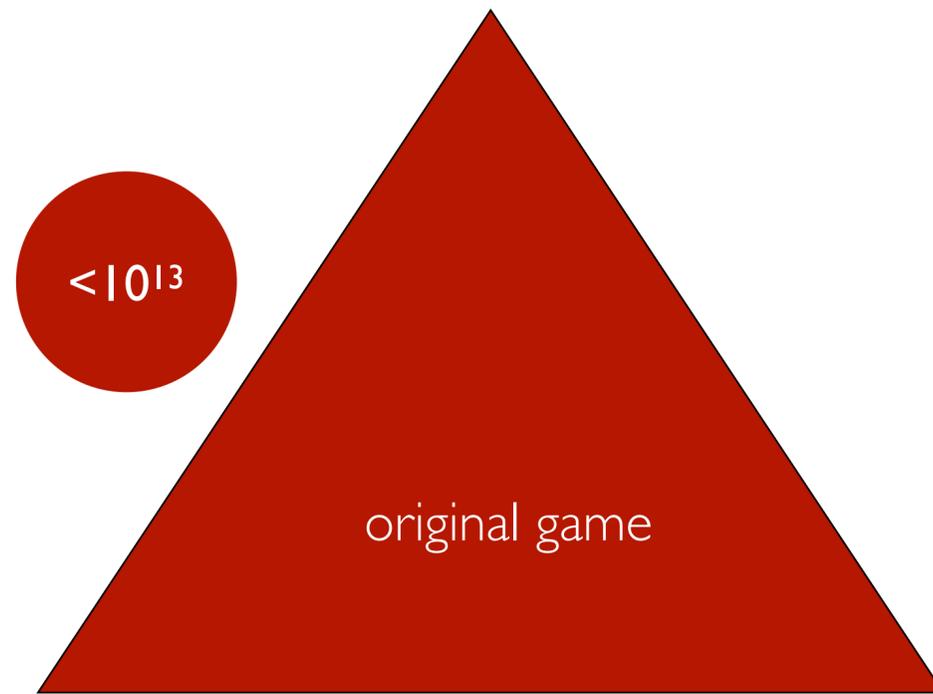
Linear programming

(Shi, Littman, 2000)

(Billings et al., 2003)

(Gilpin, Sandholm, 2007)

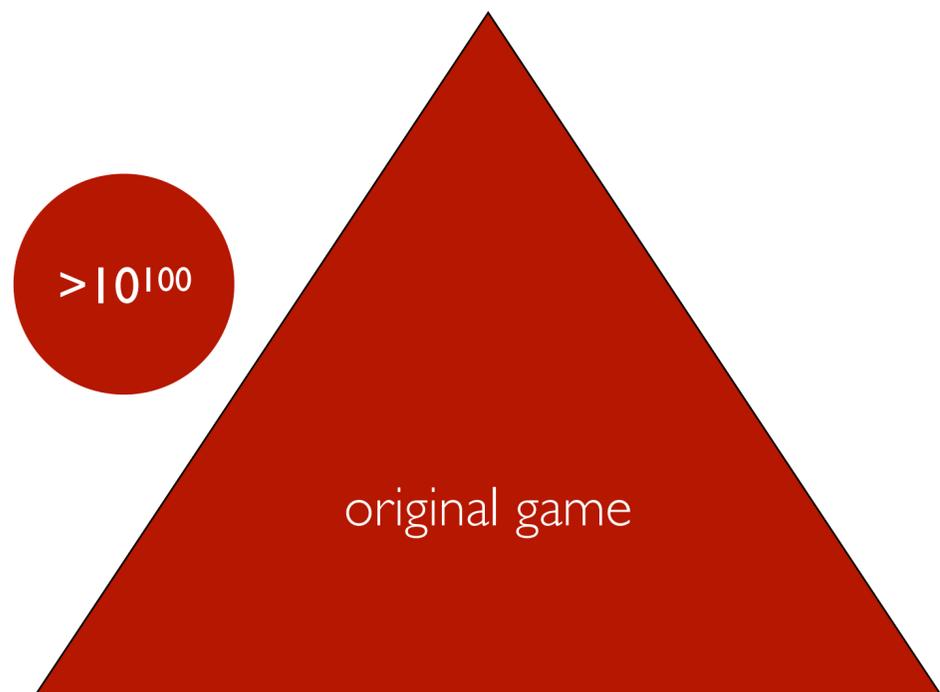
State-of-the-art game solving



No-regret learning

Zinkevich et al., 2008

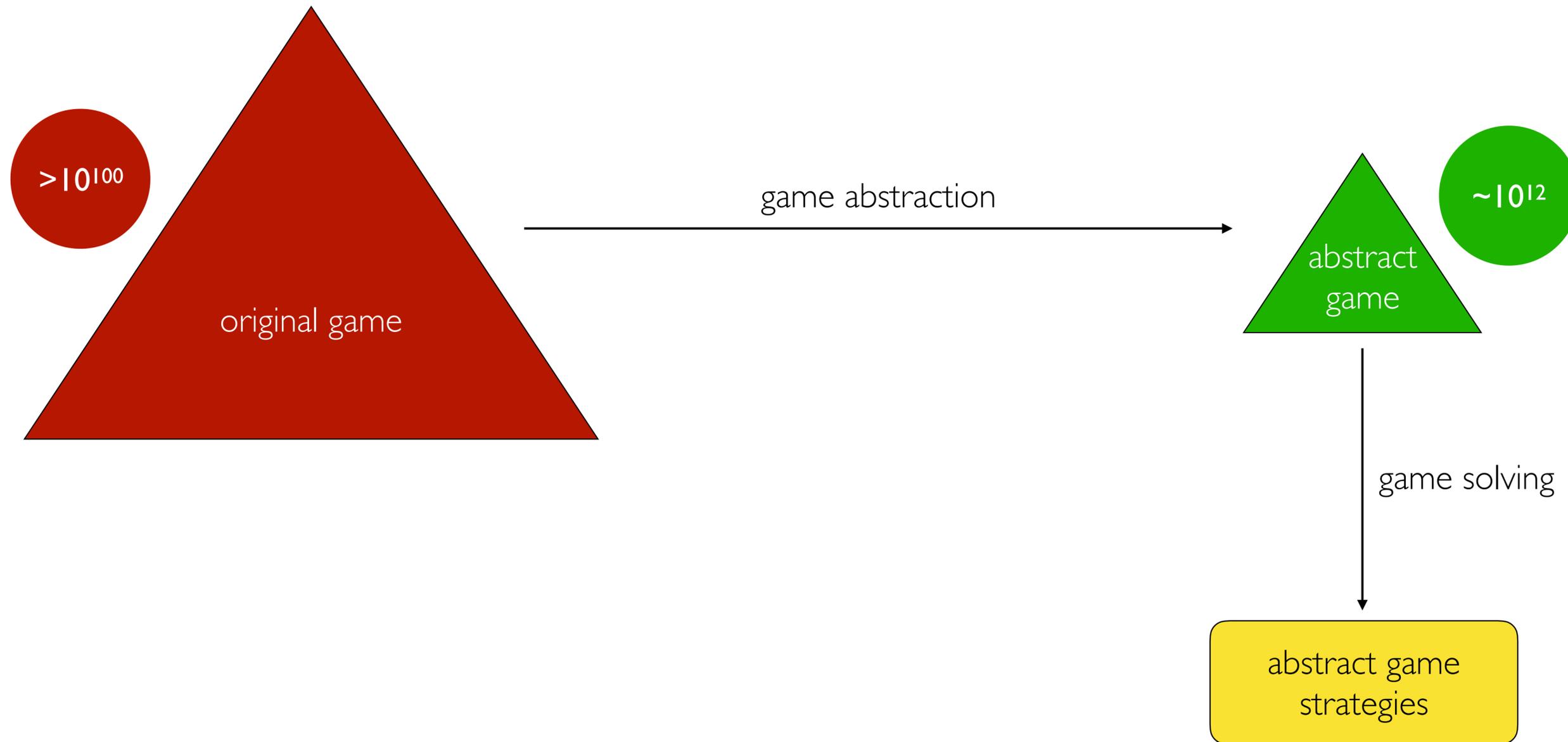
State-of-the-art huge-game solving: *preplay*



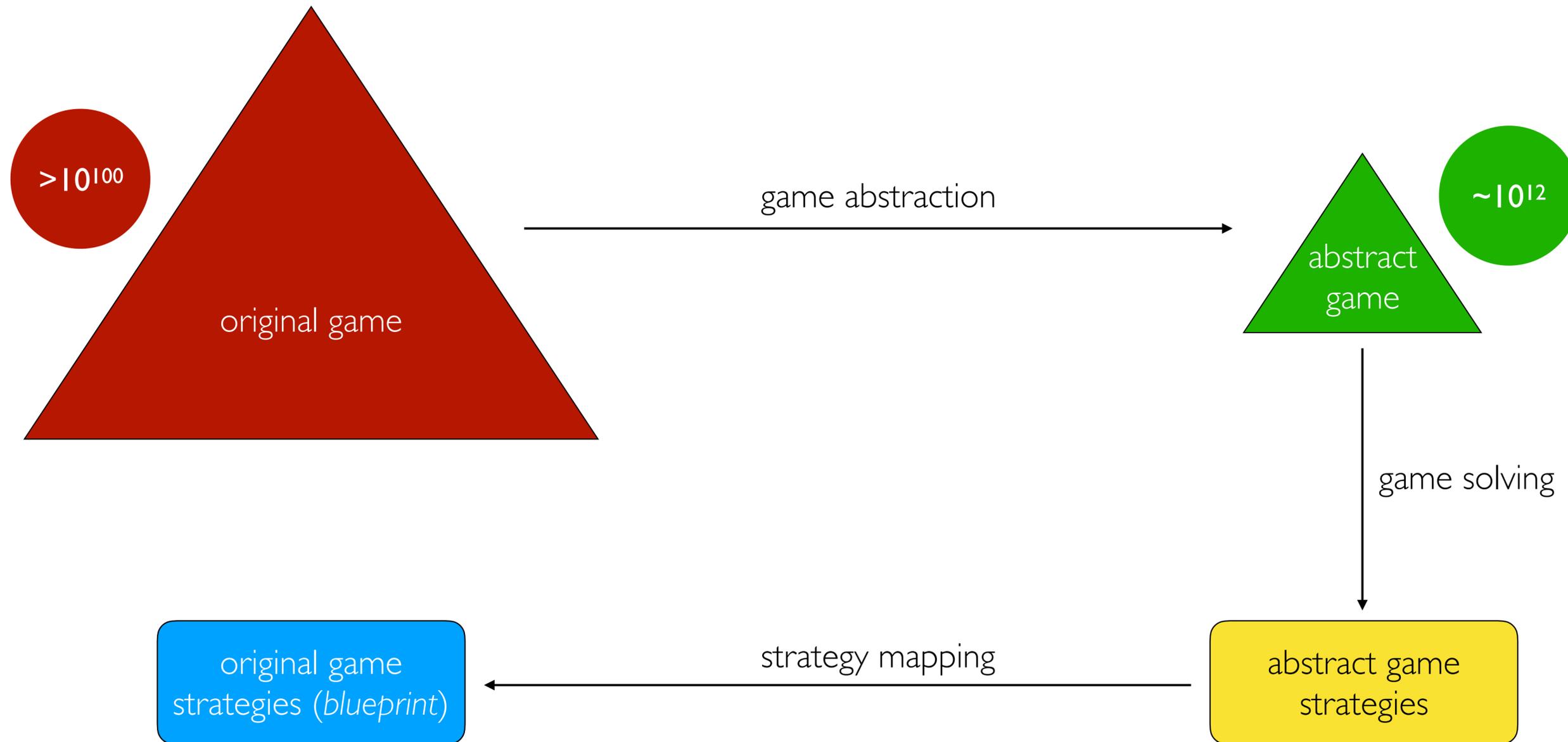
State-of-the-art huge-game solving: *preplay*



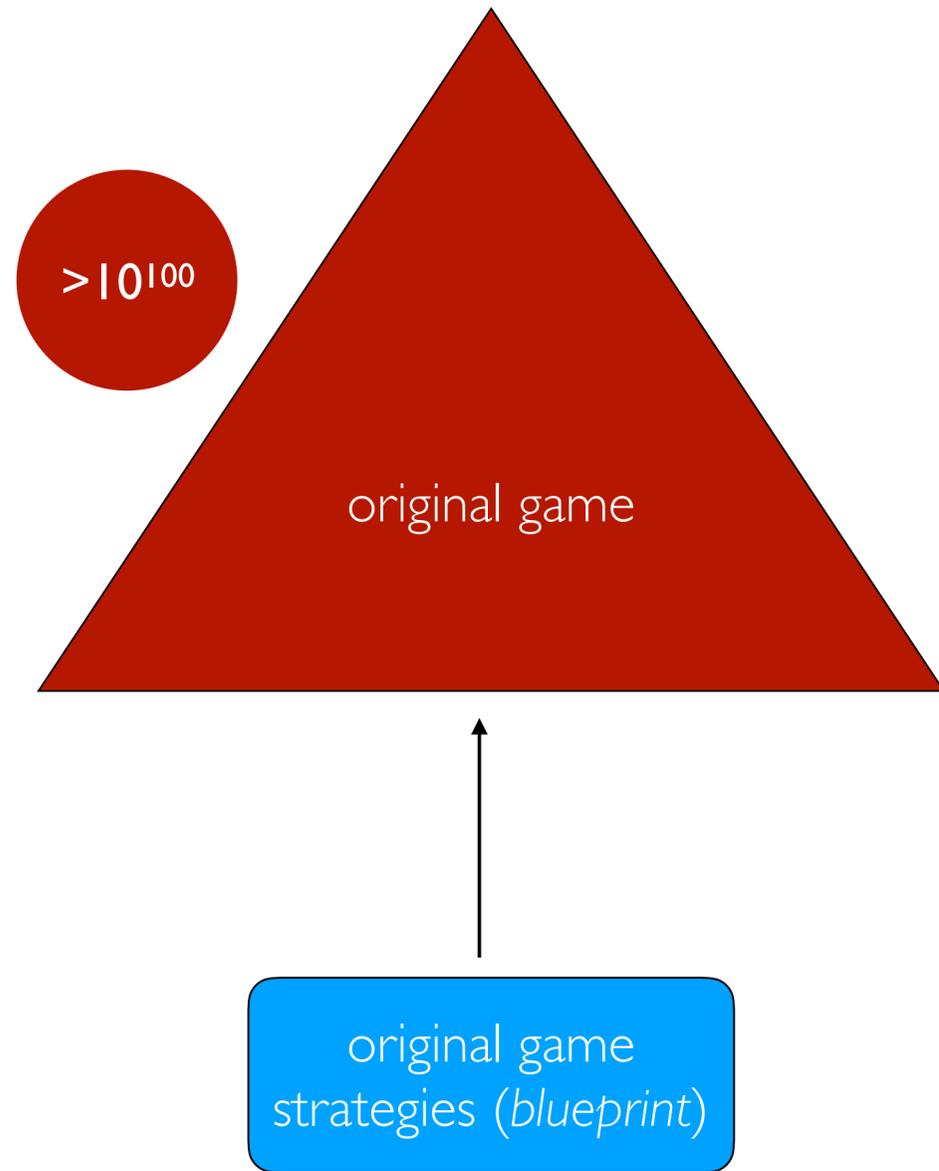
State-of-the-art huge-game solving: *preplay*



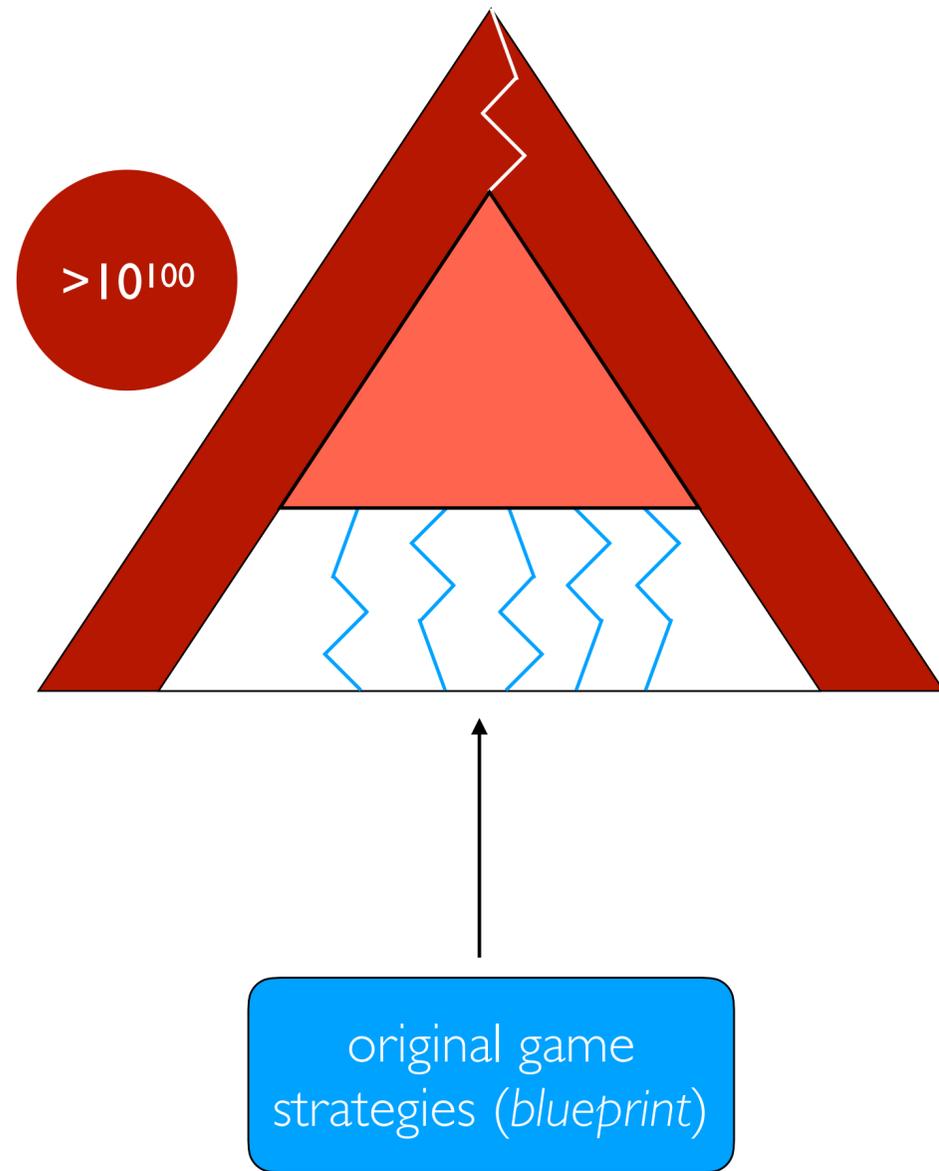
State-of-the-art huge-game solving: *preplay*



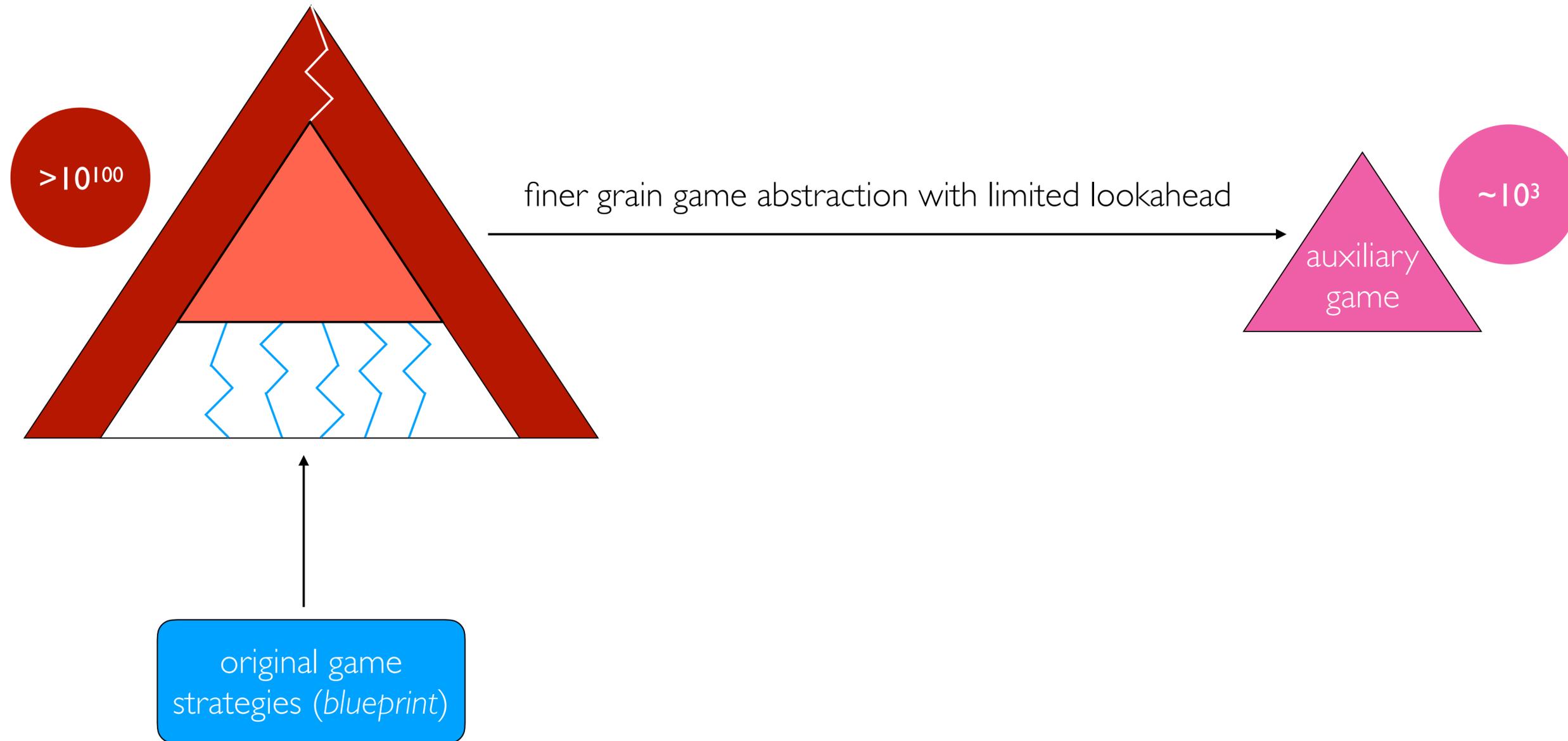
State-of-the-art huge-game solving: *play*



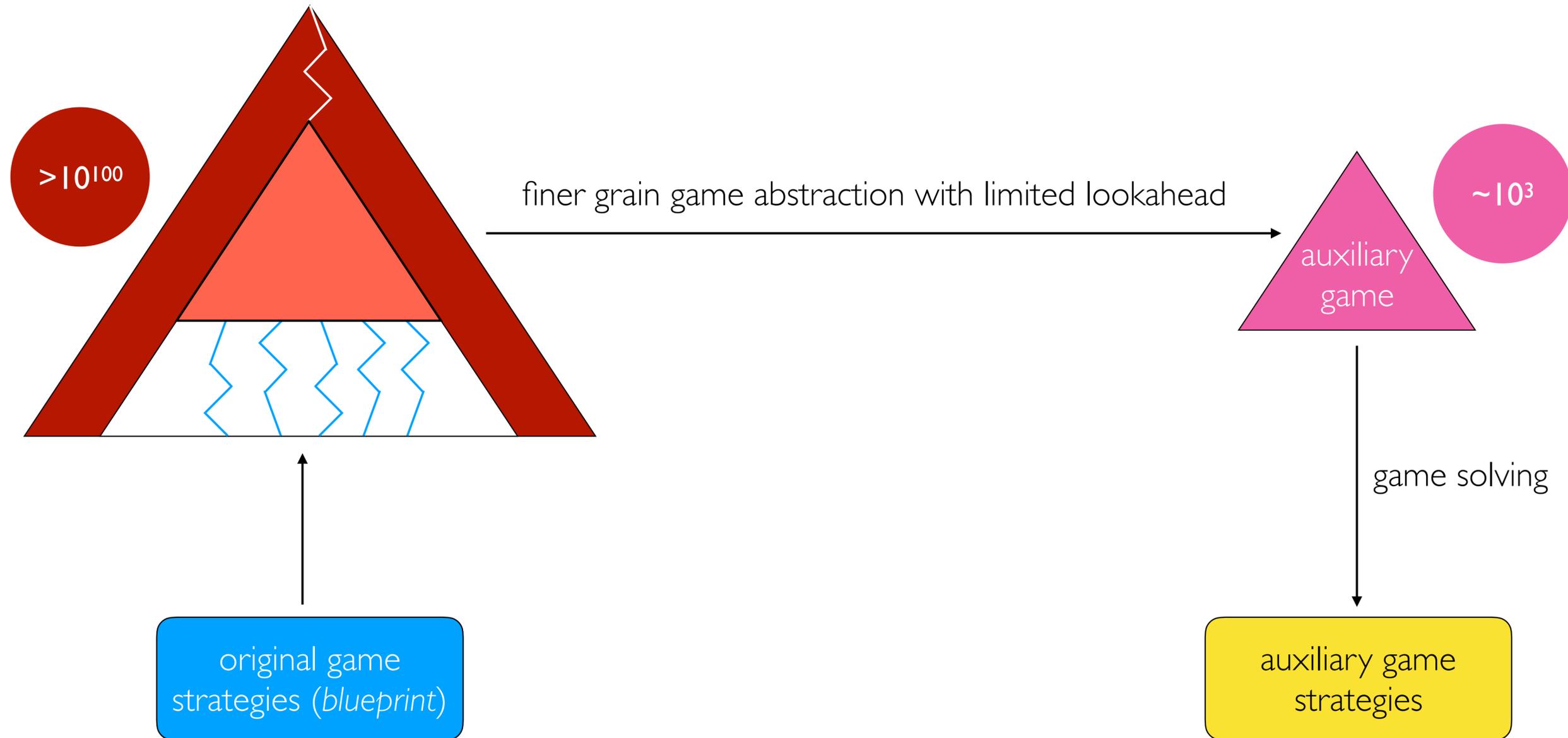
State-of-the-art huge-game solving: *play*



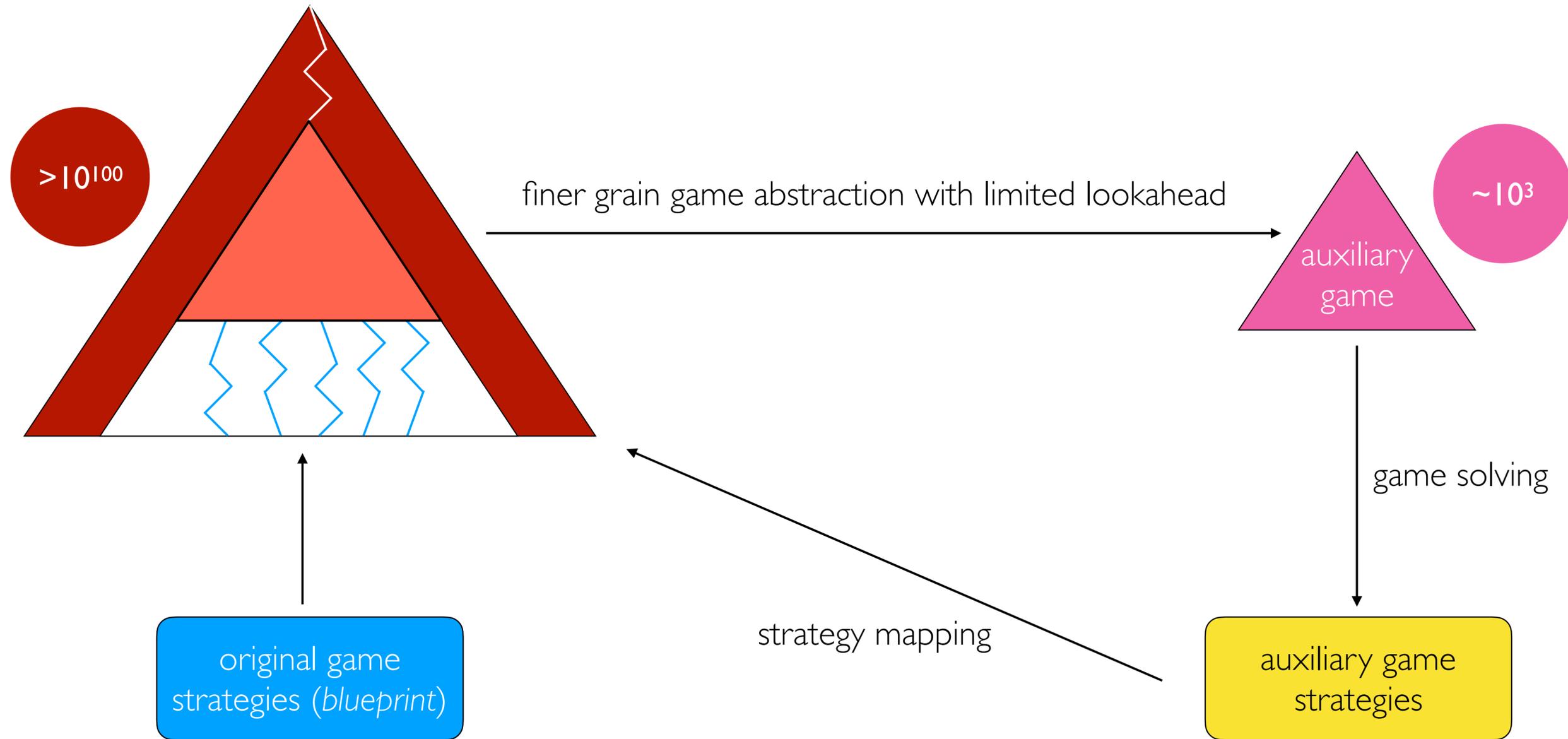
State-of-the-art huge-game solving: *play*



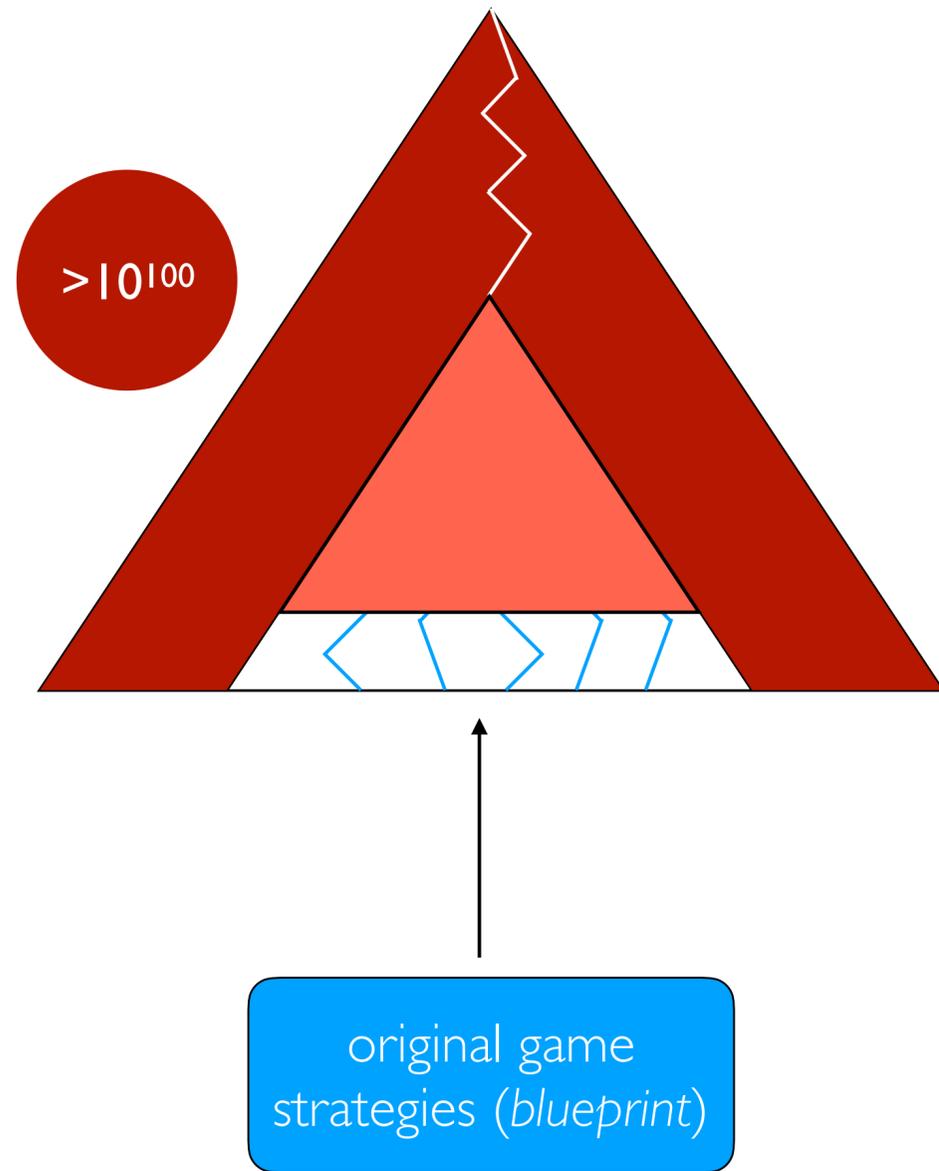
State-of-the-art huge-game solving: *play*



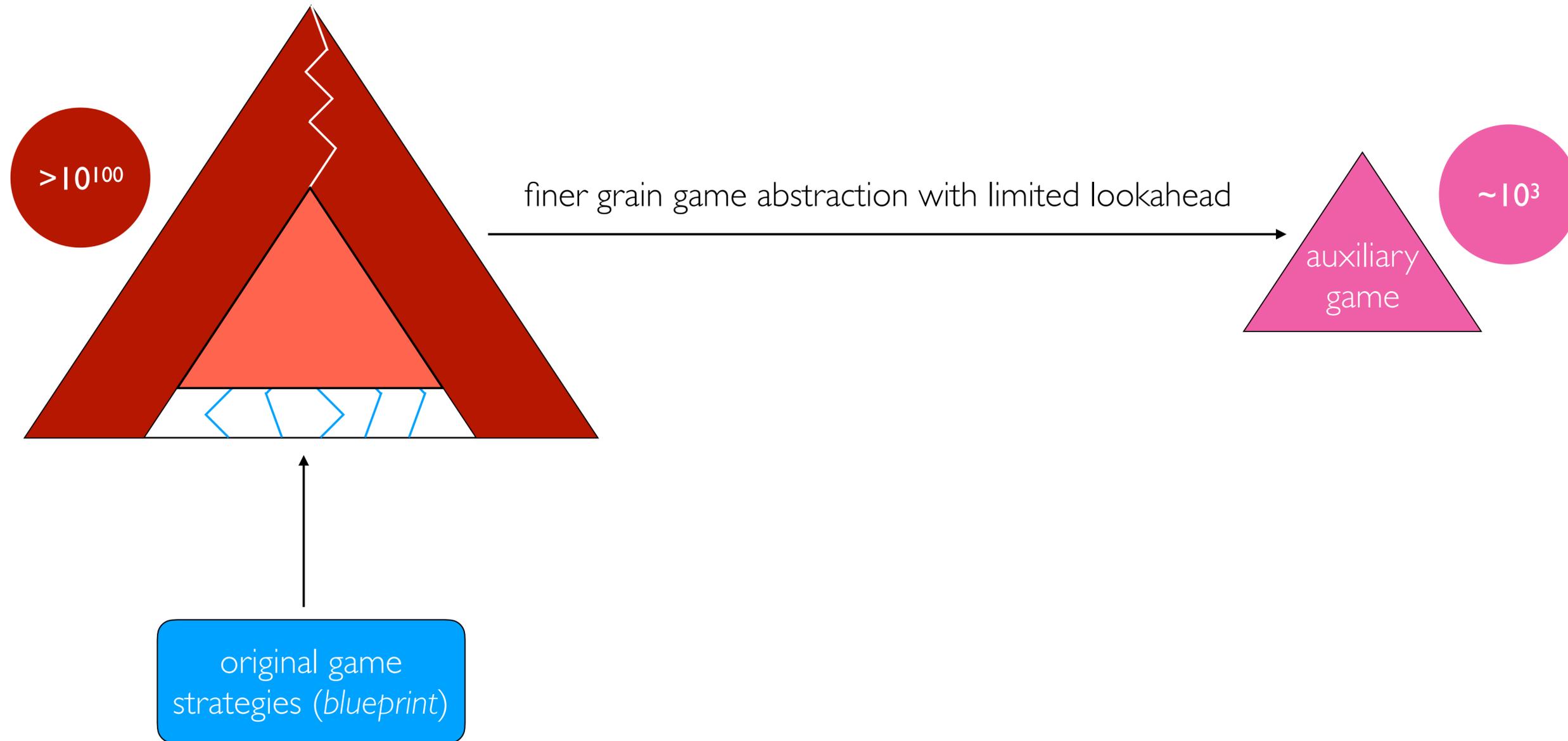
State-of-the-art huge-game solving: *play*



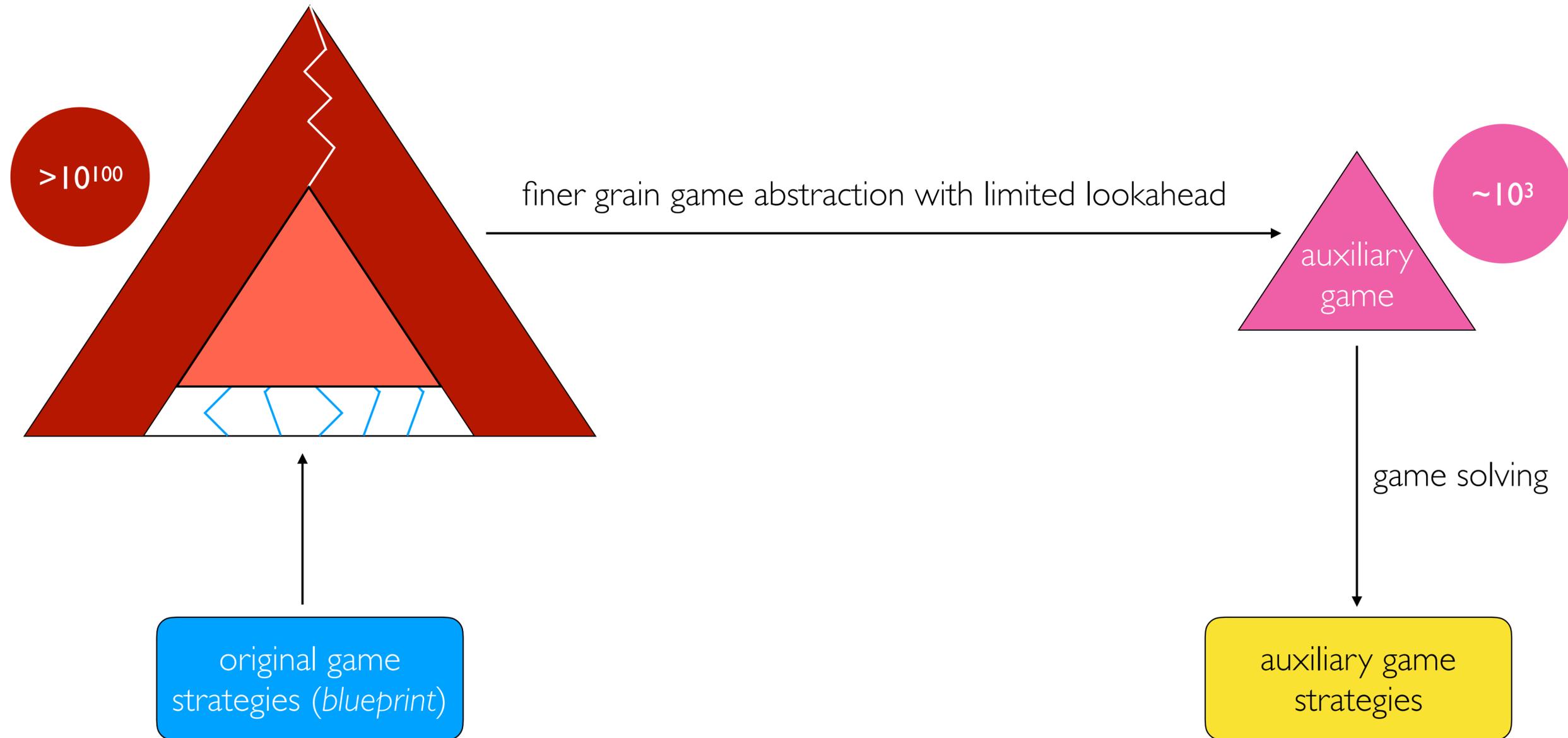
State-of-the-art huge-game solving: *play*



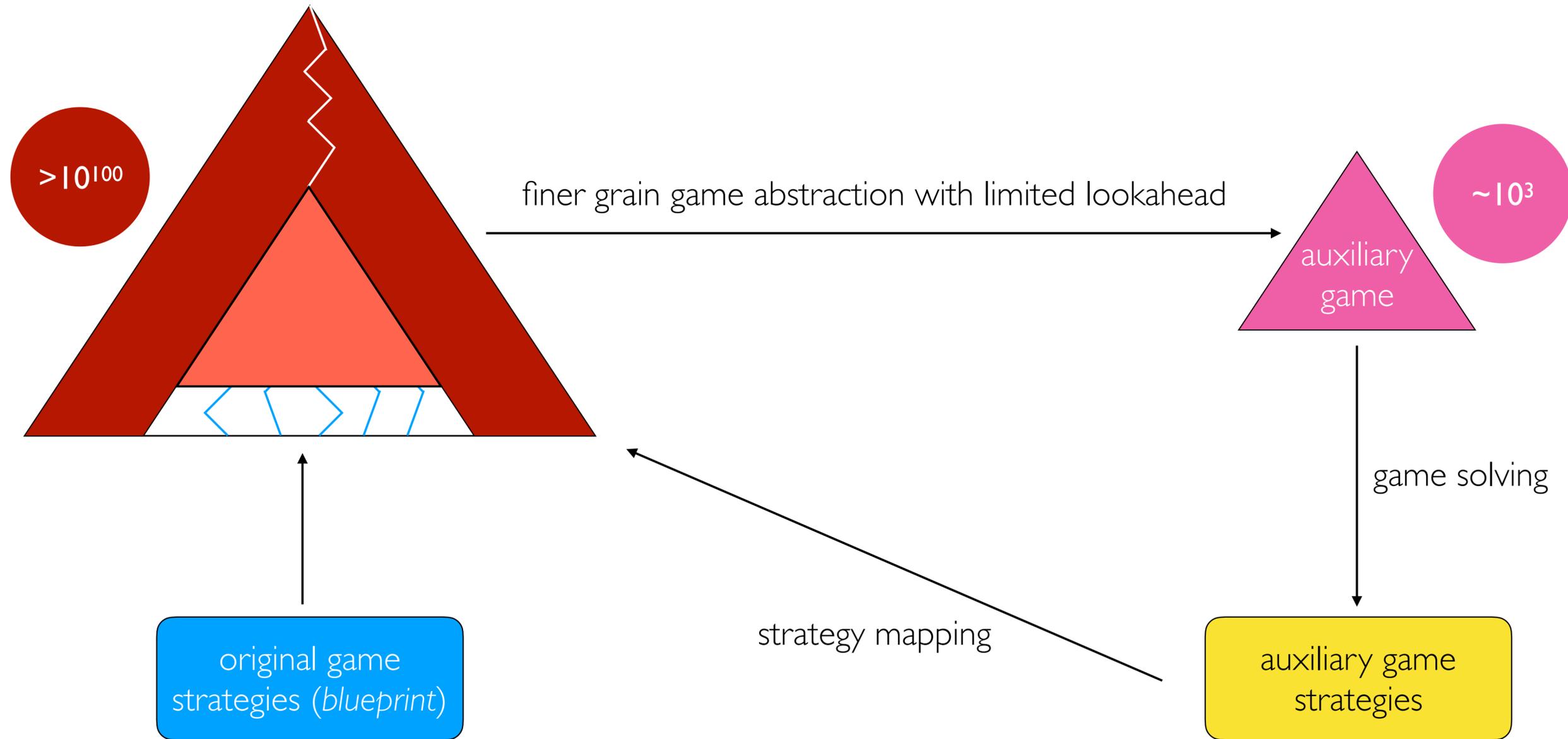
State-of-the-art huge-game solving: *play*



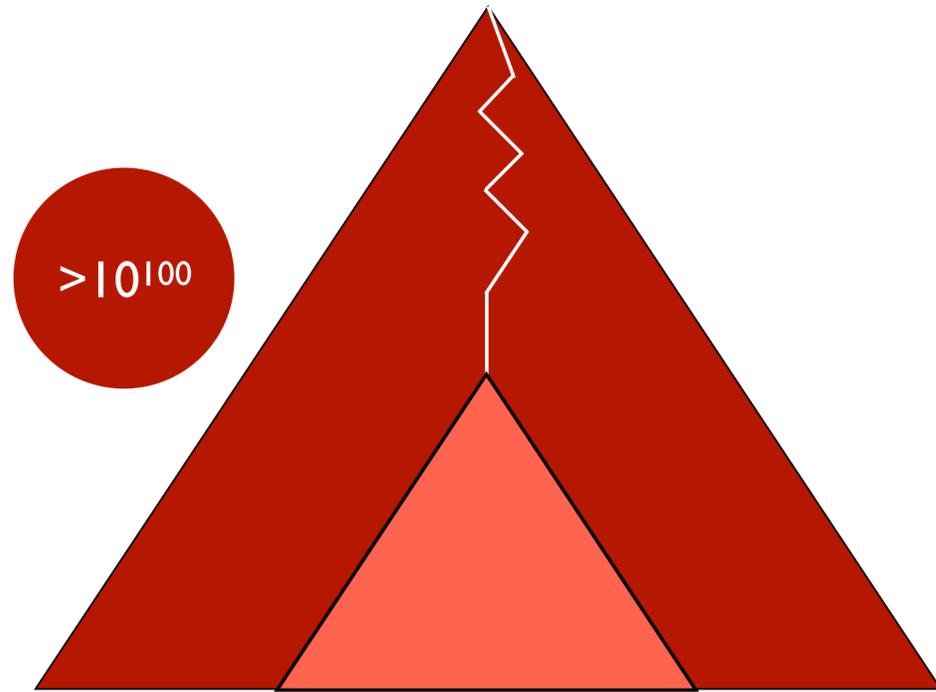
State-of-the-art huge-game solving: *play*



State-of-the-art huge-game solving: *play*

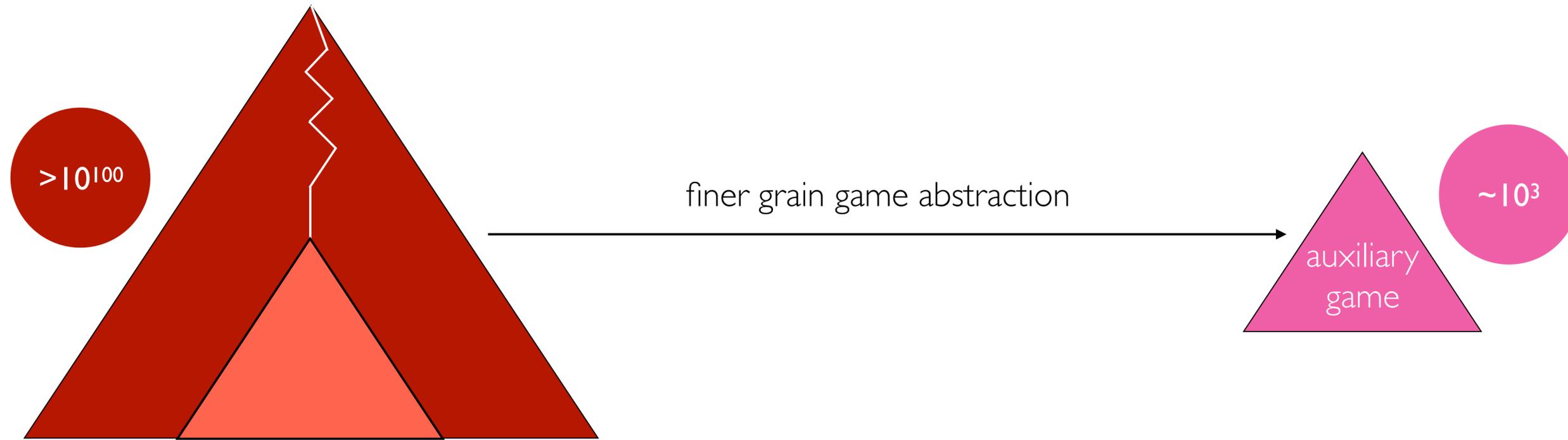


State-of-the-art huge-game solving: *play*



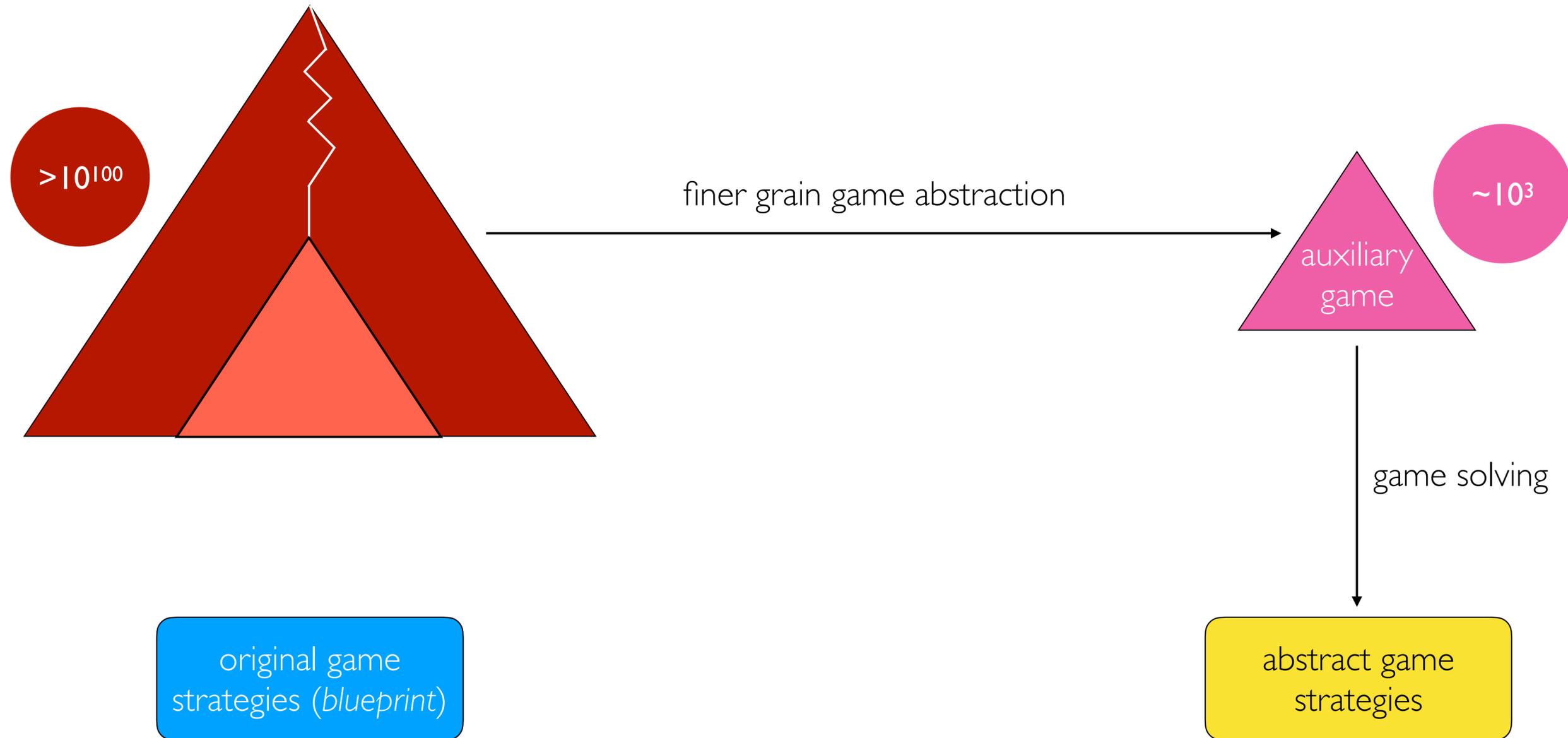
original game
strategies (*blueprint*)

State-of-the-art huge-game solving: *play*

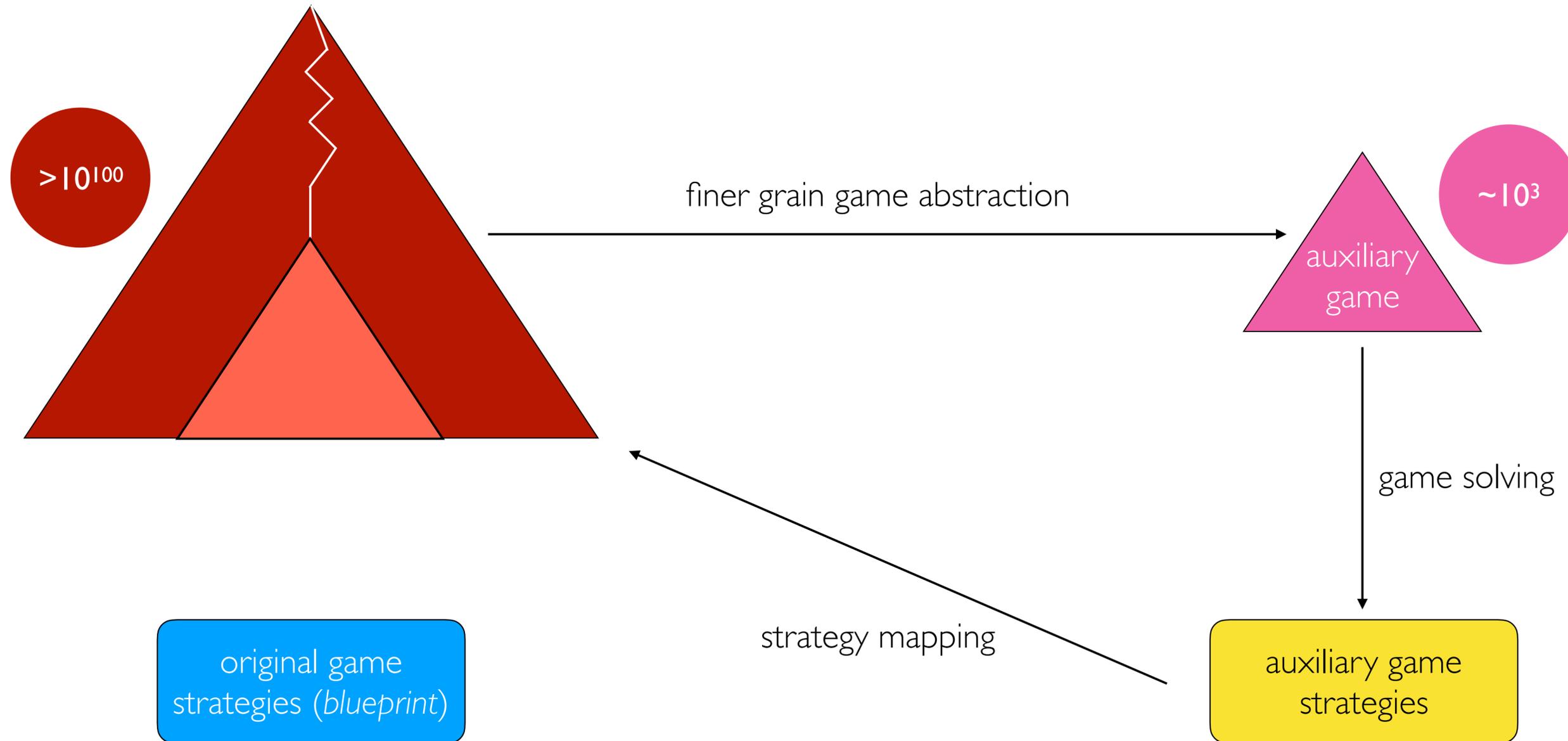


original game
strategies (*blueprint*)

State-of-the-art huge-game solving: *play*



State-of-the-art huge-game solving: *play*



Game solving

- Optimization problem definition
- Counter Factual Regret minimization plus (CFR+) (Zinkevich et al., *Counter Factual Regret minimization in games with incomplete information*, NeurIPS, 2008)
- Monte Carlo CFR+ (Lanctot et al., *Monte Carlo sampling for regret minimization in extensive gaems*, NeurIPS, 2009)

Abstractions

- Smaller version of the game capturing the most essential properties of the real domain
- Abstracted game solution provides a useful approximation of the optimal strategy

Abstractions

- Smaller version of the game capturing the most essential properties of the real domain
- Abstracted game solution provides a useful approximation of the optimal strategy
- Lossless information abstractions (Gilpin, Sandholm, 2007)
- Lossy information abstractions (Gilpin, Sandholm, 2007)

Abstractions

- Information abstractions
 - Linear programming and bucketing
 - Expectation-based and potential-aware abstractions (Gilpin, Sandholm, 2007)
- Action abstractions
 - Actions discretization and game refinement (Brown, Sandholm, 2015)
- Simulation-based abstractions (Tuyls et al. 2018), (Viqueira et al., 2019)

Beyond abstractions



Libratus, 2017

**Carnegie
Mellon
University**

**Game Abstraction + MCCFR self-play
Nested subgame solving
Self-improvement**

Beyond abstractions



Libratus, 2017

Game Abstraction + MCCFR self-play
Nested subgame solving
Self-improvement



Pluribus, 2019

facebook Artificial Intelligence

Extended to 6 players



Simulation-based abstractions

- Bottom-up approach based on *data* (artificial learning)
- Game *traces* (observed vs generated)
- Query an *oracle* for noisy payoff given a strategy
- *Model-free*

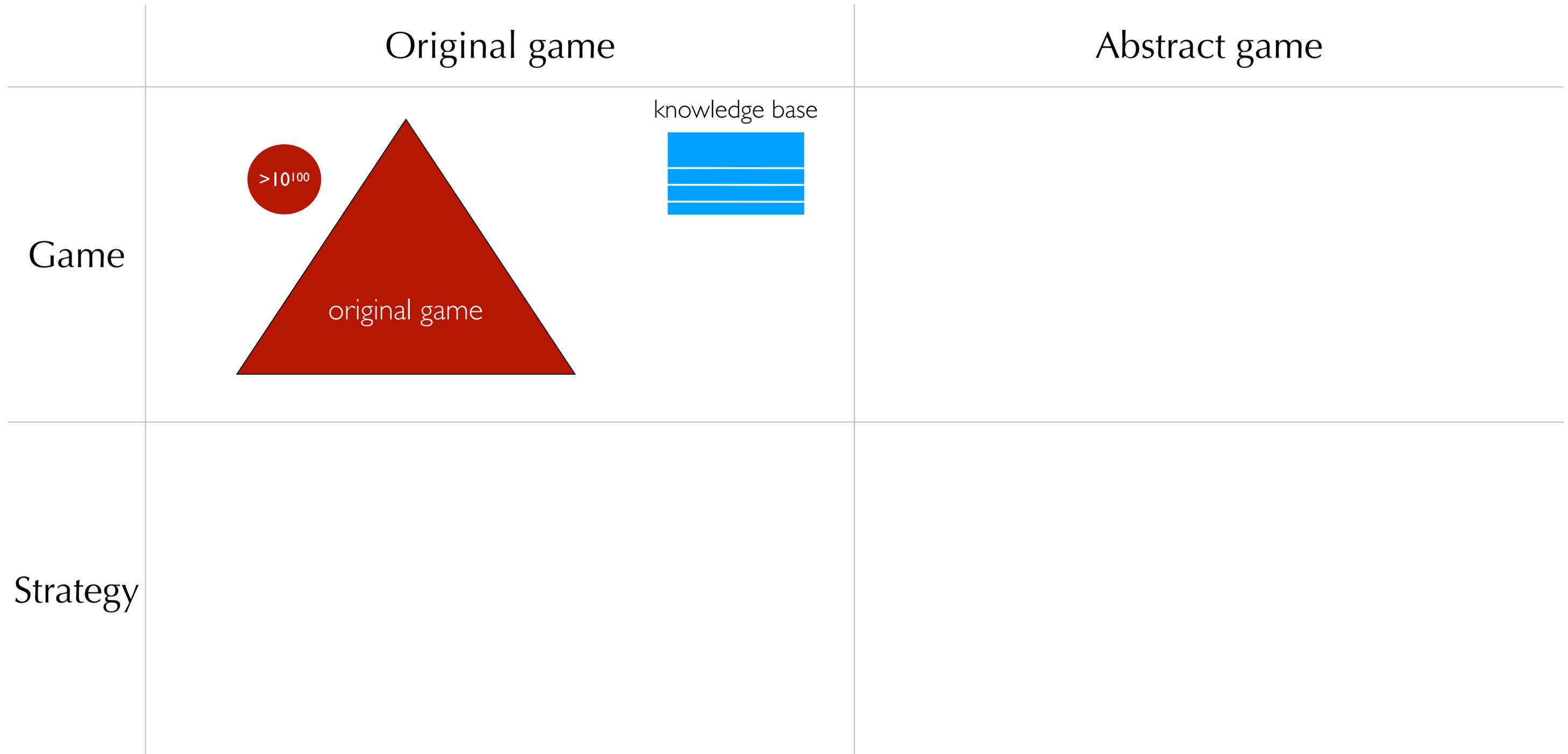
The problem

- Real-world games and strategic scenarios are *too large* to be represented
- No clear *domain-independent* abstraction approach was presented to solve these games
- Poker as the main reference application

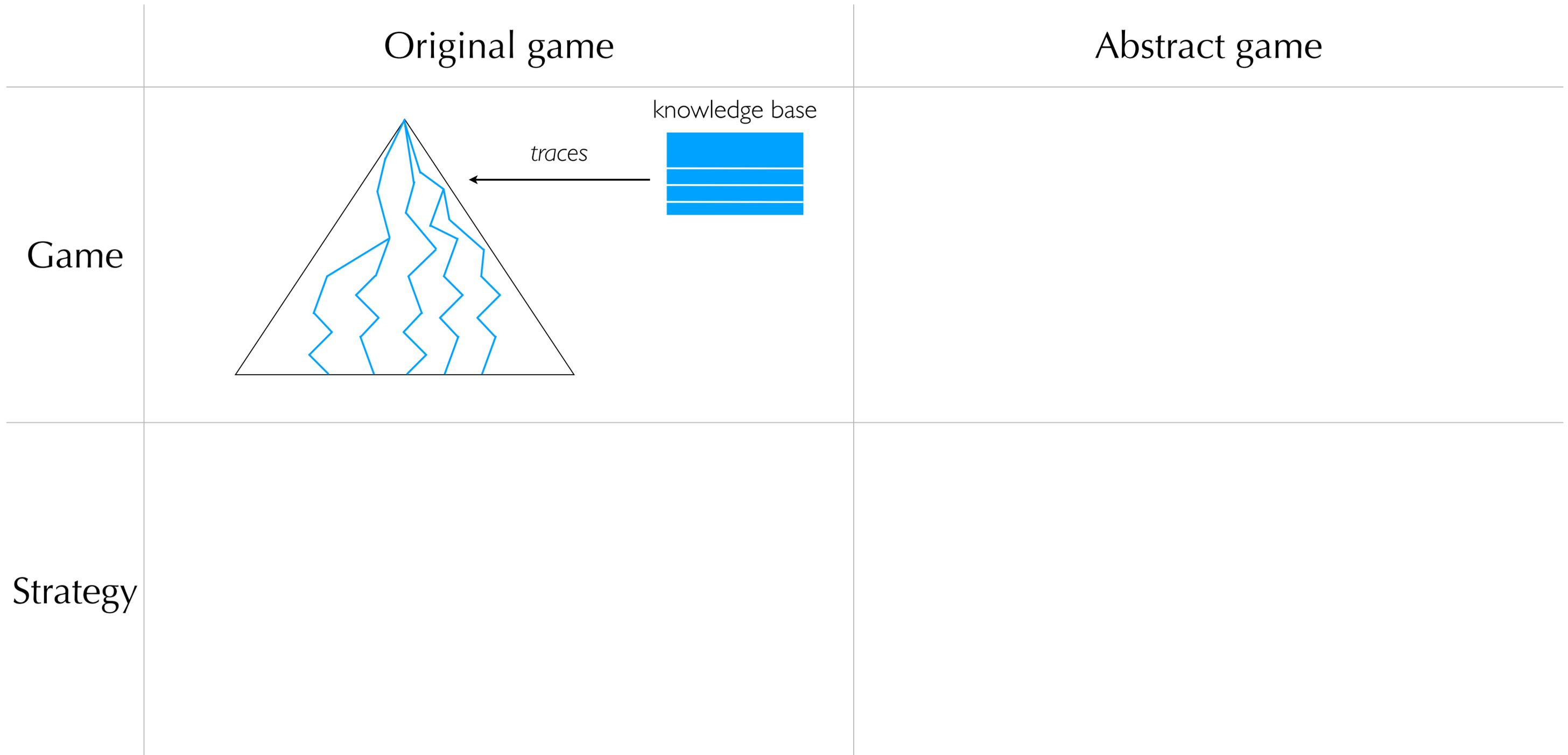
Our goal

Develop a *bottom-up model-free* abstraction approach, supported by *theoretical guarantees*, able to find *mixed strategy* Nash equilibria in *any* extensive-form game in a *simulation-based* fashion.

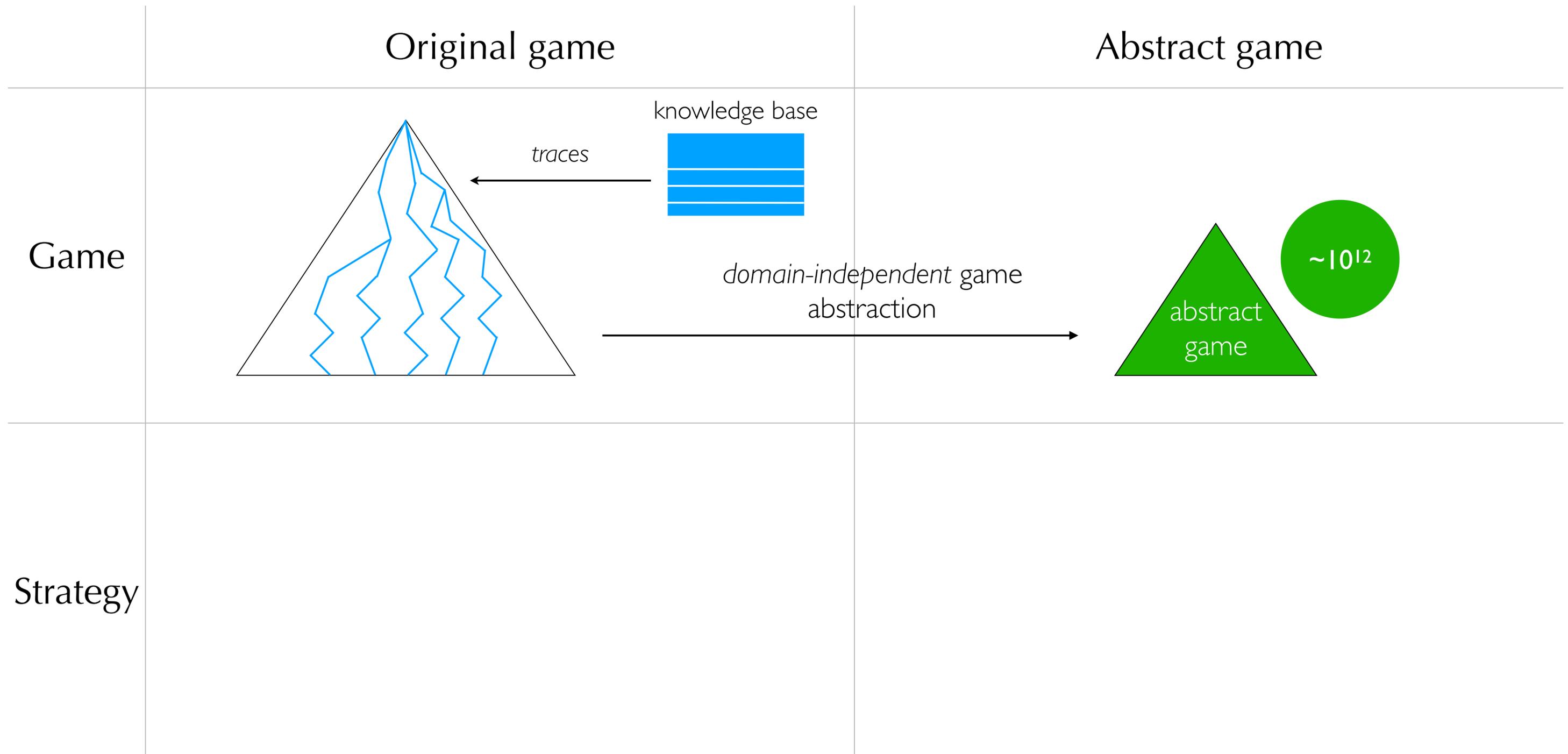
Idea



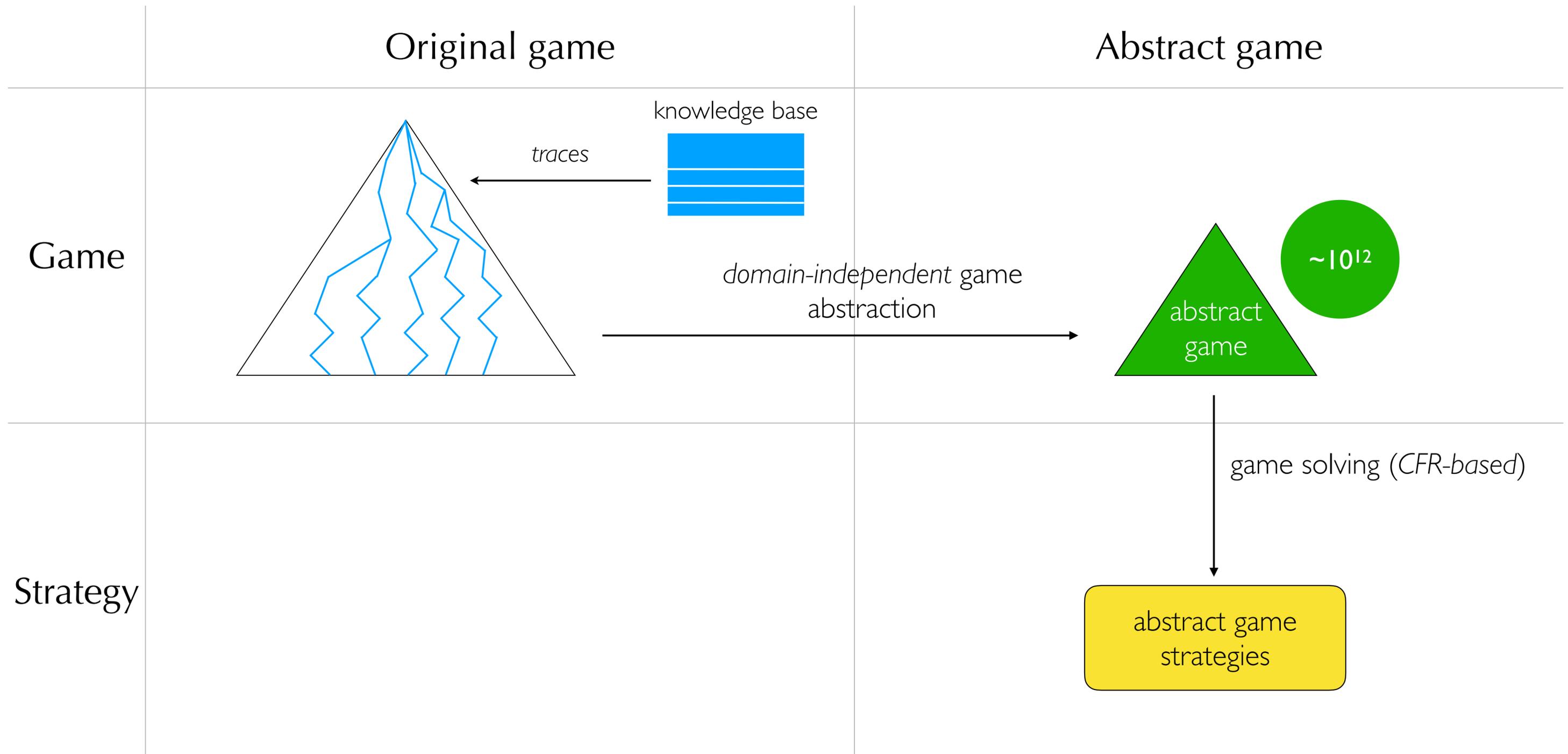
Idea



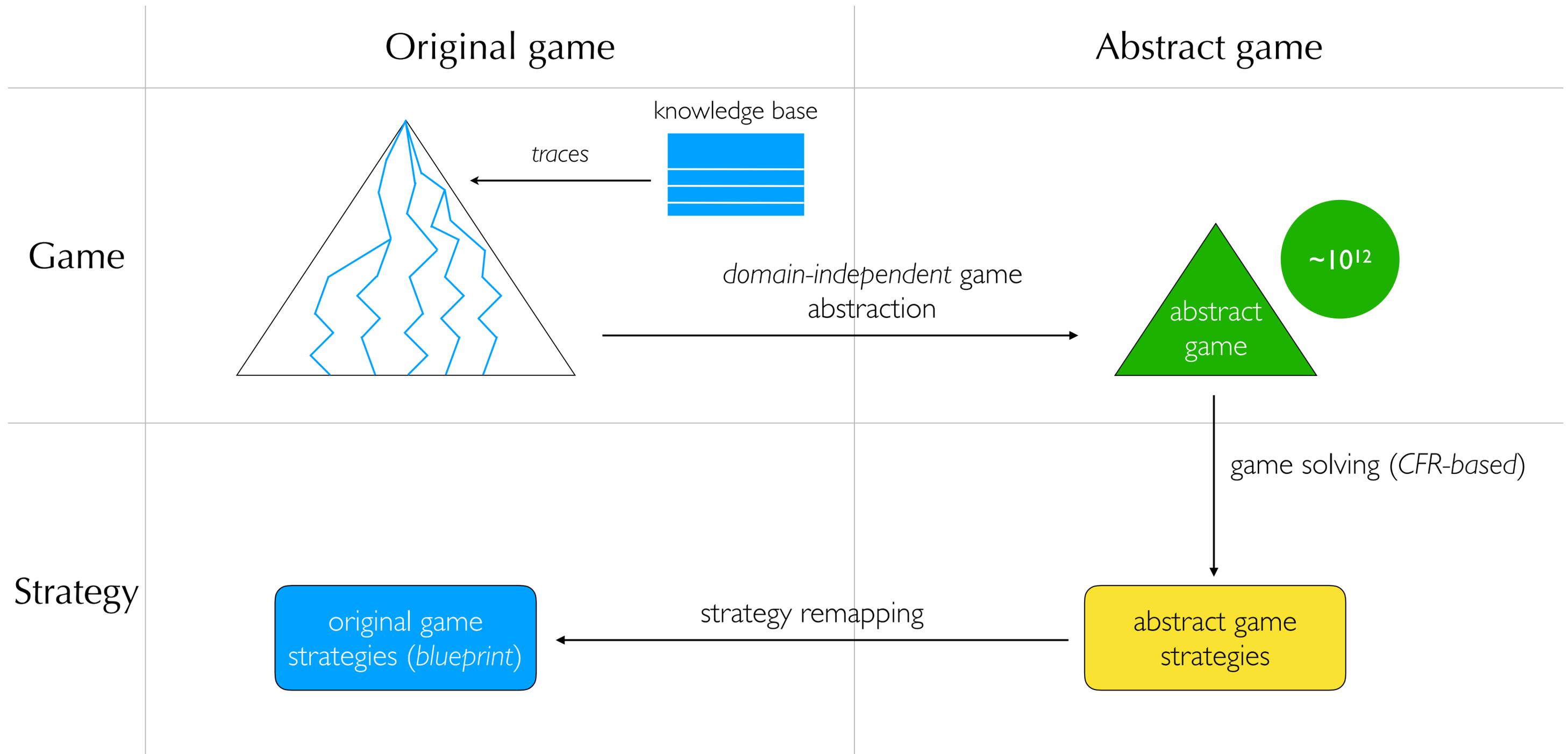
Idea



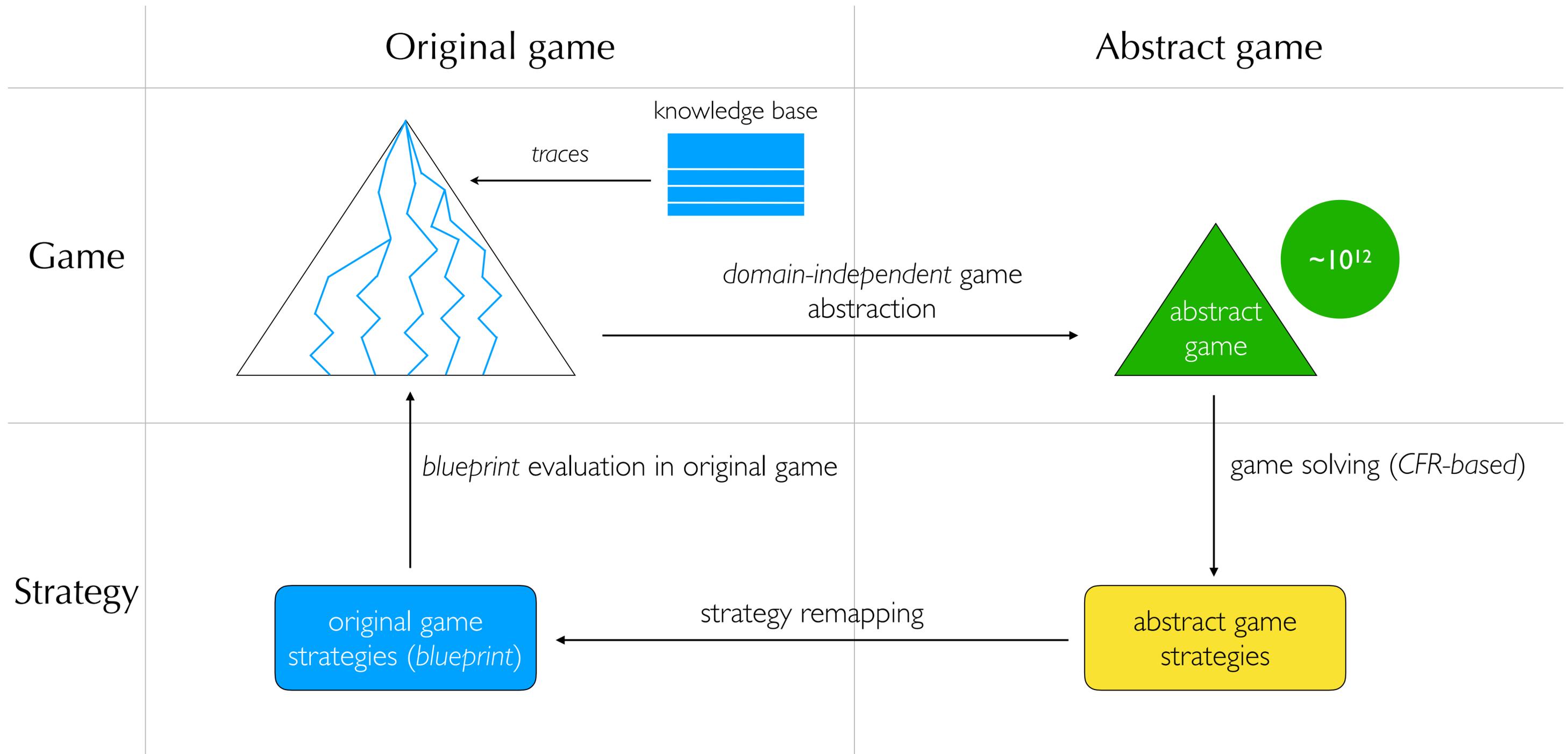
Idea



Idea



Idea



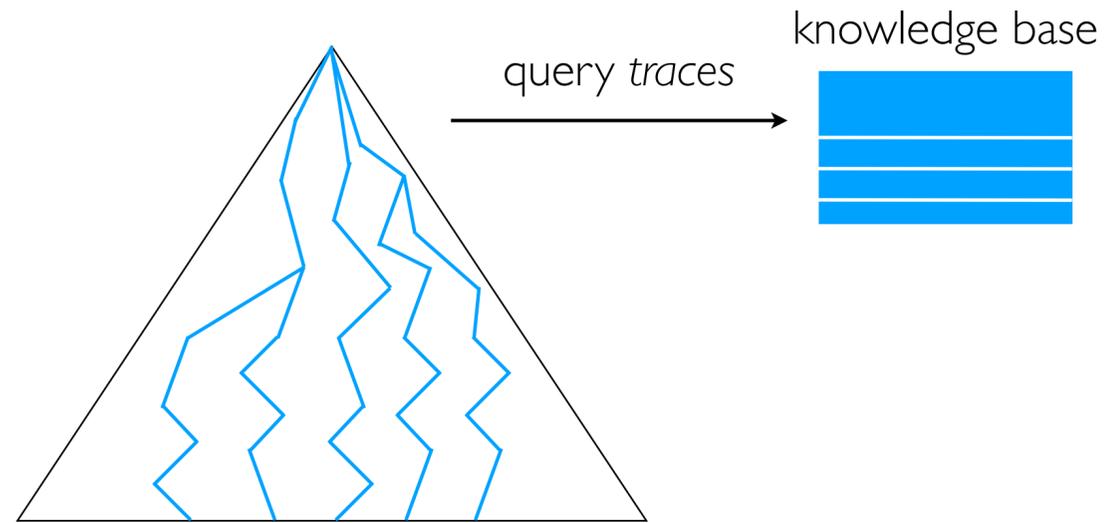
Idea

Blueprint exploitability evaluation
Traces gathering to refine abstraction

Original game

Abstract game

Game



blueprint evaluation in original game

Strategy

original game
strategies (*blueprint*)

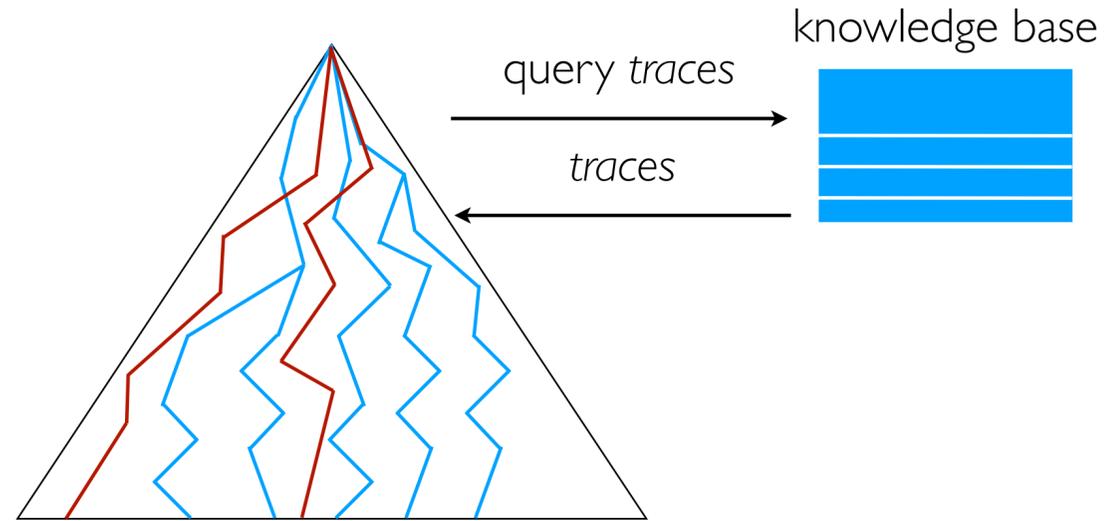
Idea

Blueprint exploitability evaluation
Traces gathering to refine abstraction

Original game

Abstract game

Game



blueprint evaluation in original game

Strategy

original game
strategies (*blueprint*)

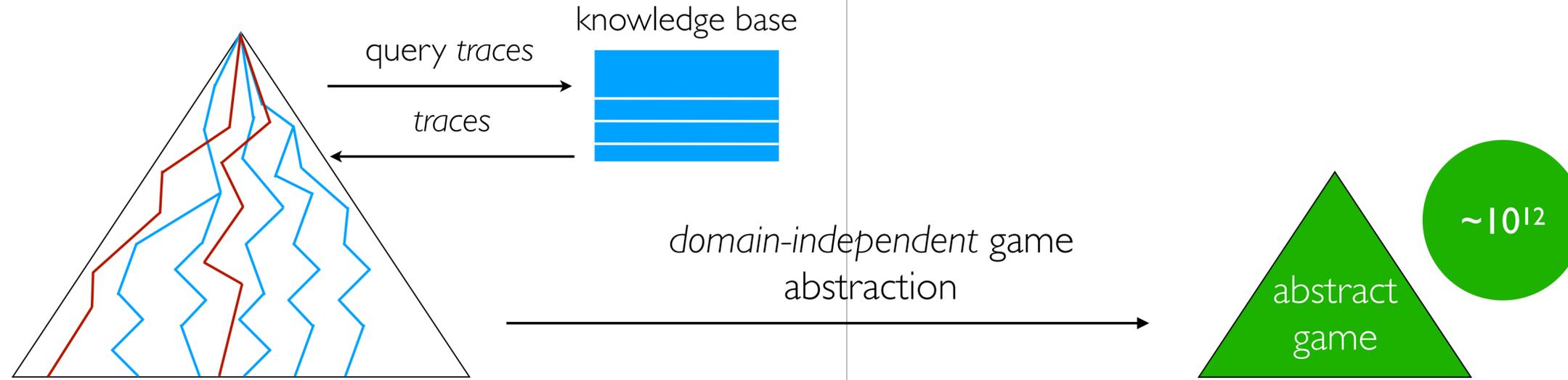
Idea

Blueprint exploitability evaluation
Traces gathering to refine abstraction

Original game

Abstract game

Game



Strategy

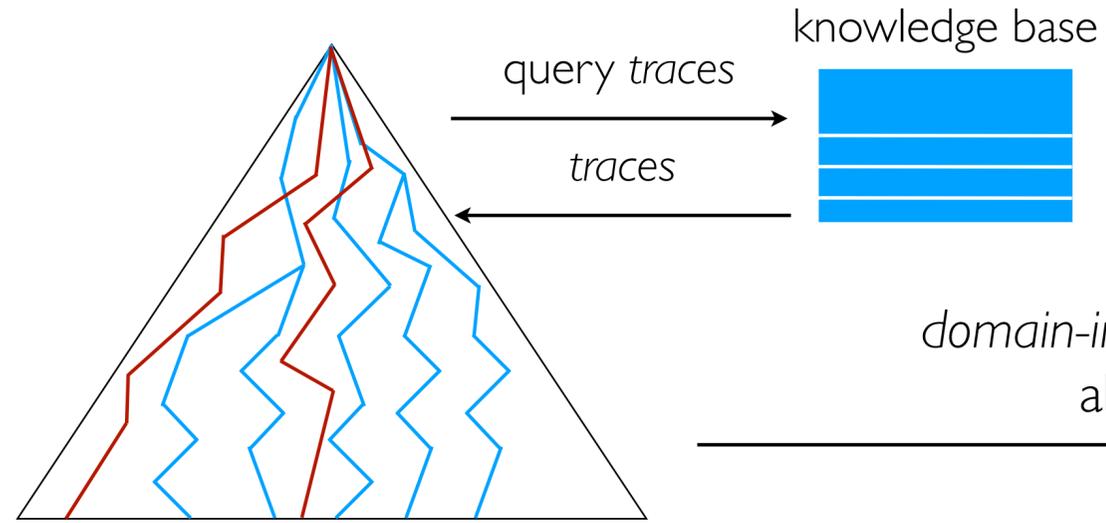
Idea

Blueprint exploitability evaluation
Traces gathering to refine abstraction

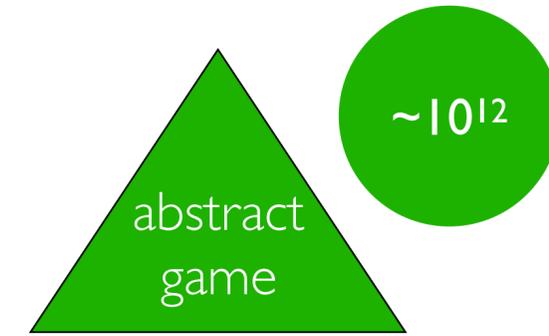
Original game

Abstract game

Game



domain-independent game
abstraction



Strategy

game solving (CFR-based)

abstract game
strategies

Idea

Blueprint exploitability evaluation
Traces gathering to refine abstraction

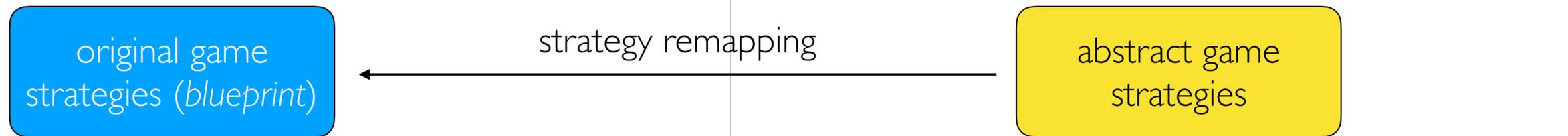
Original game

Abstract game

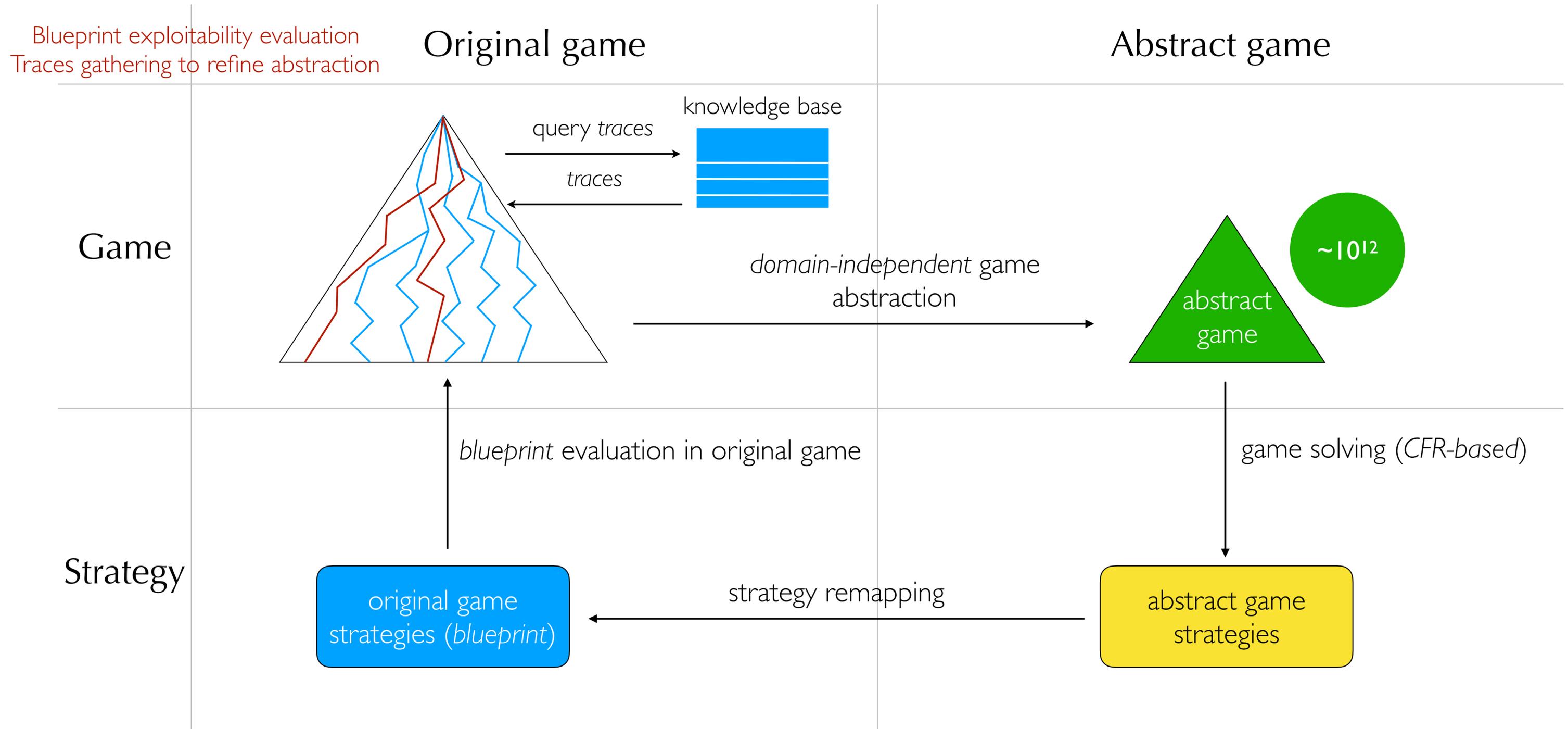
Game



Strategy



Idea



In a nutshell...

- Real-world strategic conditions are *too large* to be represented and analyzed
- Need for a *domain-independent* way to solve large games
- Exploit *data availability (traces)* and *artificial learning techniques*

In a nutshell...

- Real-world strategic conditions are *too large* to be represented and analyzed
- Need for a *domain-independent* way to solve large games
- Exploit *data availability (traces)* and *artificial learning techniques*

- How to abstract the game starting from *traces*?
- How to choose future *traces*?

Applications



Contract Bridge

Applications



Contract Bridge



Car Racing

Applications



Contract Bridge



Car Racing

Cybersecurity

