

Research Project Proposal: Abstractions in Extensive-Form Games

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CSE Track



POLITECNICO
MILANO 1863



HONOURS
PROGRAMME
HP-SR
in Information Technology

Recreational games



Chess

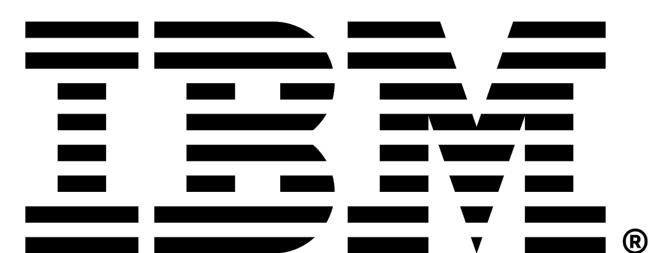
Recreational games



Chess

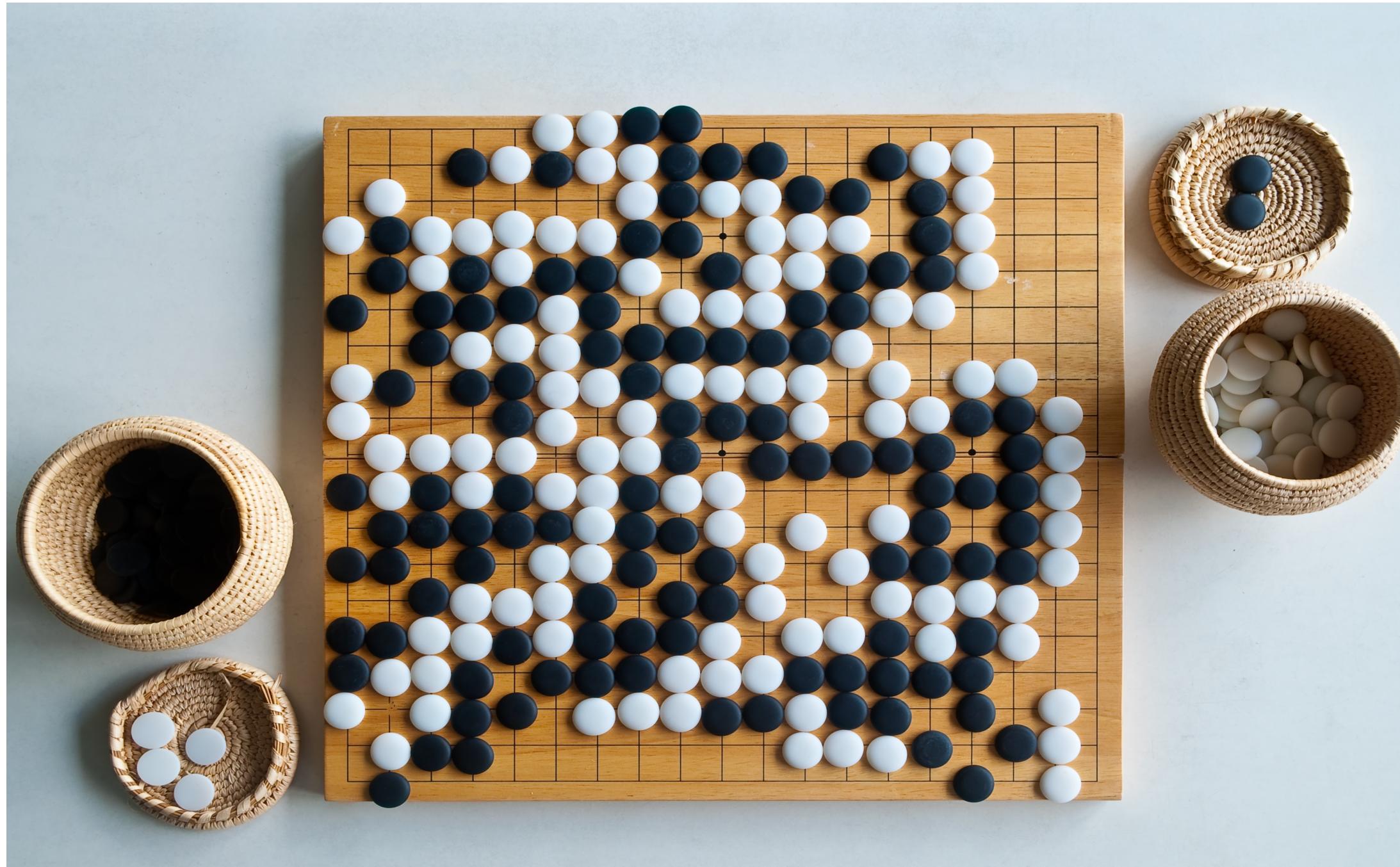


1996



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



AMG
PETRONAS
MOTORSPORT



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

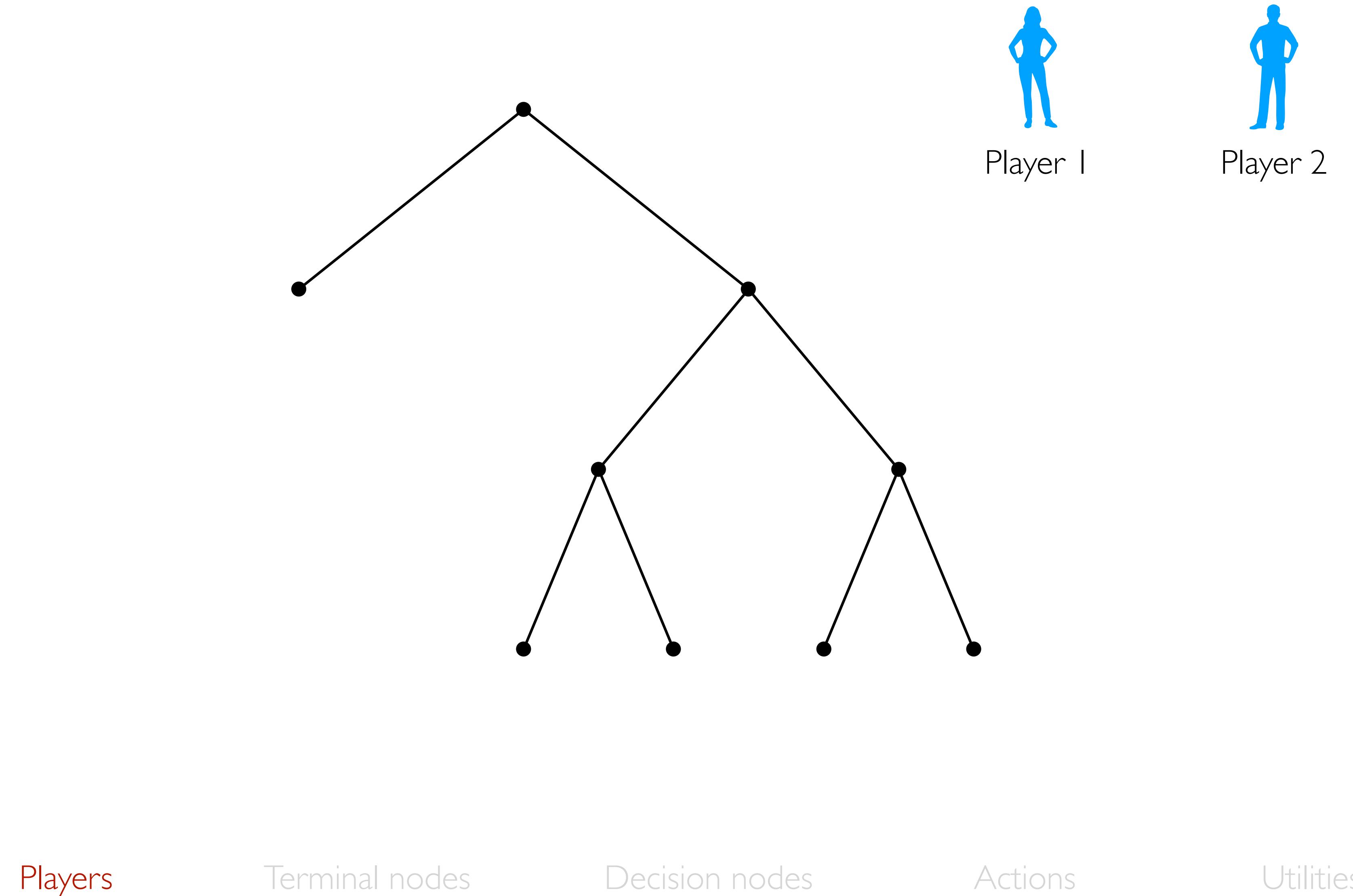
Terminal nodes

Decision nodes

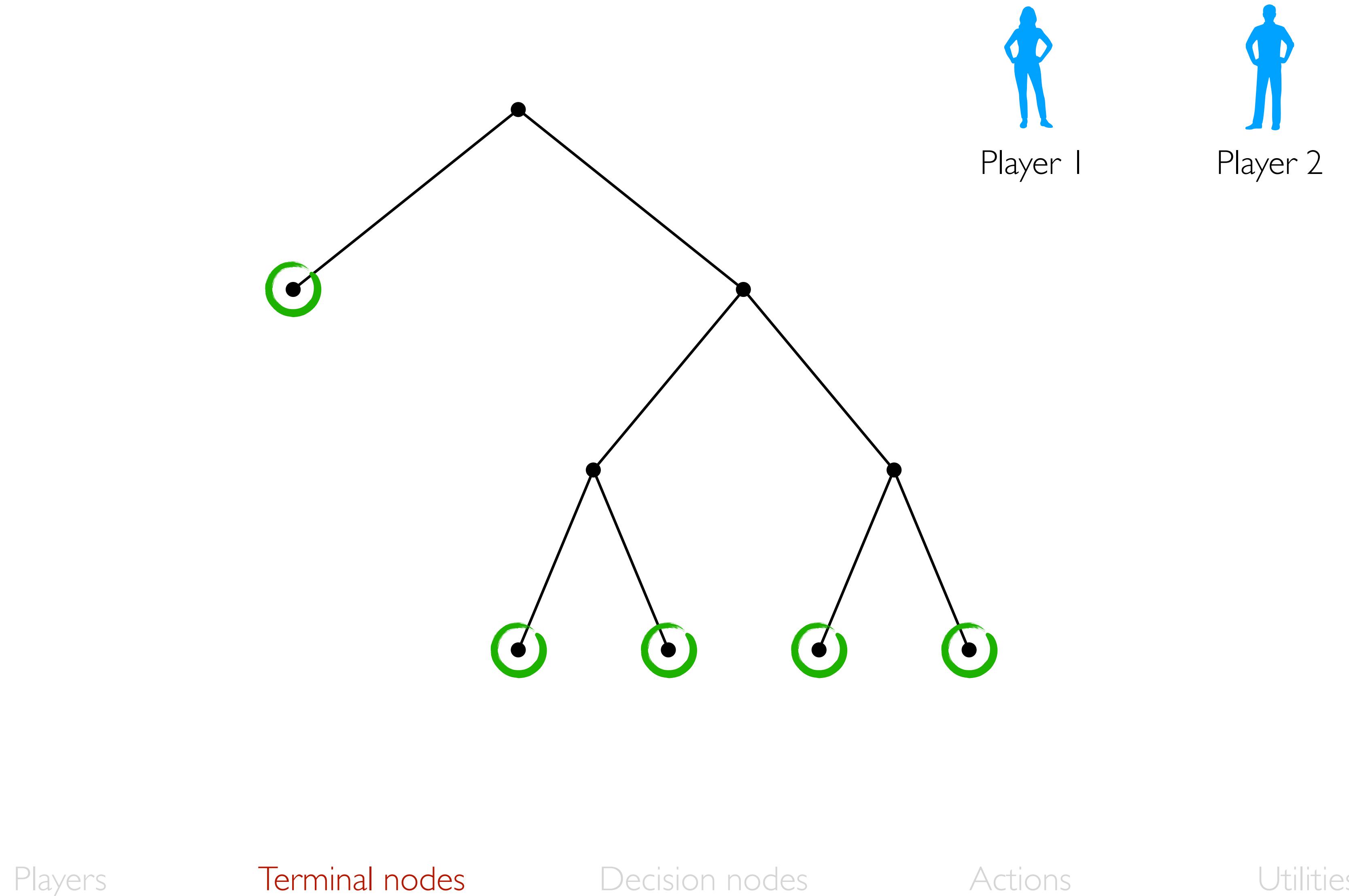
Actions

Utilities

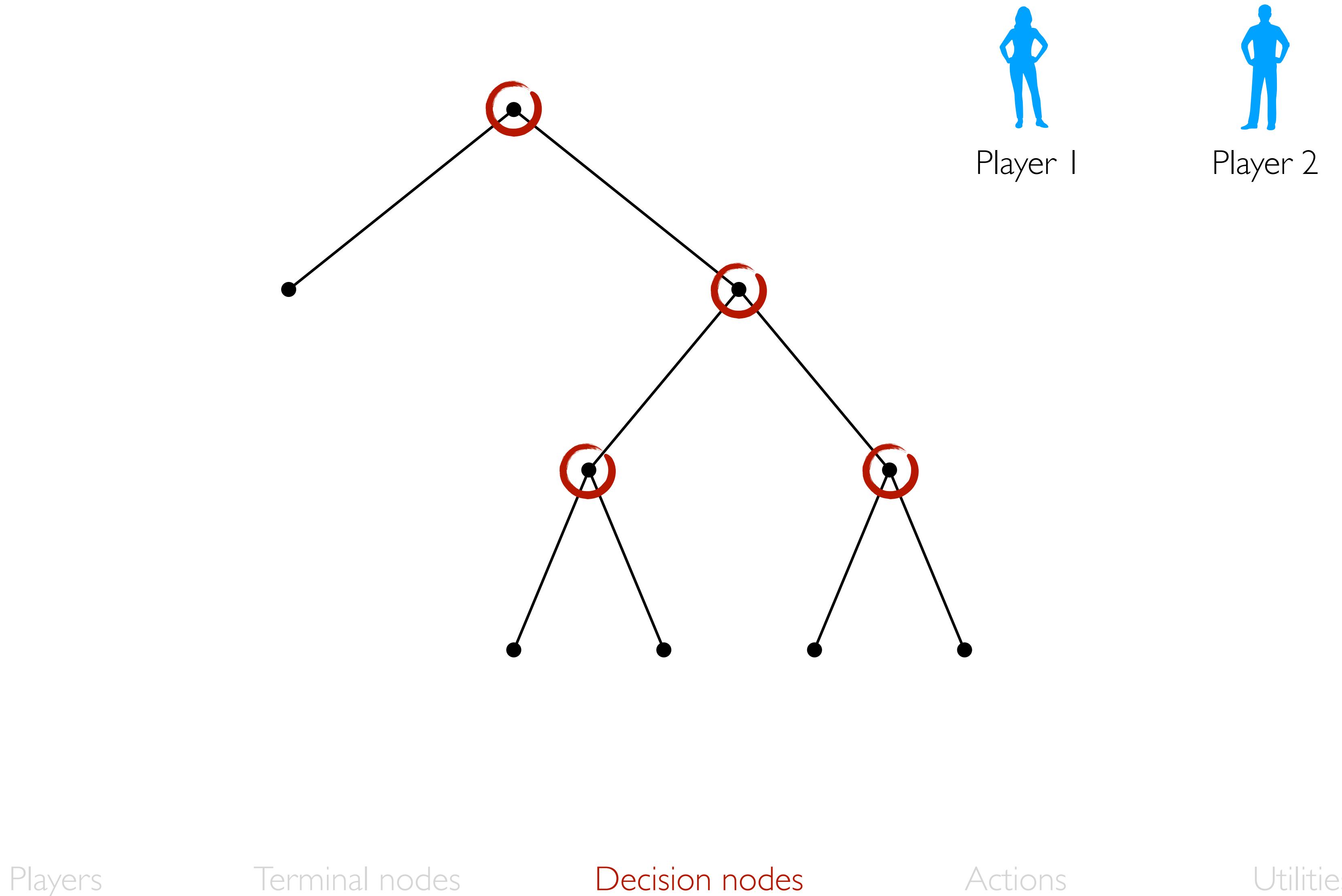
Game tree representation



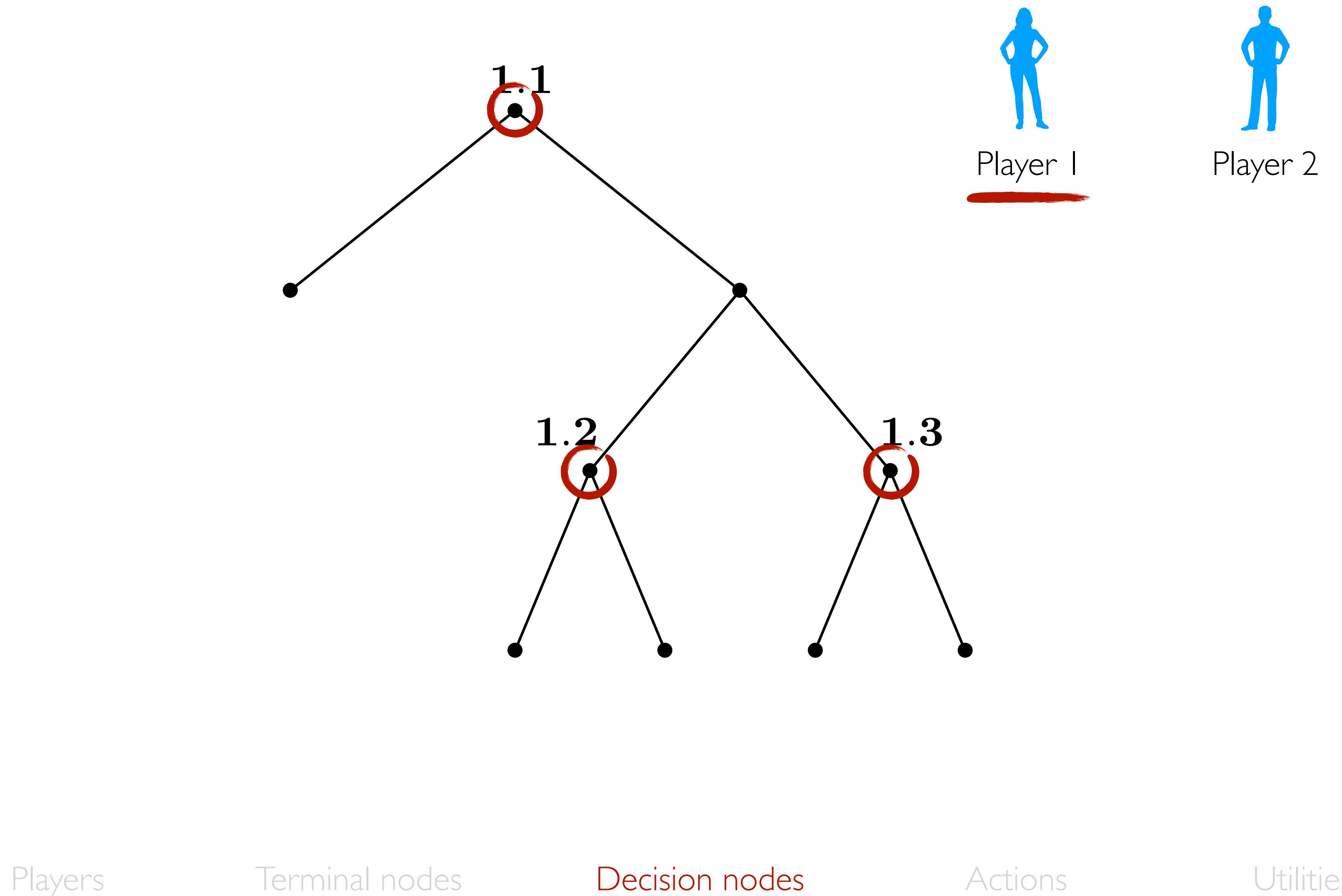
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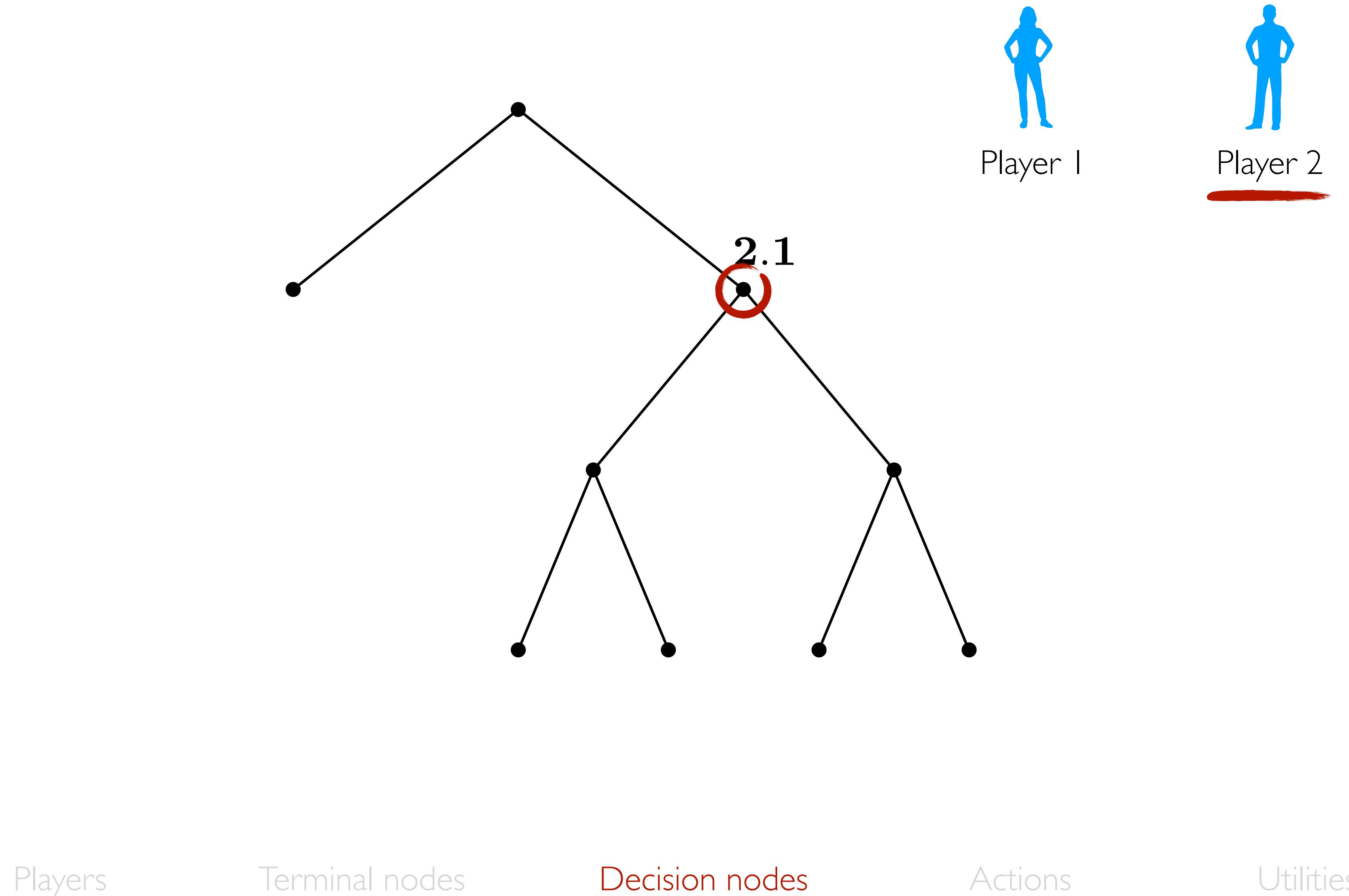
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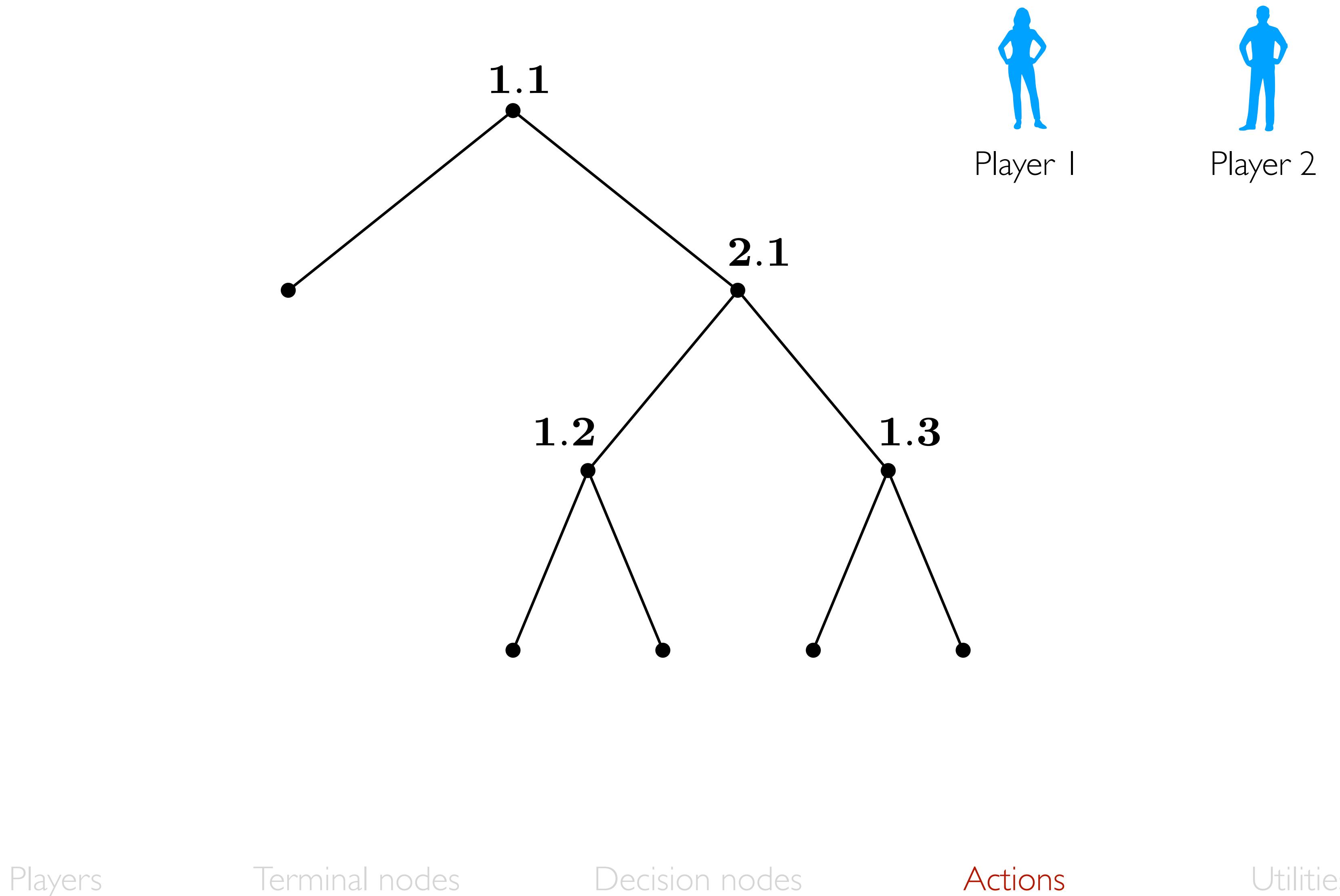
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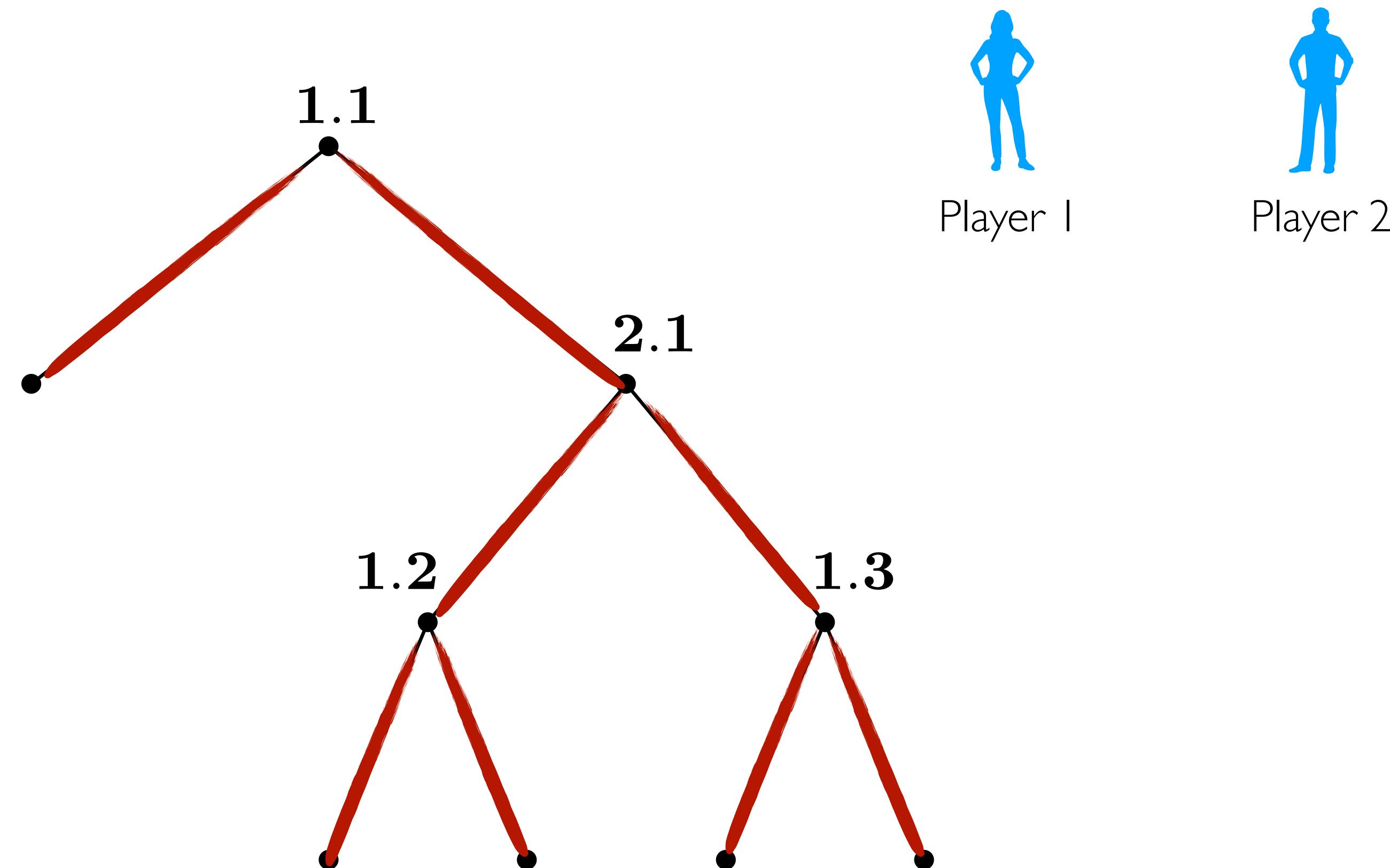
Game tree representation



Game tree representation



Game tree representation



Players

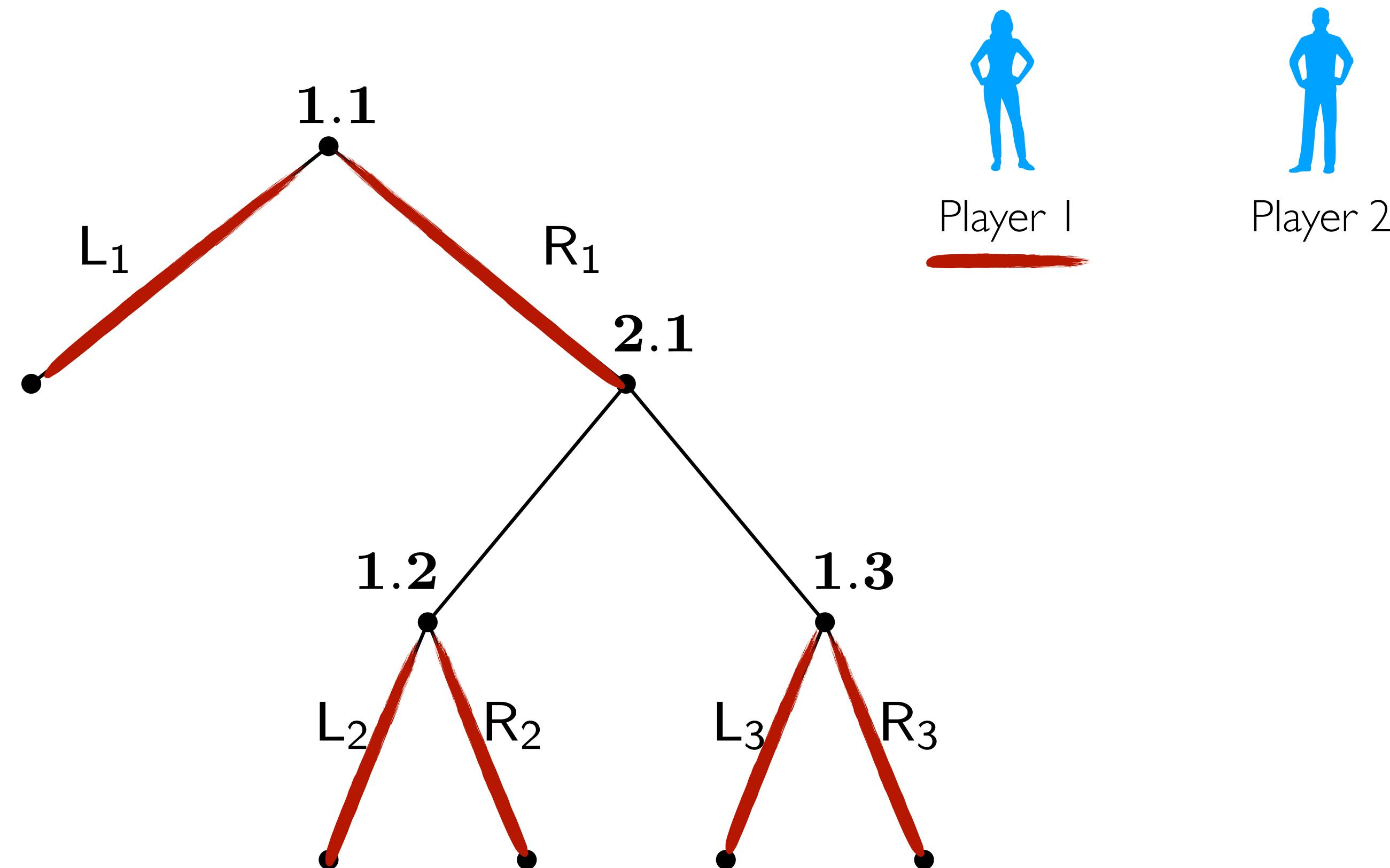
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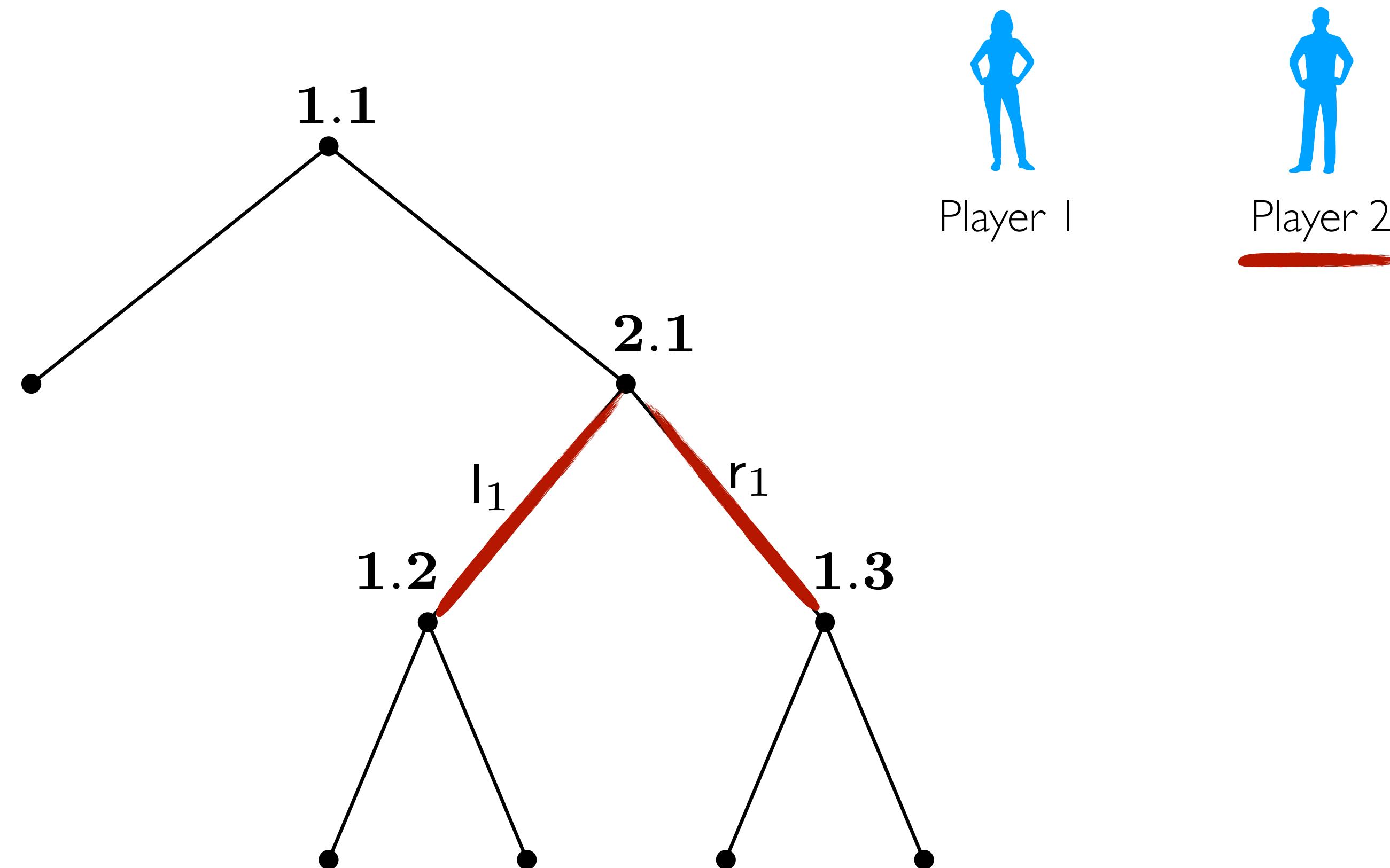
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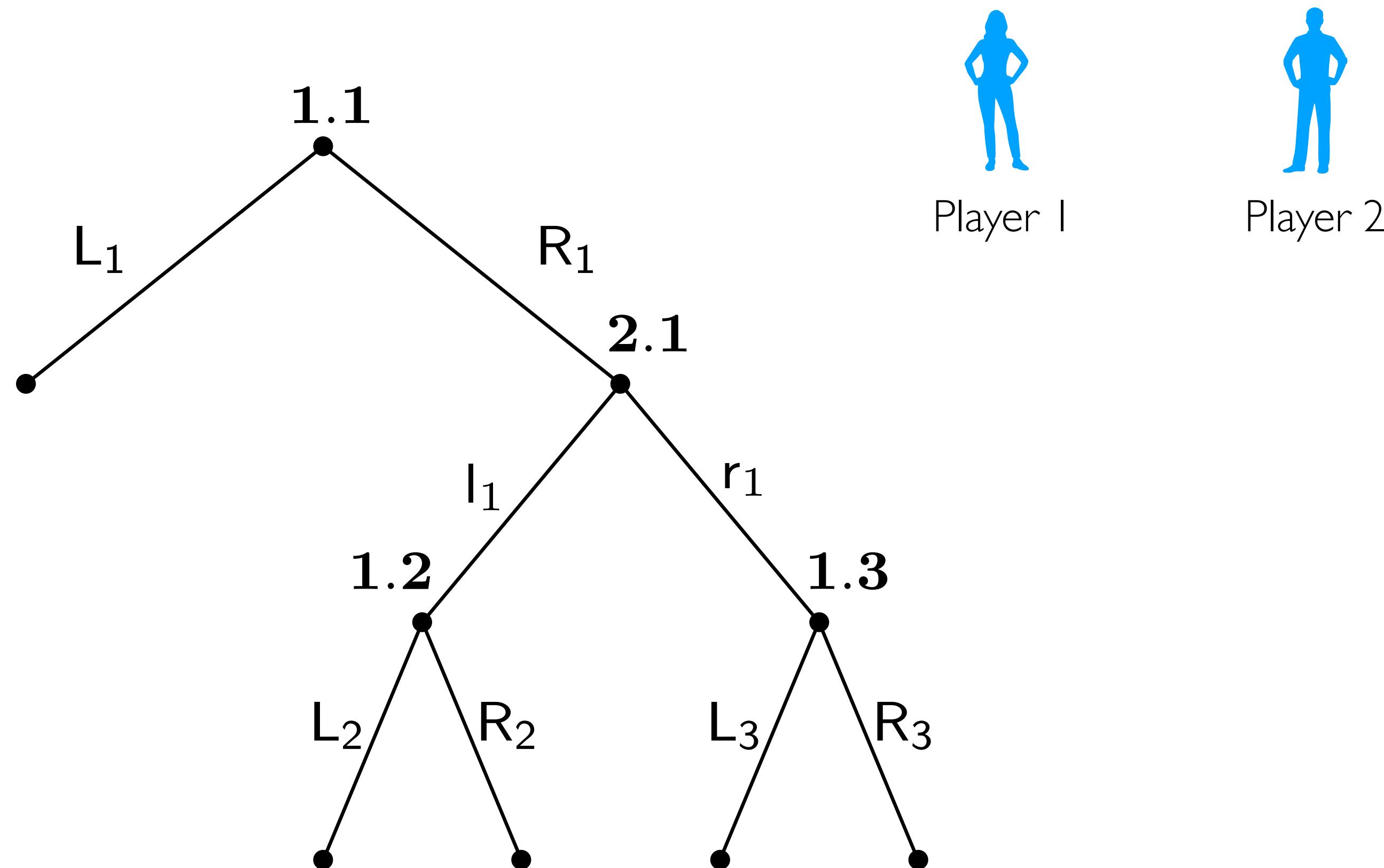
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Game tree representation



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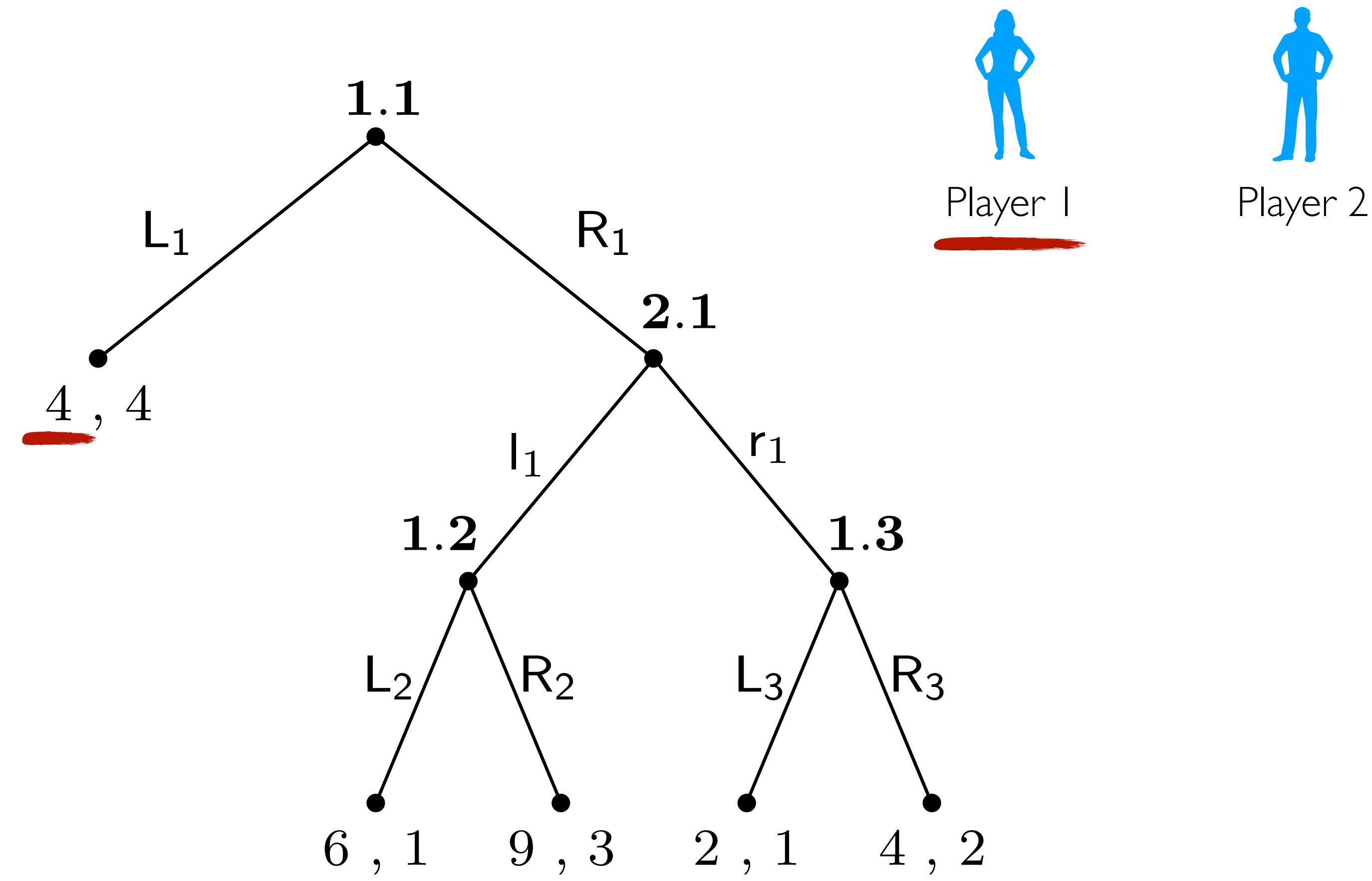
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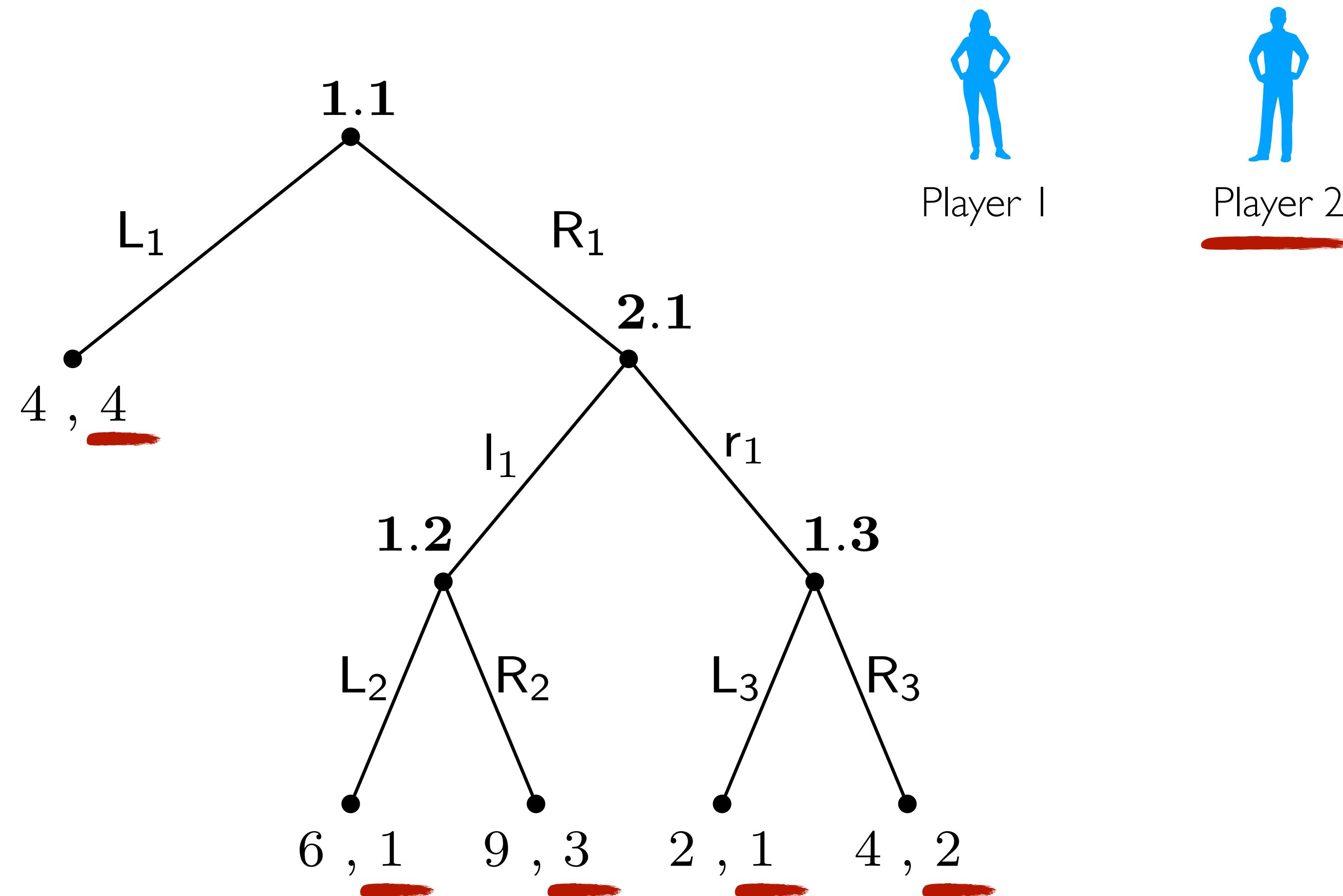
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Game tree representation



Players

Terminal nodes

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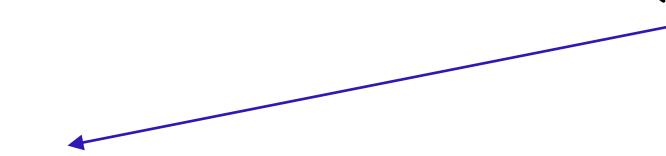
Formal model

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

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



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Set of players $\{1, 2\}$
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Set of actions $\{L_1, R_1, l_1, r_1, L_2, R_2, \dots\}$



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

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Player function: $V \longrightarrow N$

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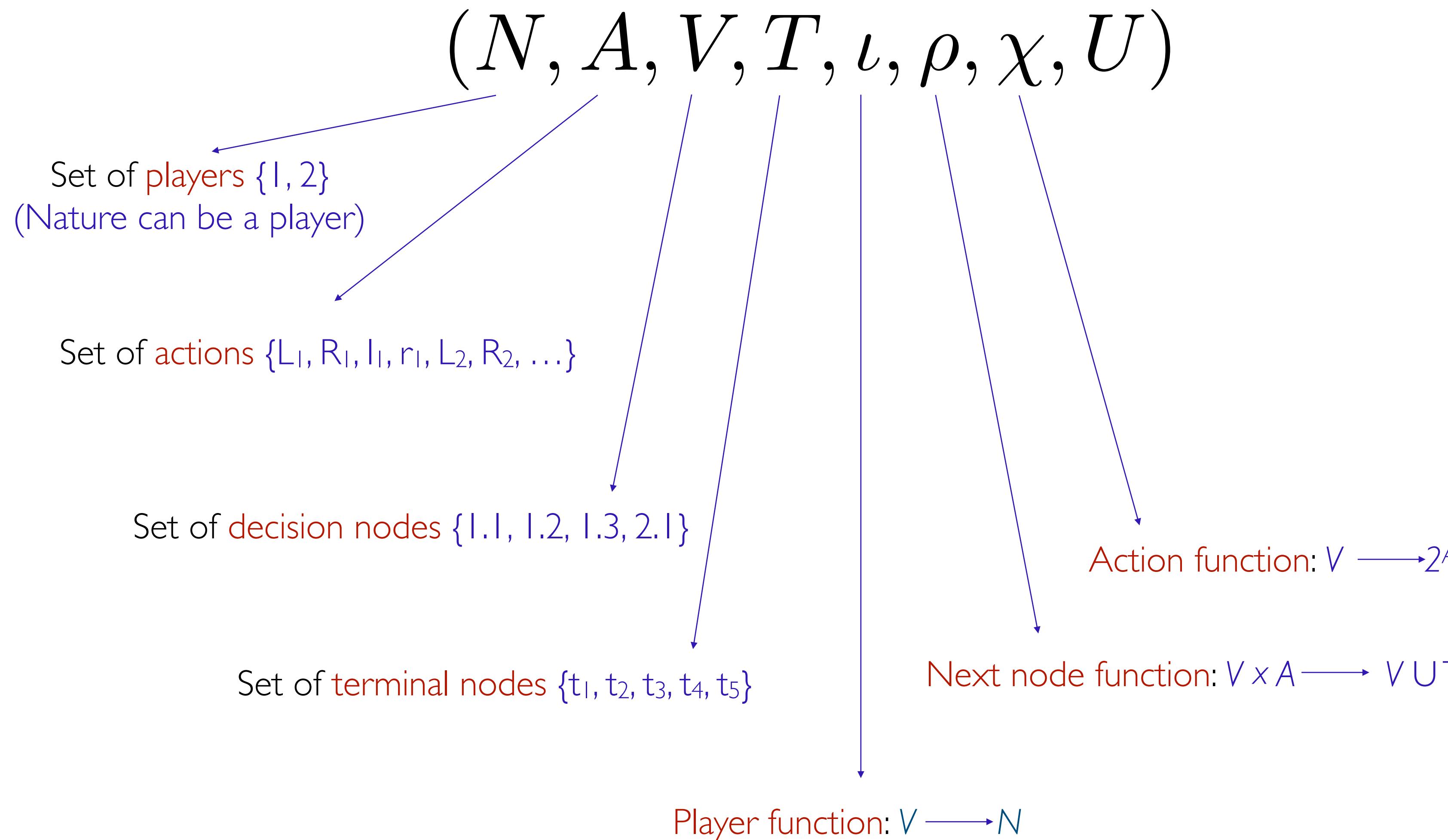
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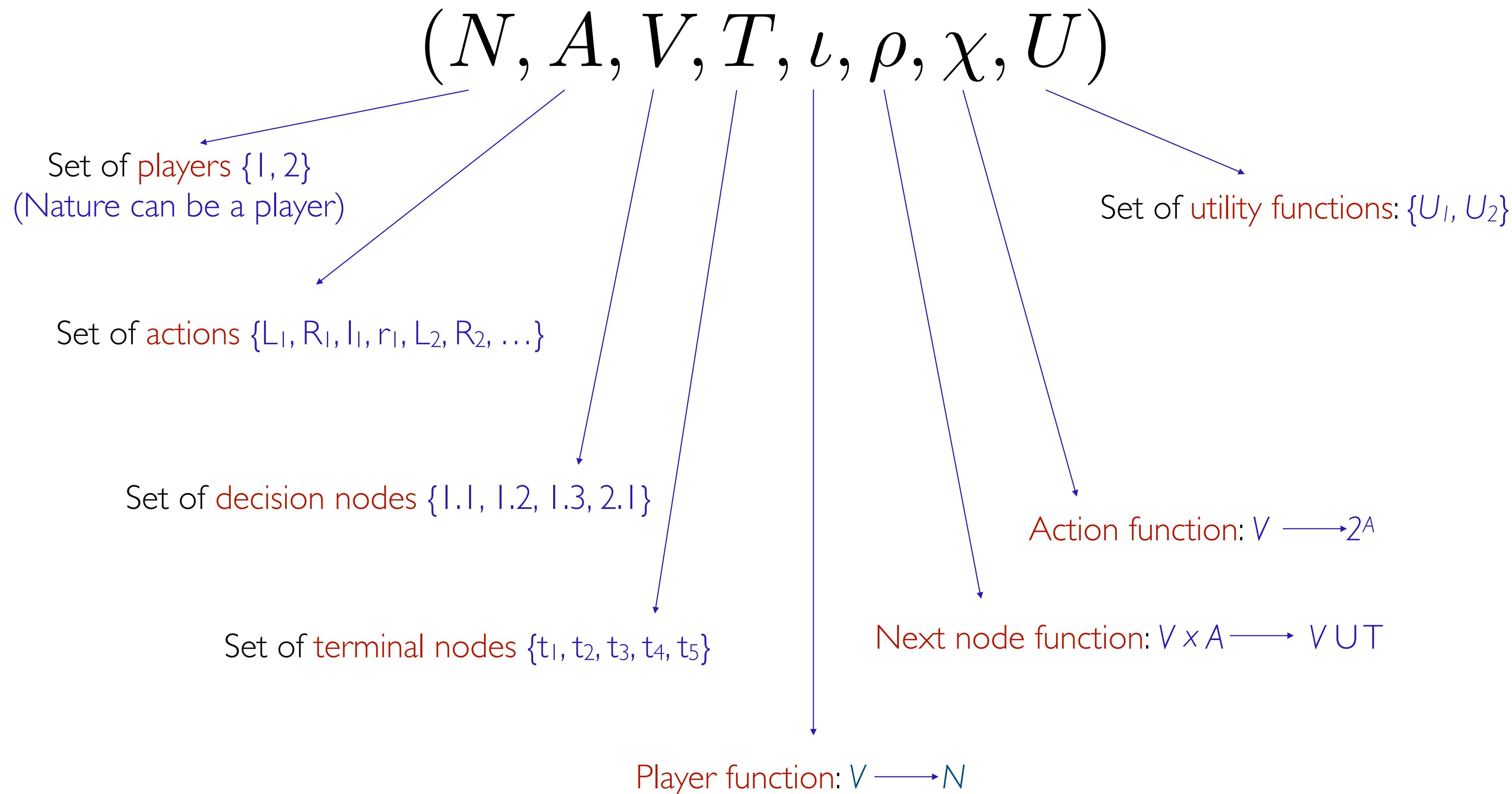
Next node function: $V \times A \longrightarrow V \cup T$

Player function: $V \longrightarrow N$

Formal model



Formal model



Perfect vs imperfect information



Perfect information

Perfect vs imperfect information



Perfect information



Imperfect information

Perfect vs imperfect information



Perfect information



Imperfect information

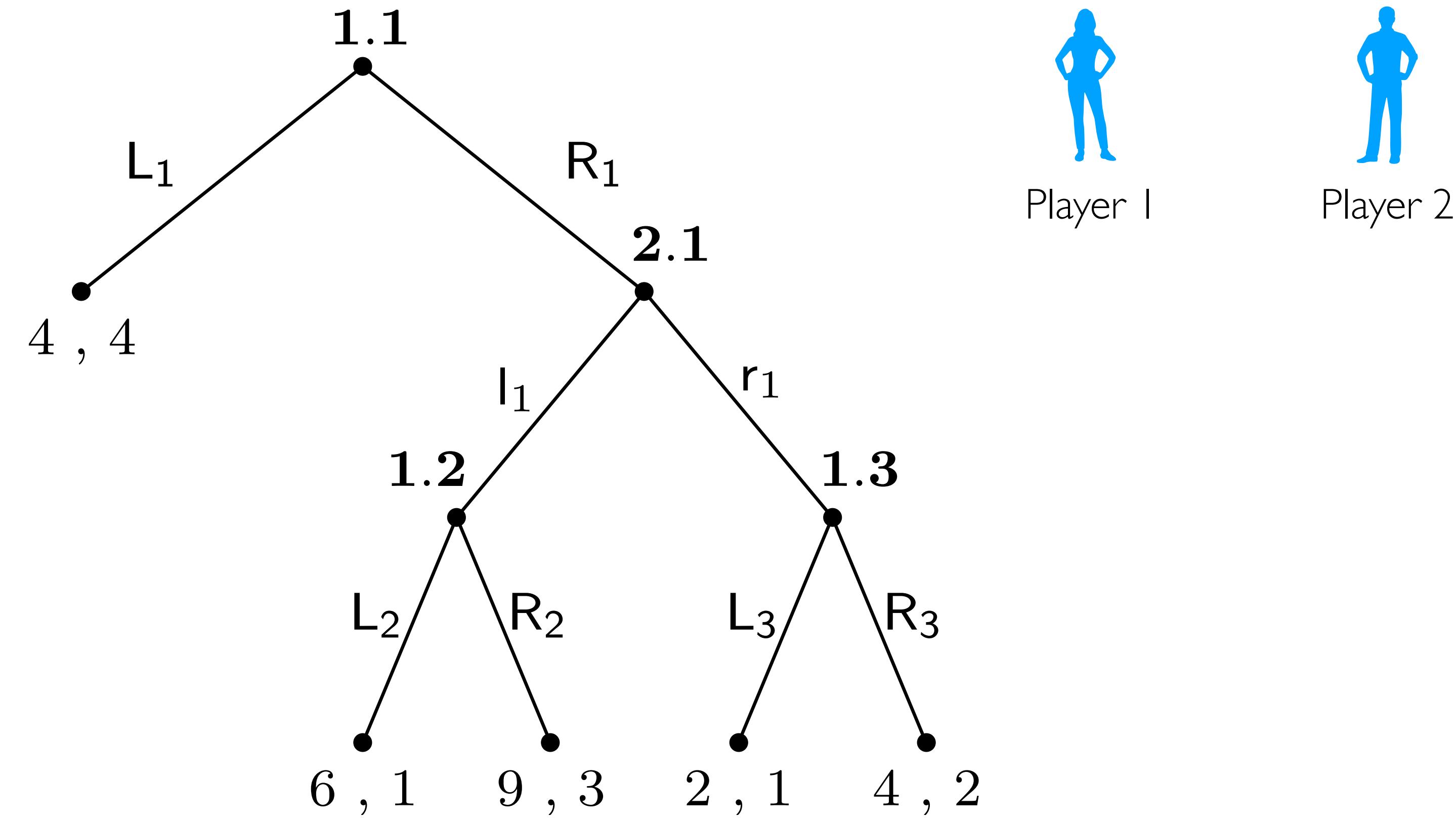
Information sets

Strategies

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

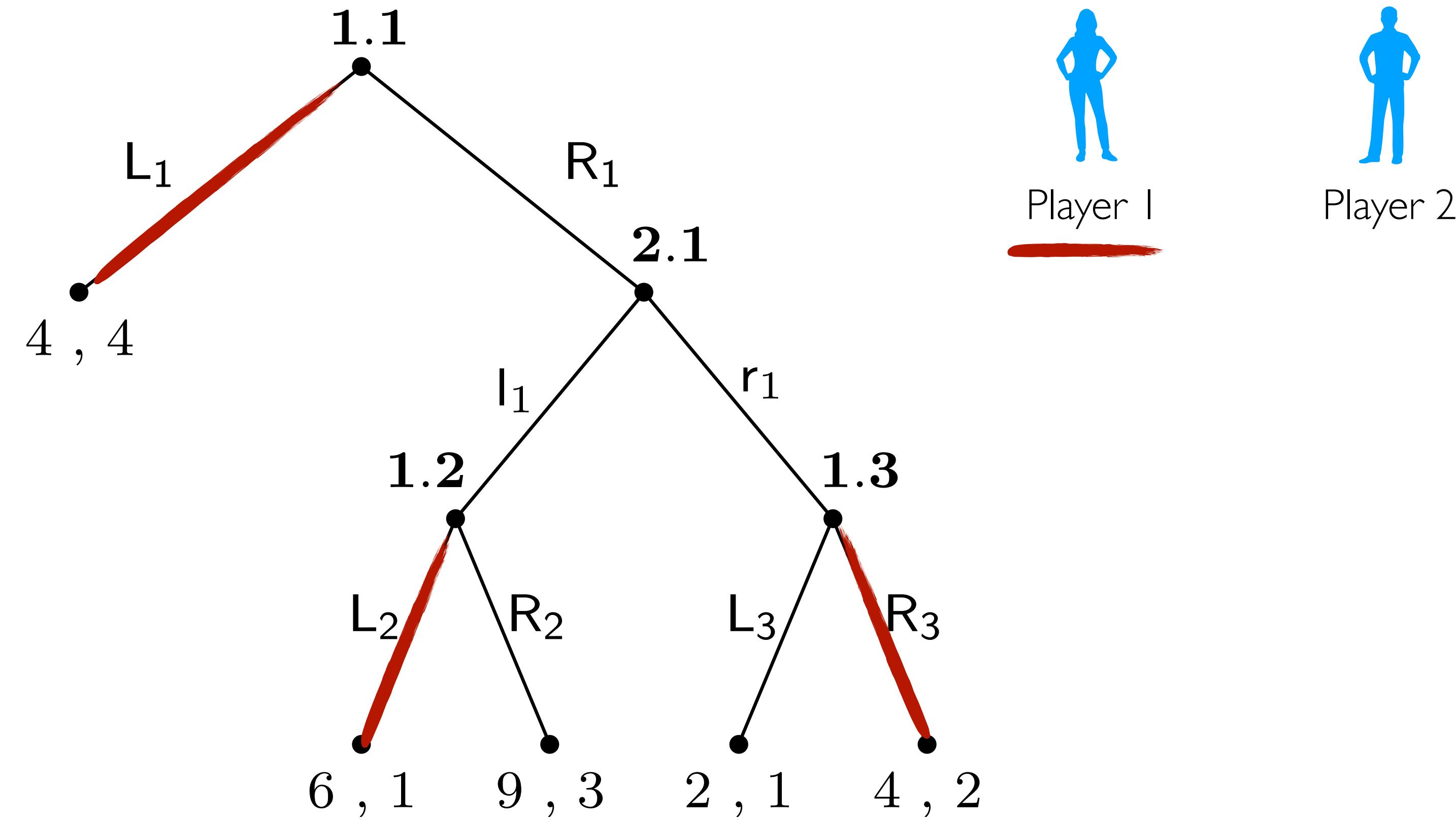
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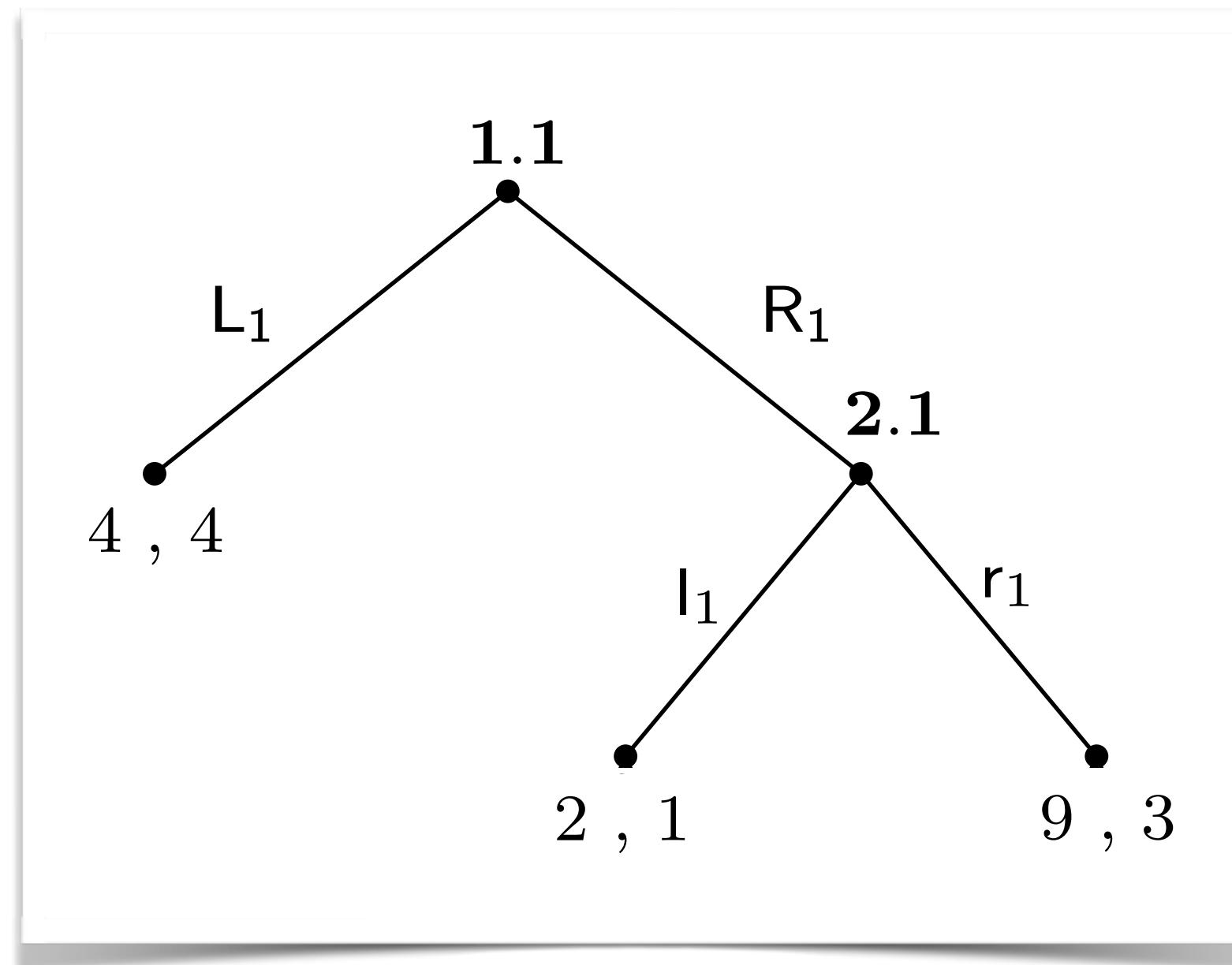


Nash equilibrium

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

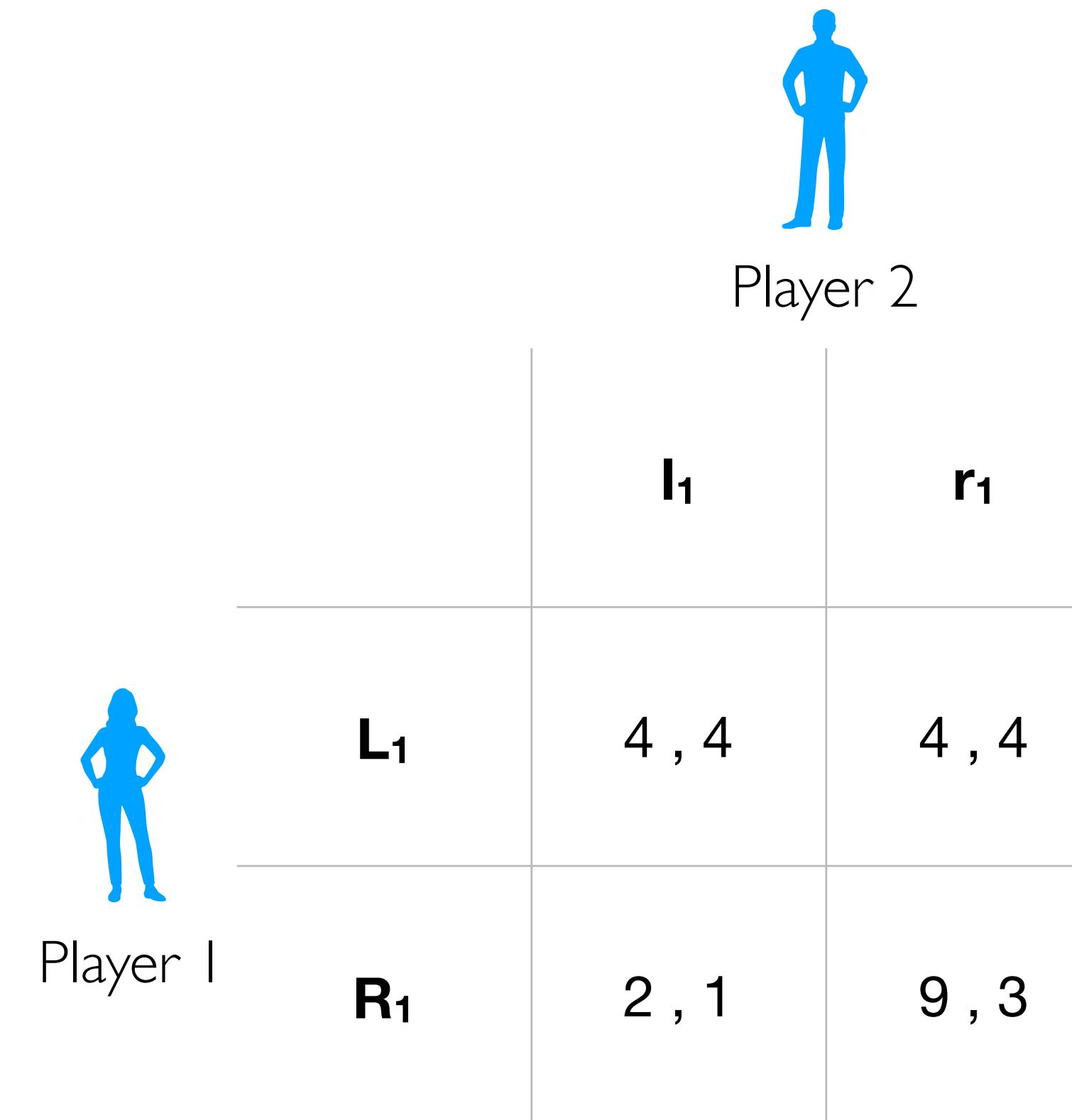
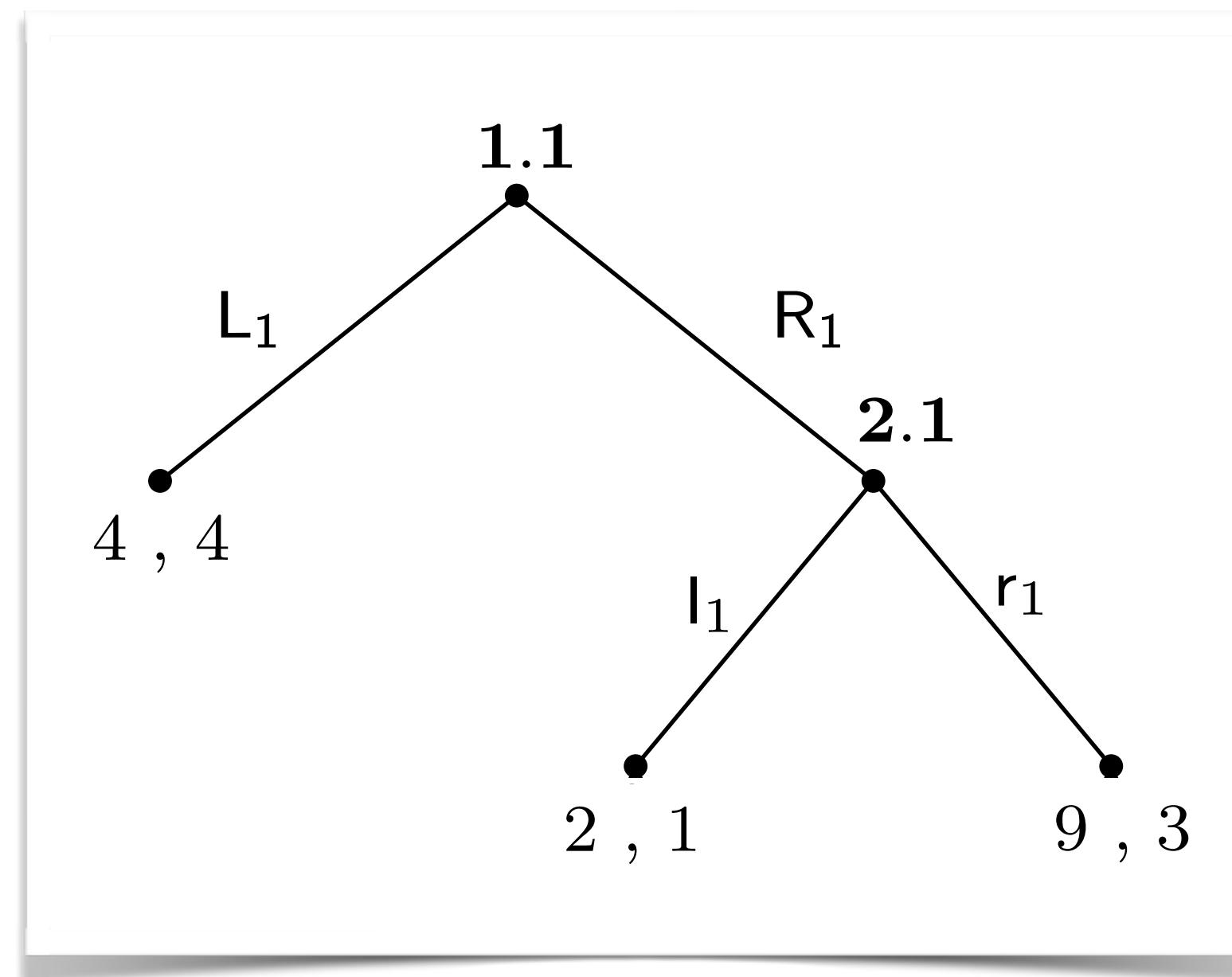
Nash equilibrium

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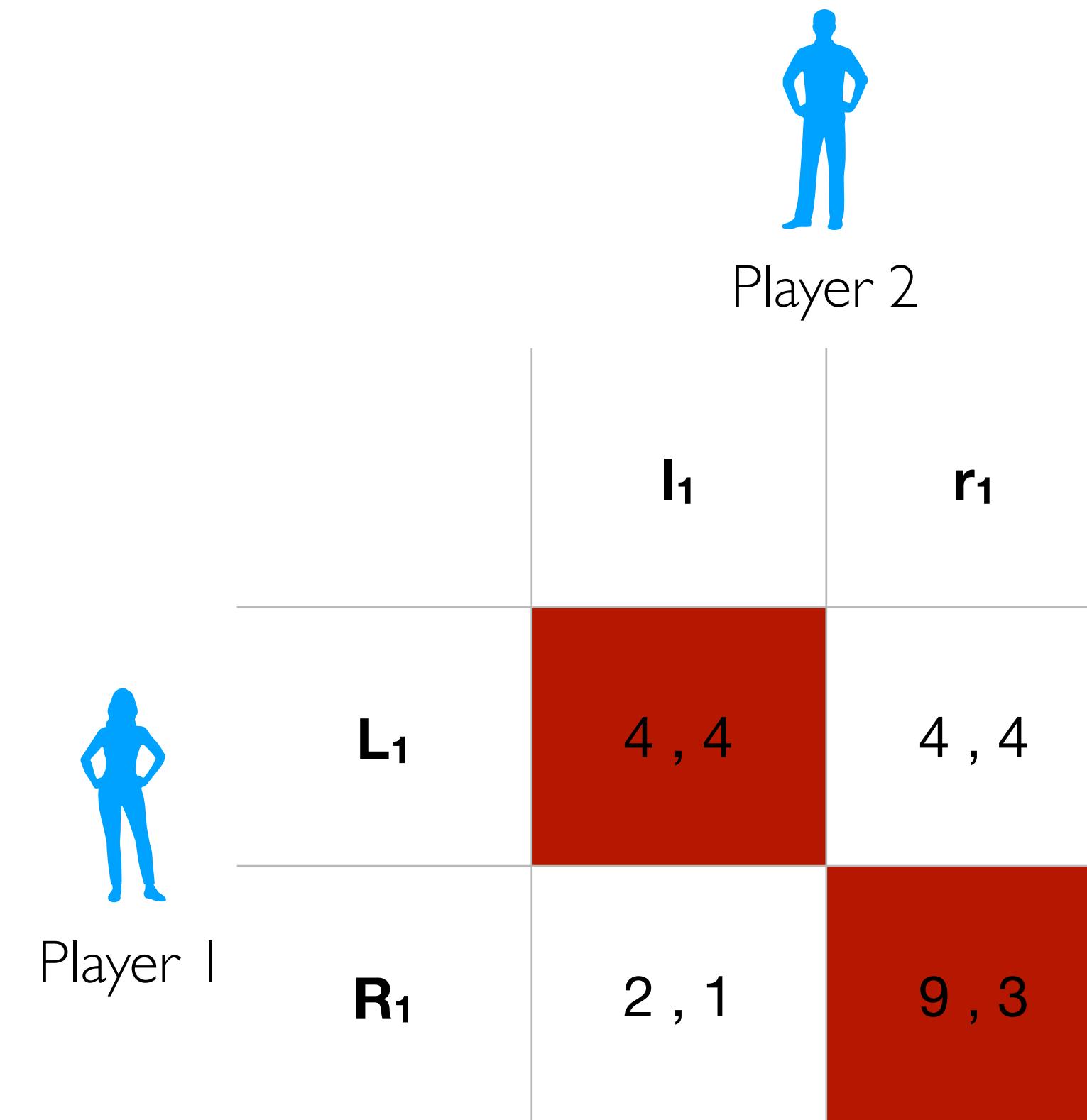
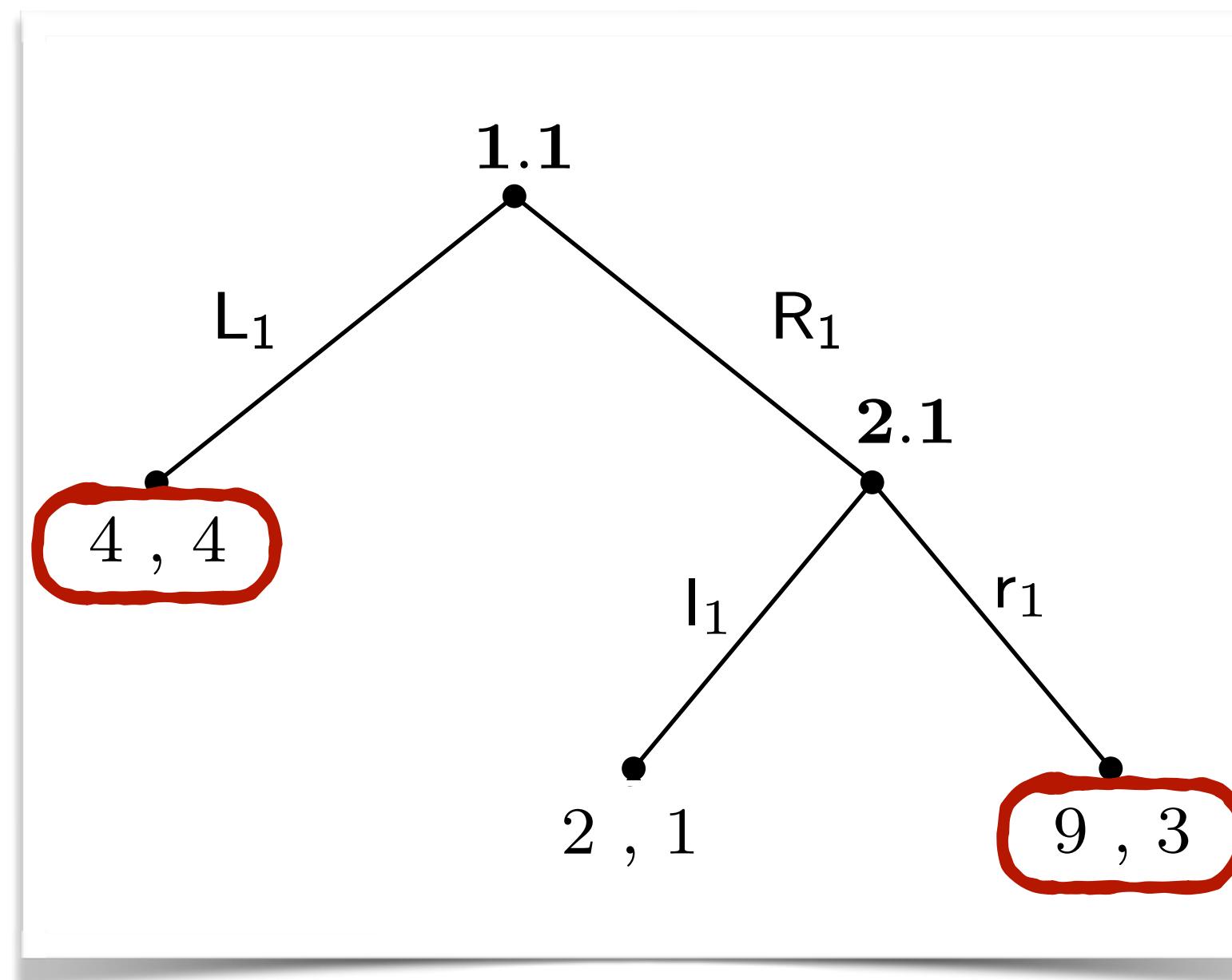
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Nash equilibrium

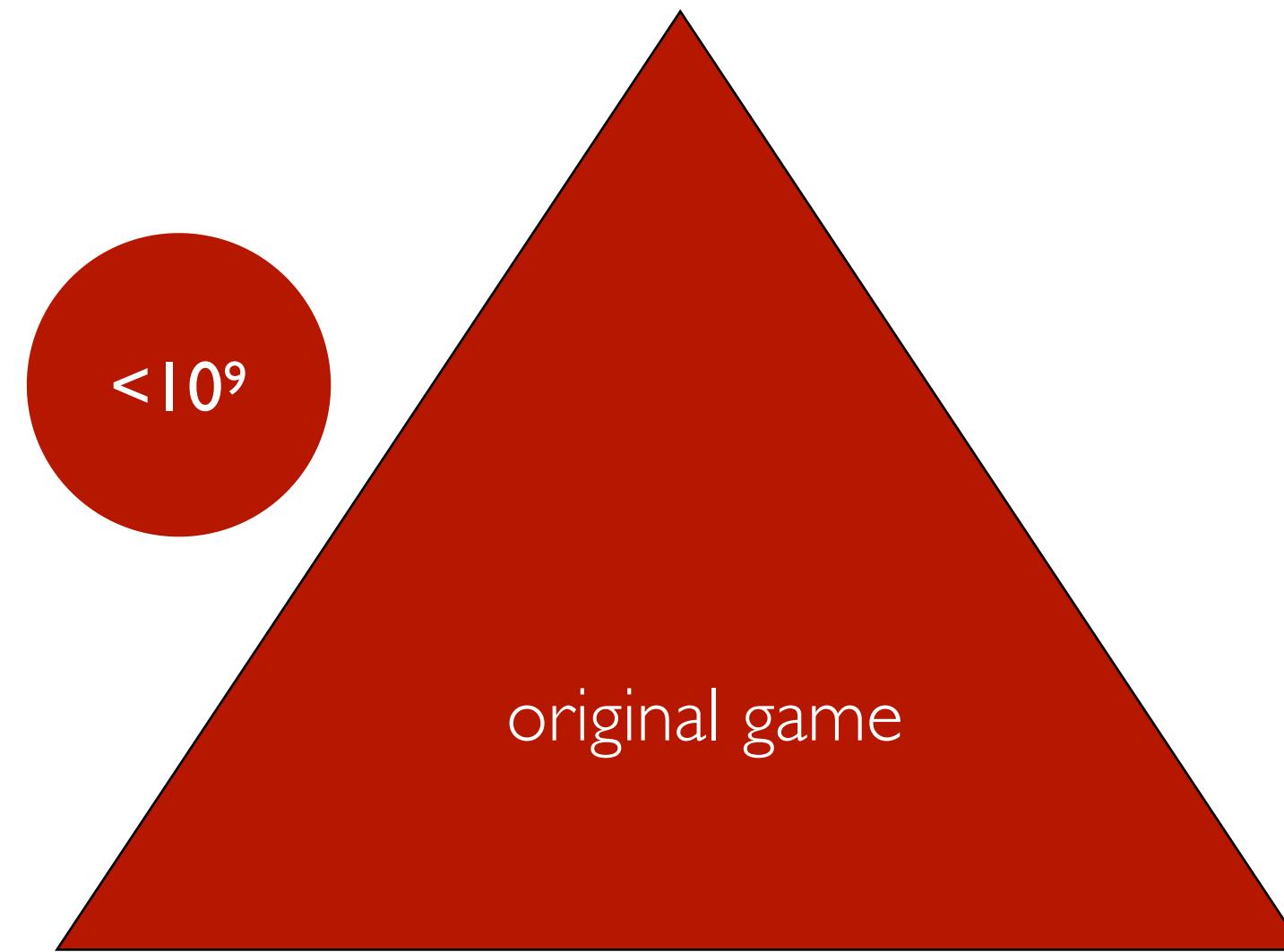
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Nash equilibrium

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

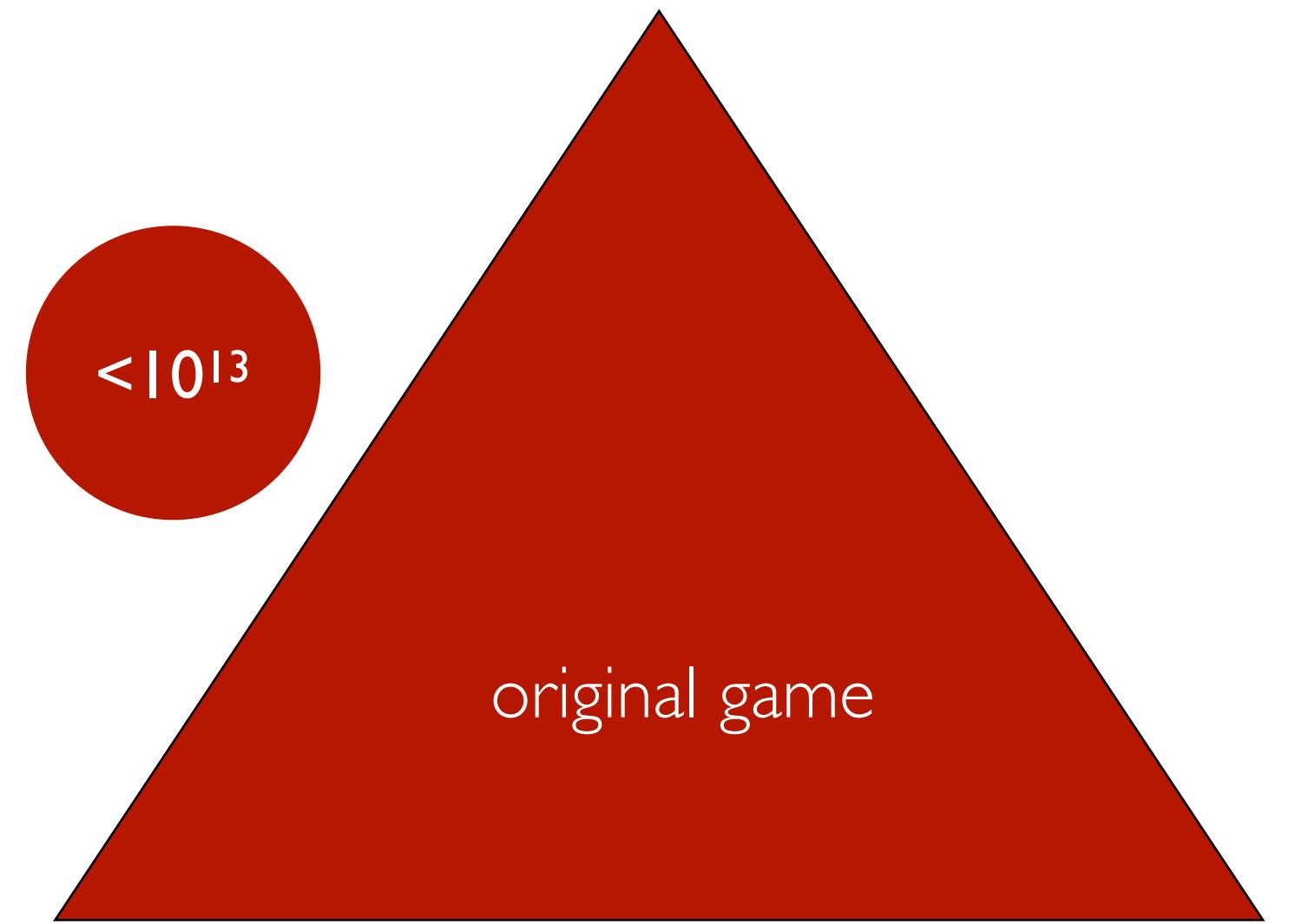
State-of-the-art game solving



Linear programming

(Shi, Littman, 2000)
(Billings et al., 2003)
(Gilpin, Sandholm, 2007)

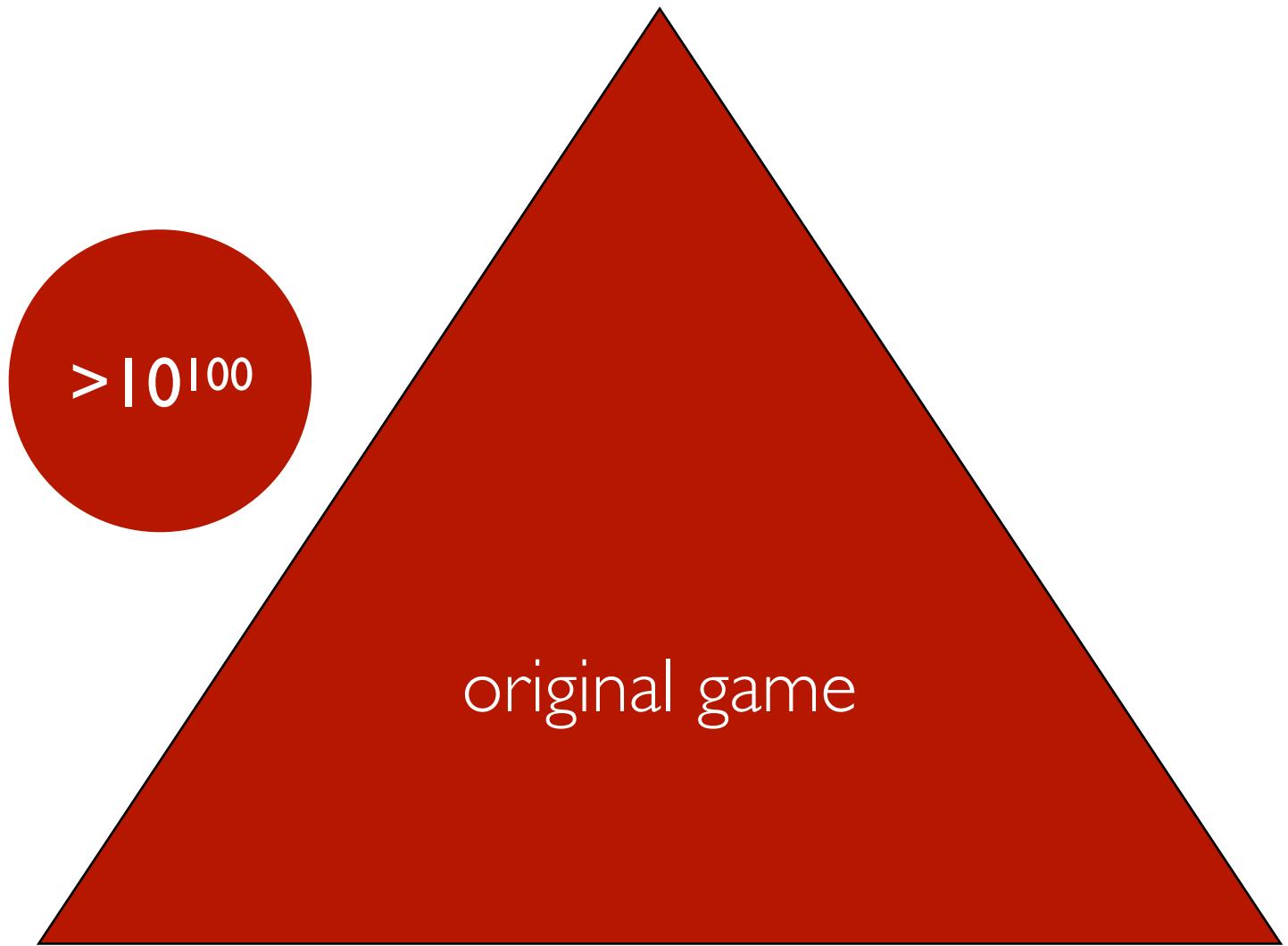
State-of-the-art game solving



No-regret learning

Zinkevich et al., 2008

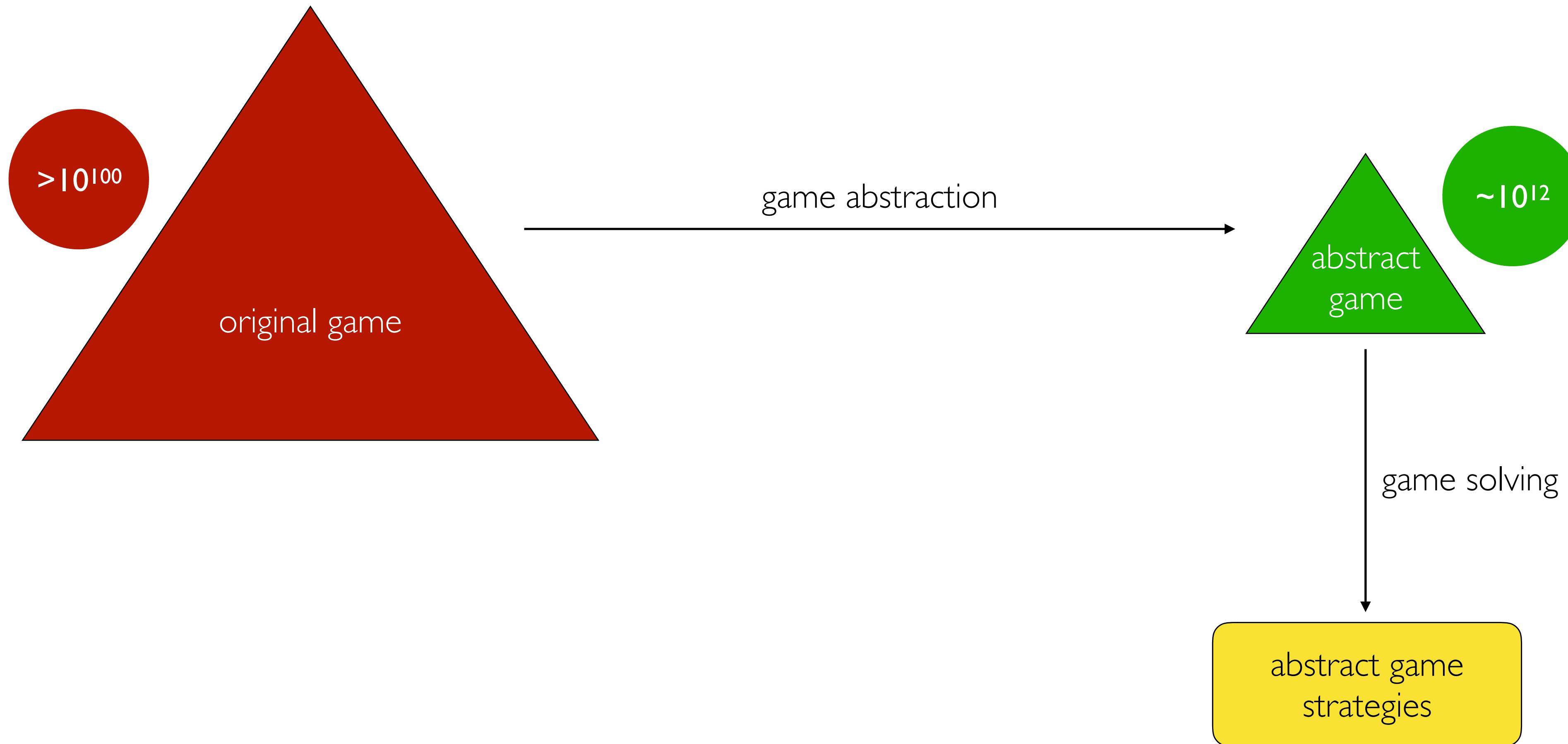
State-of-the-art huge-game solving: *preplay*



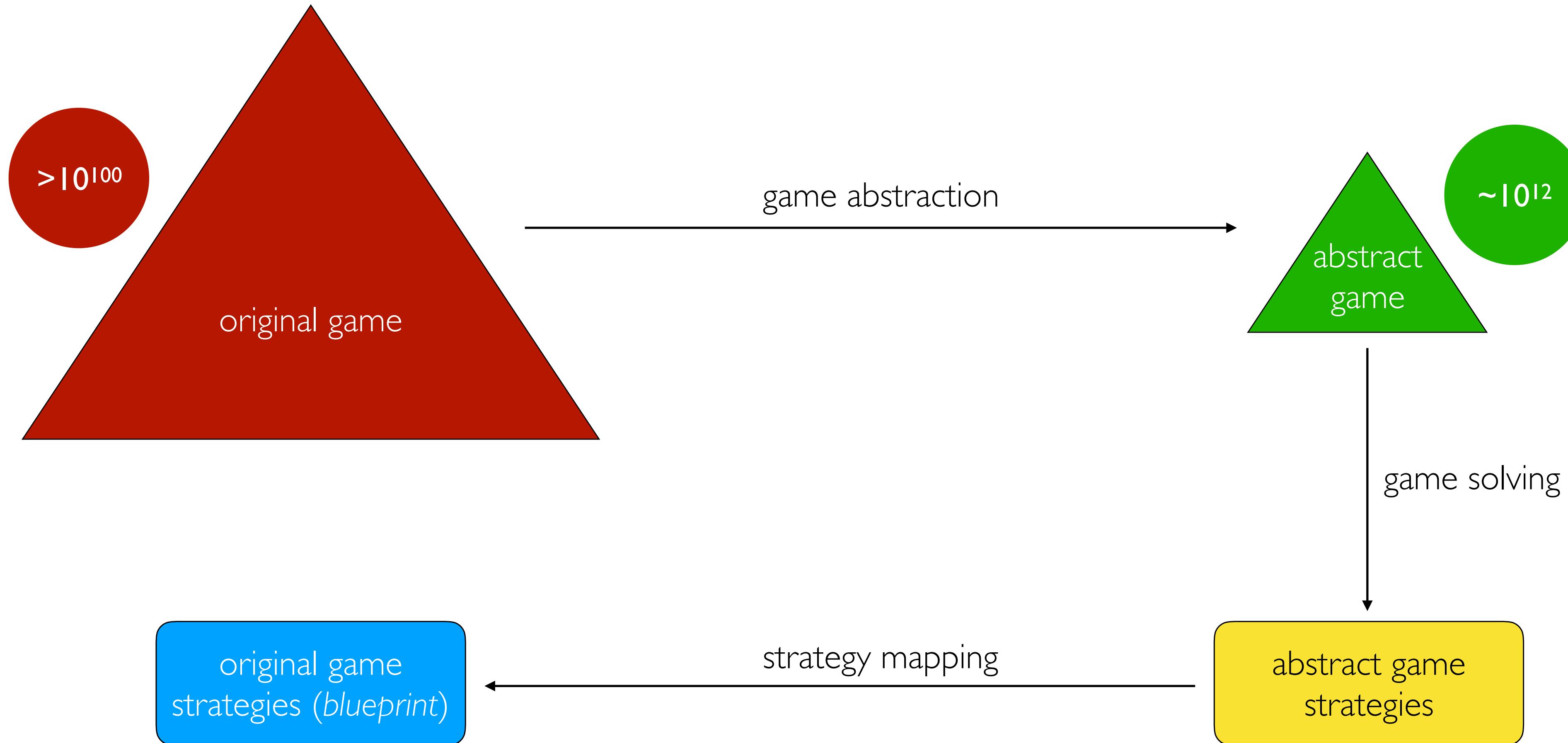
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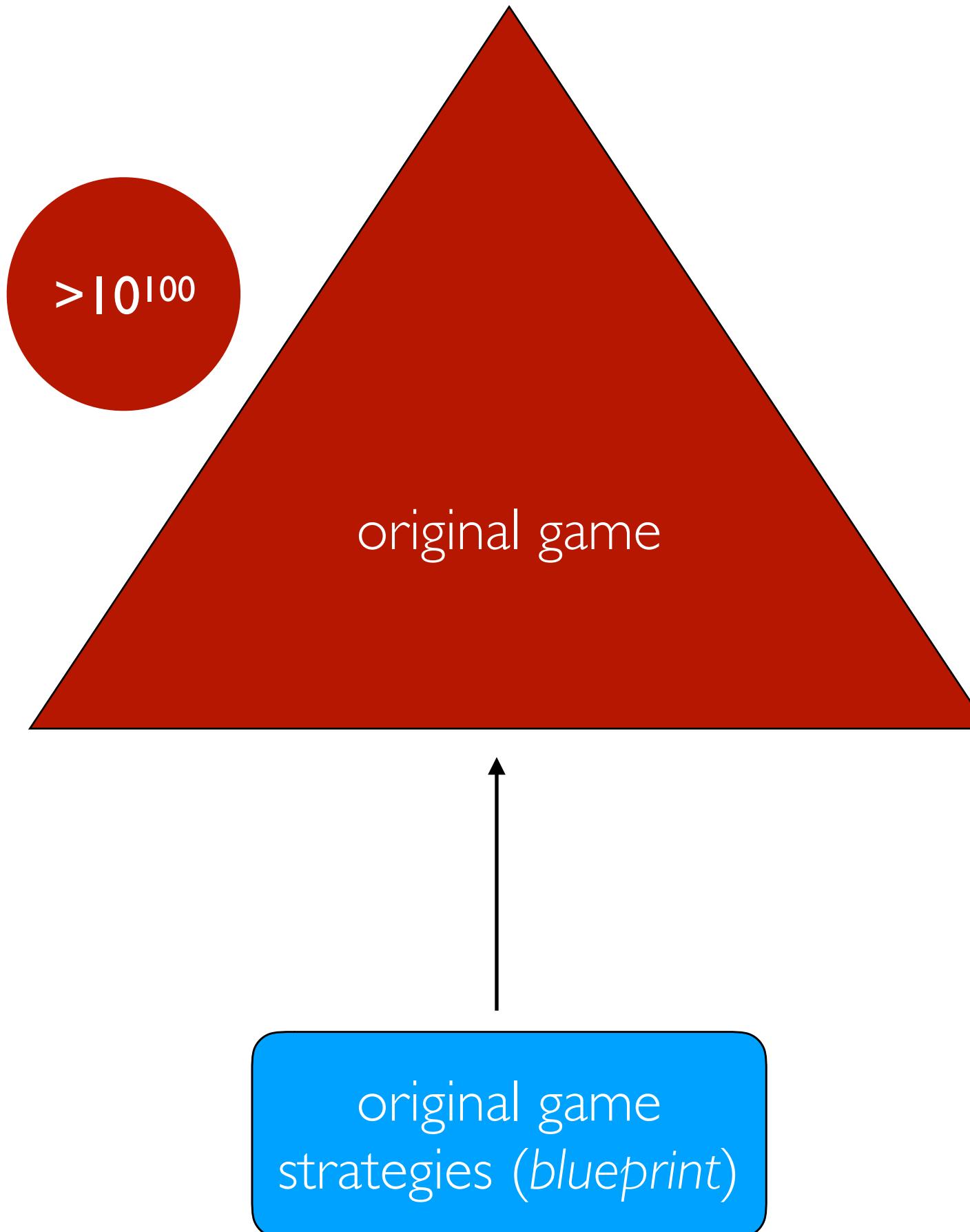
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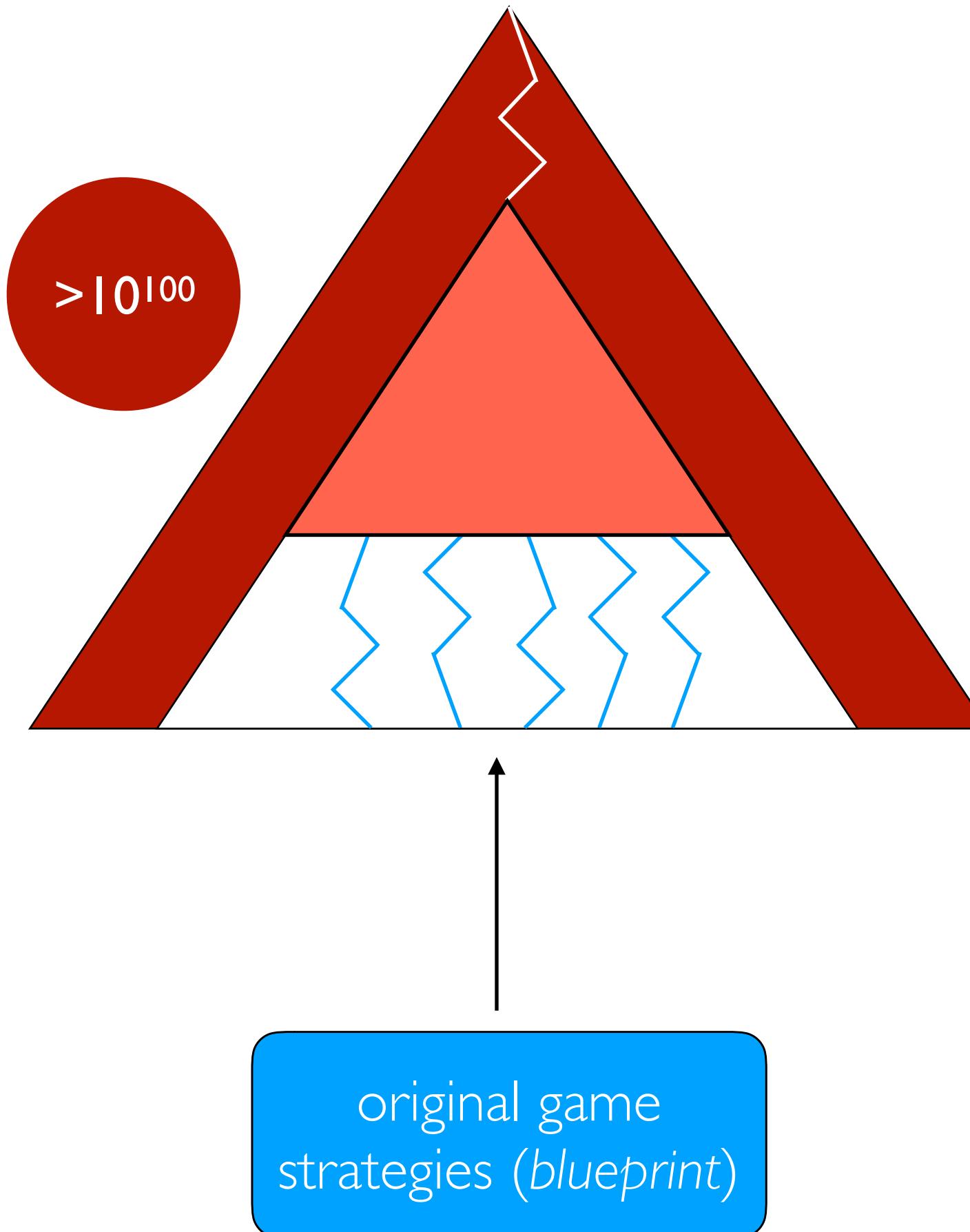
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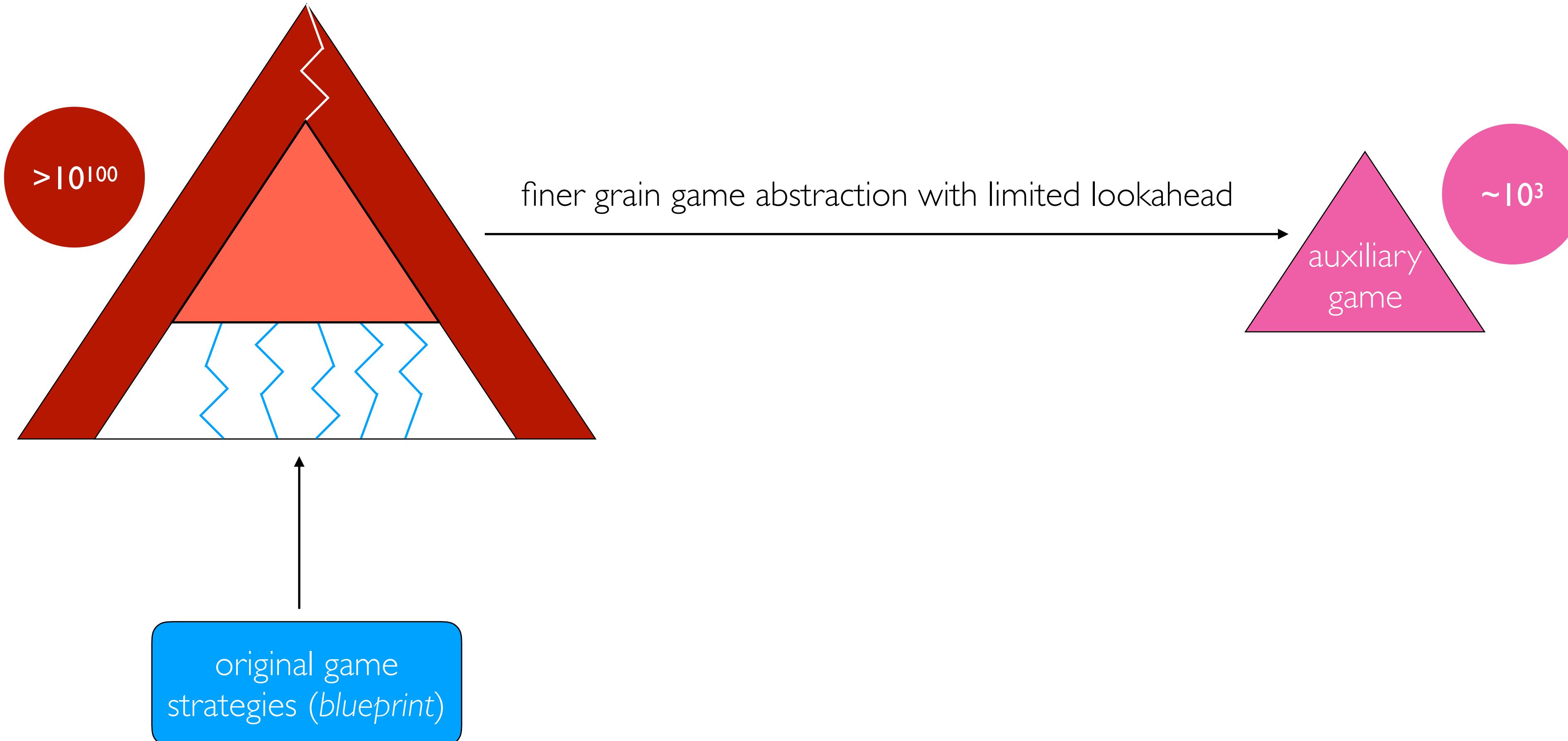
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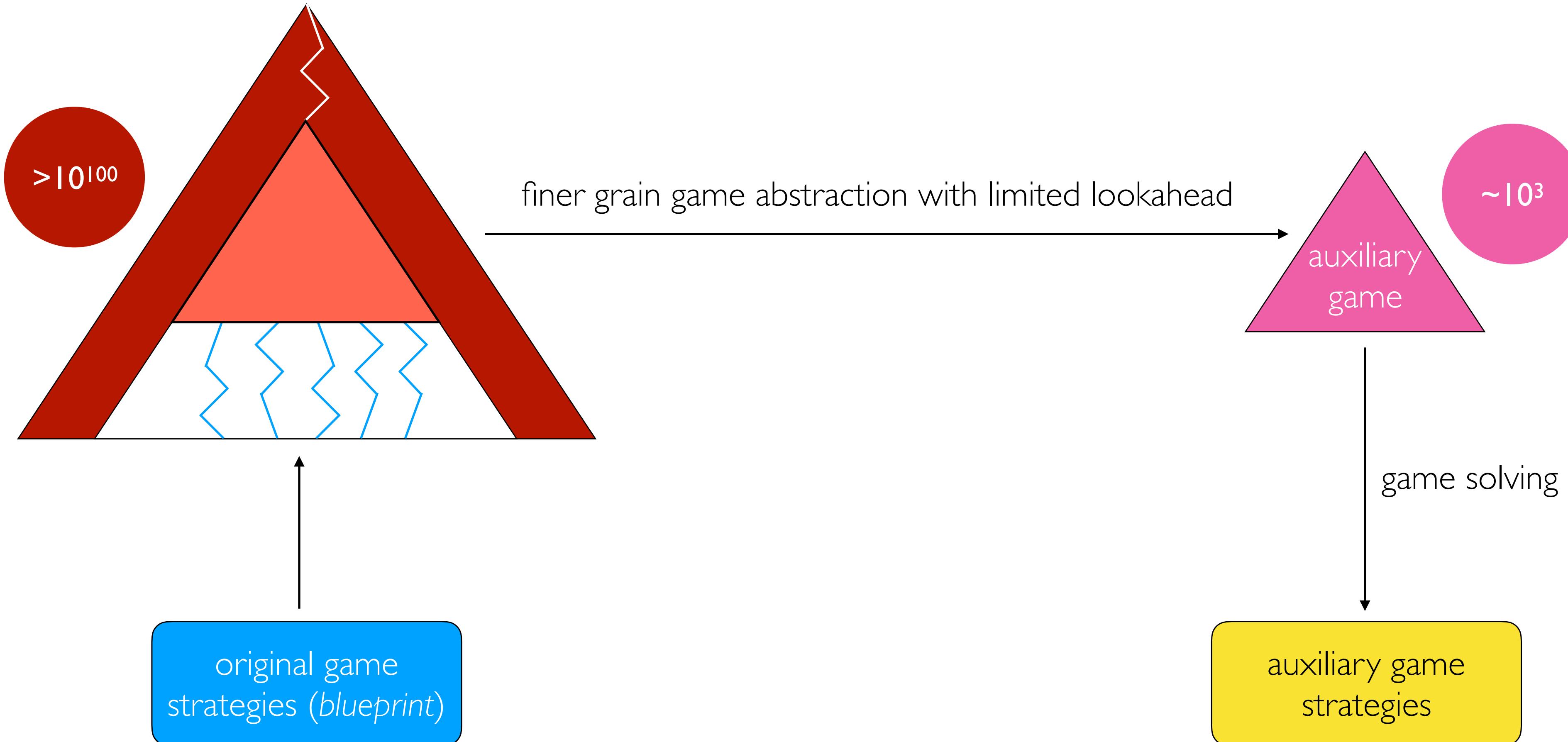
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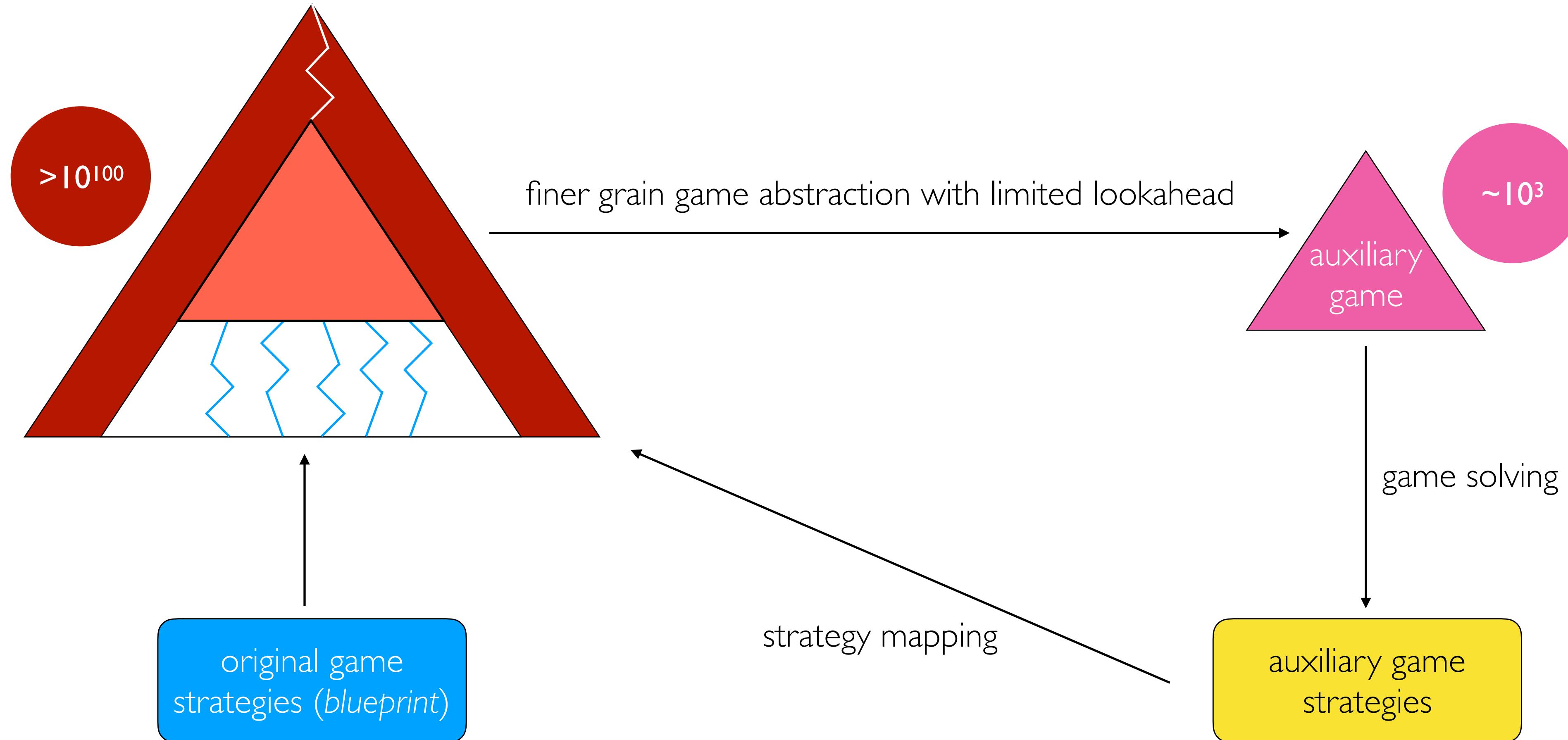
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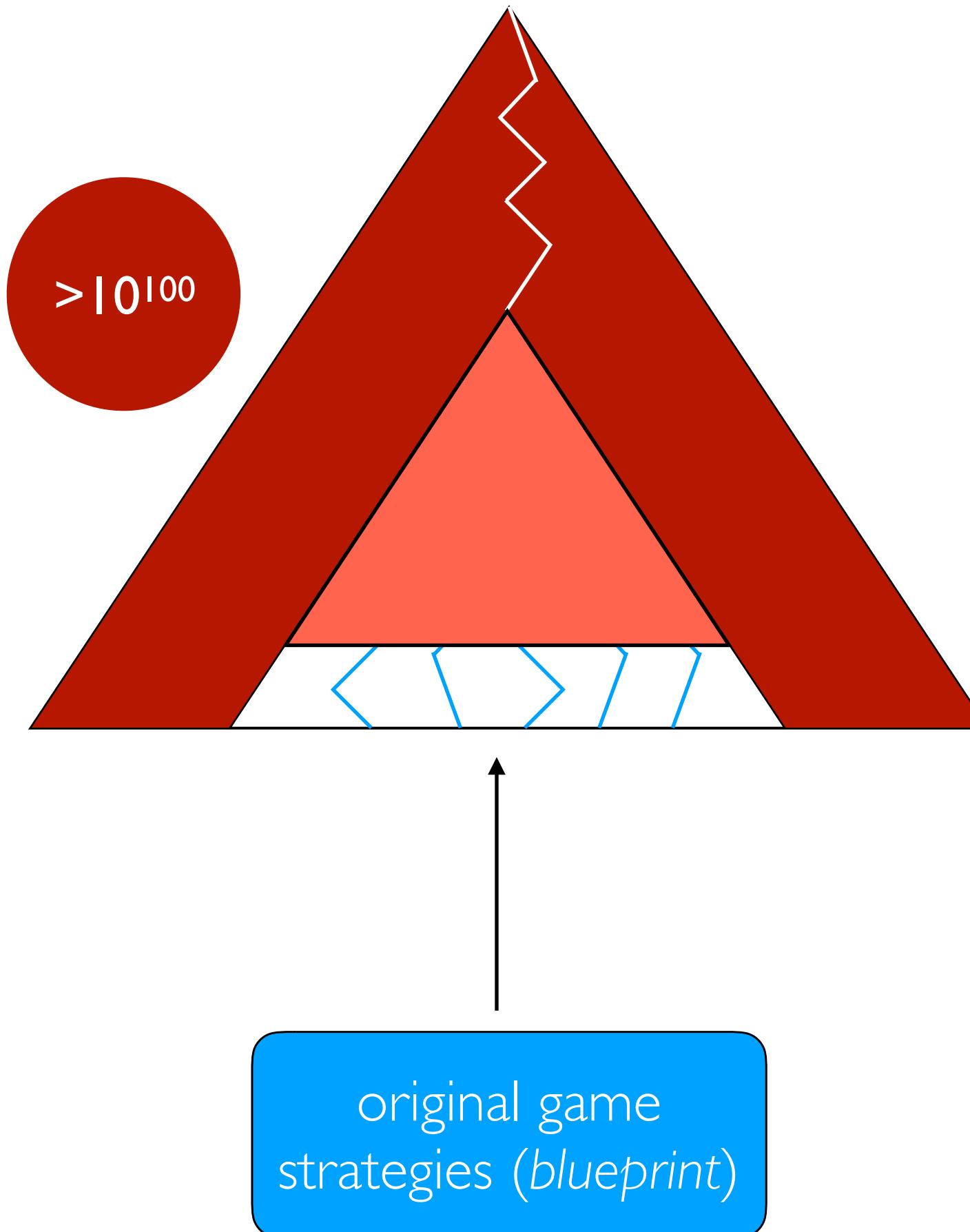
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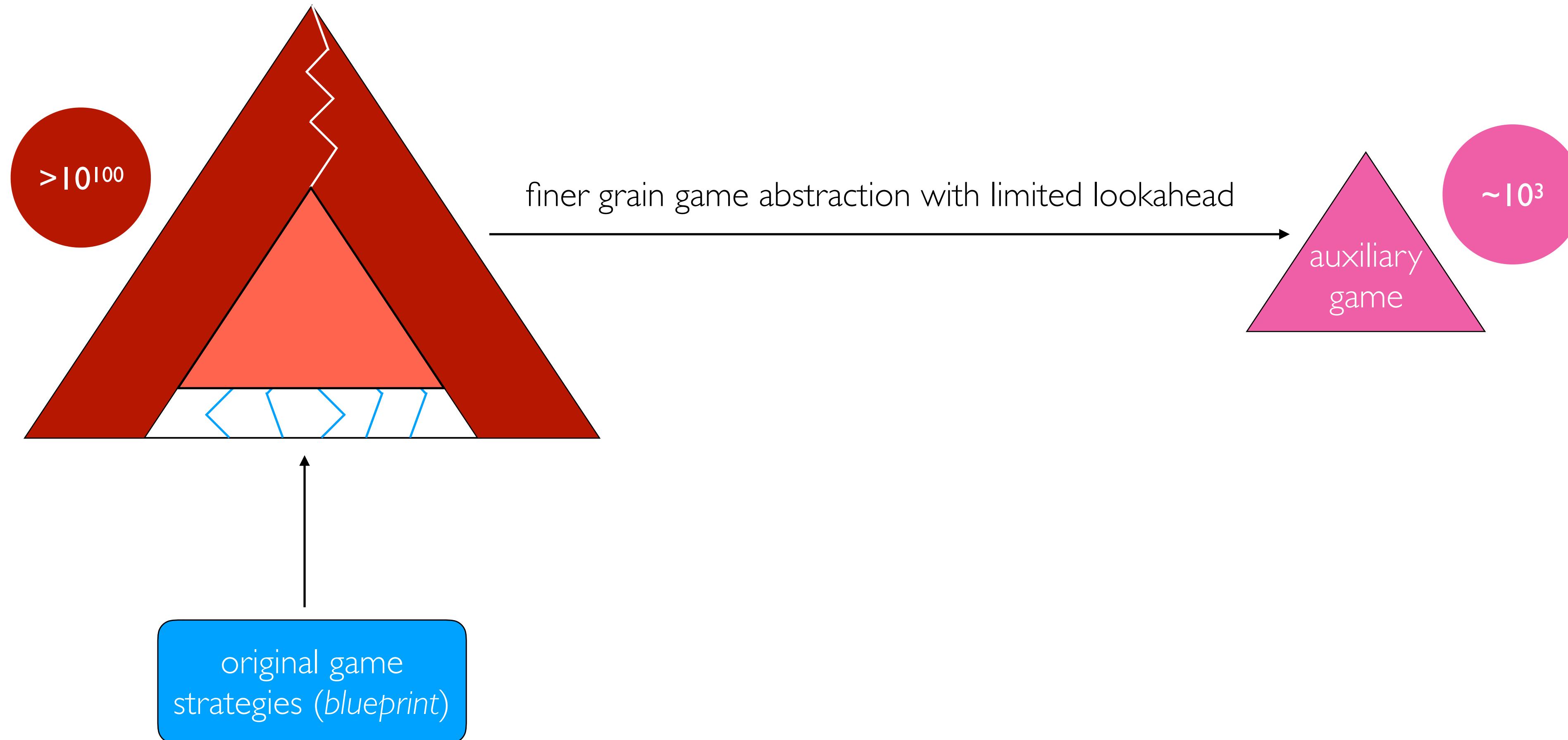
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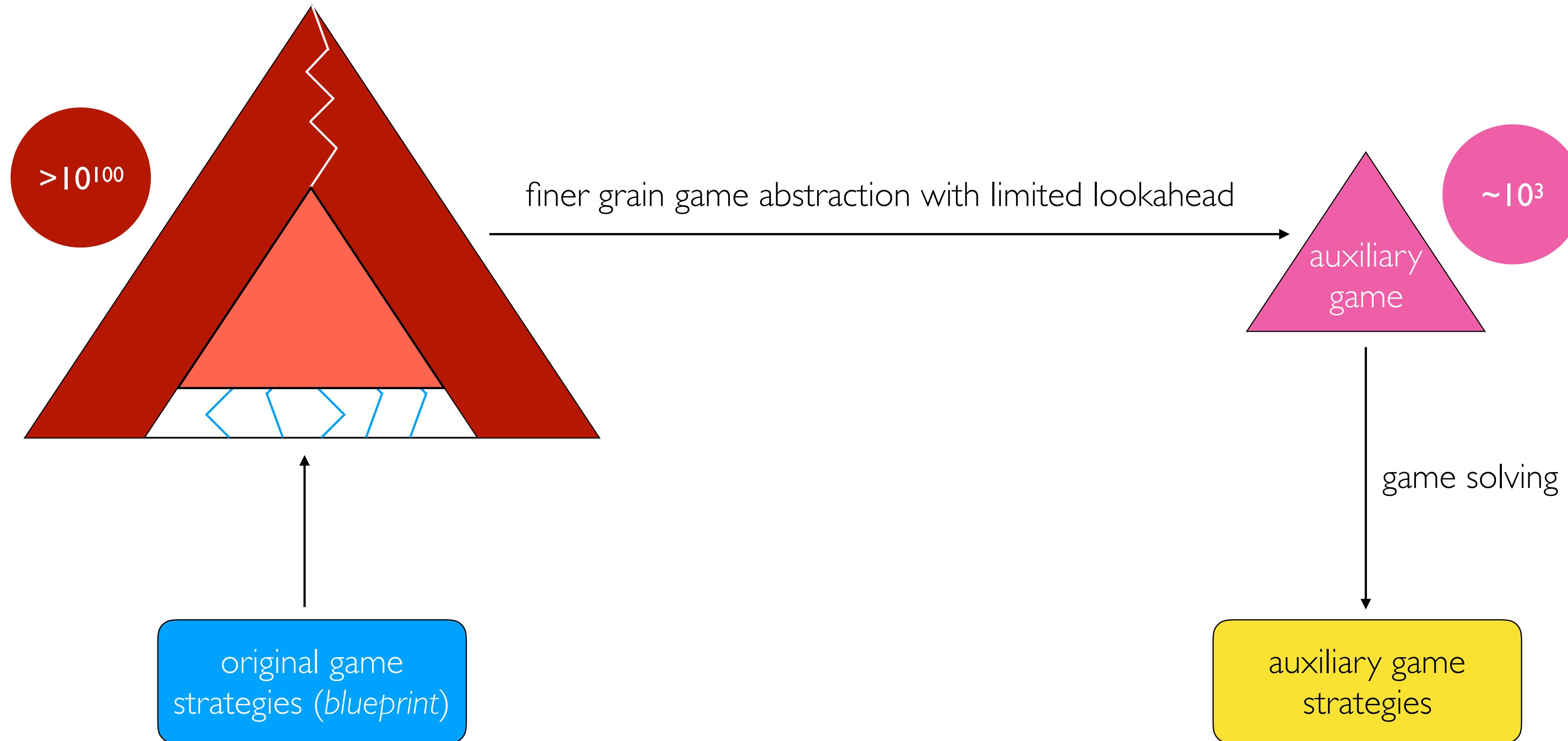
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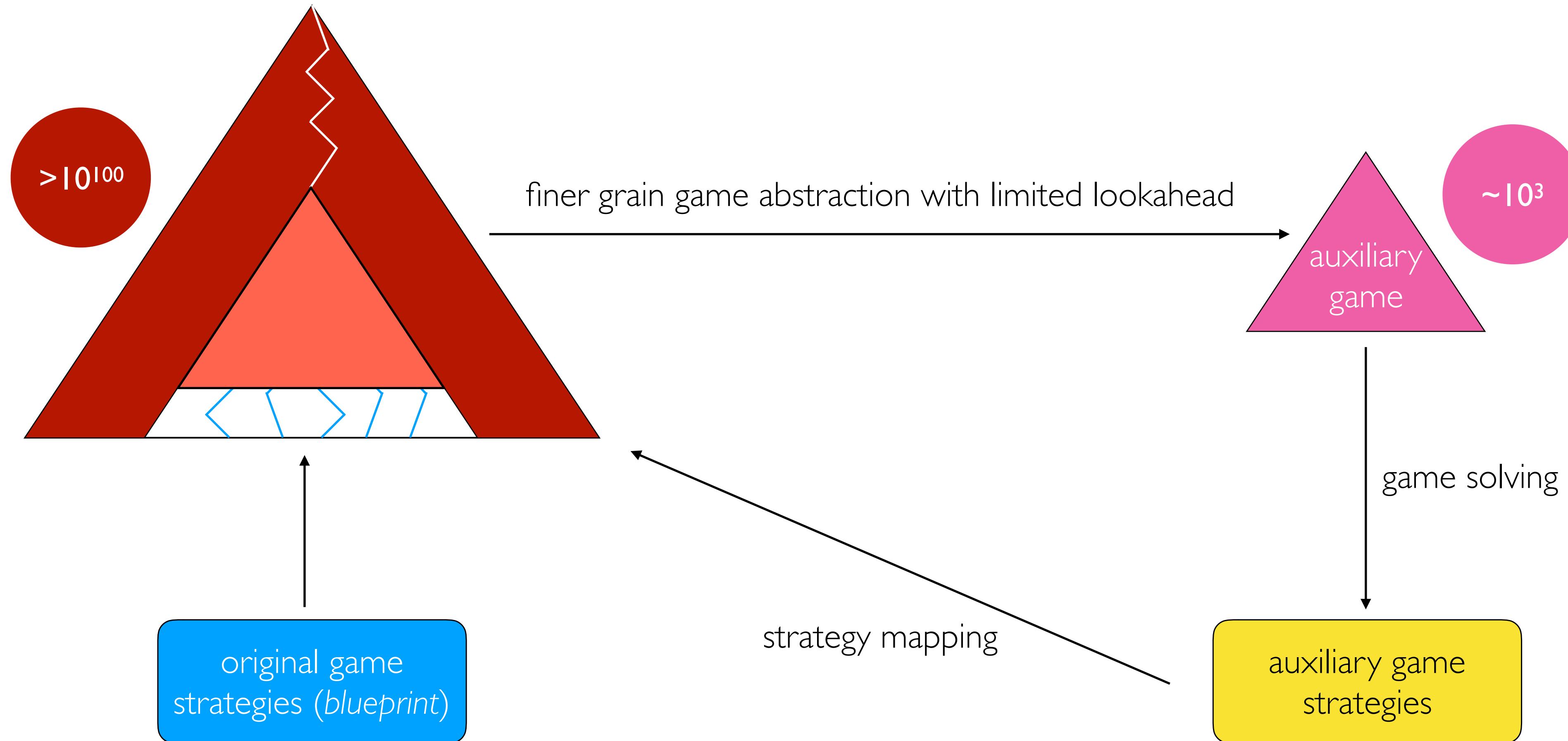
State-of-the-art huge-game solving: *play*



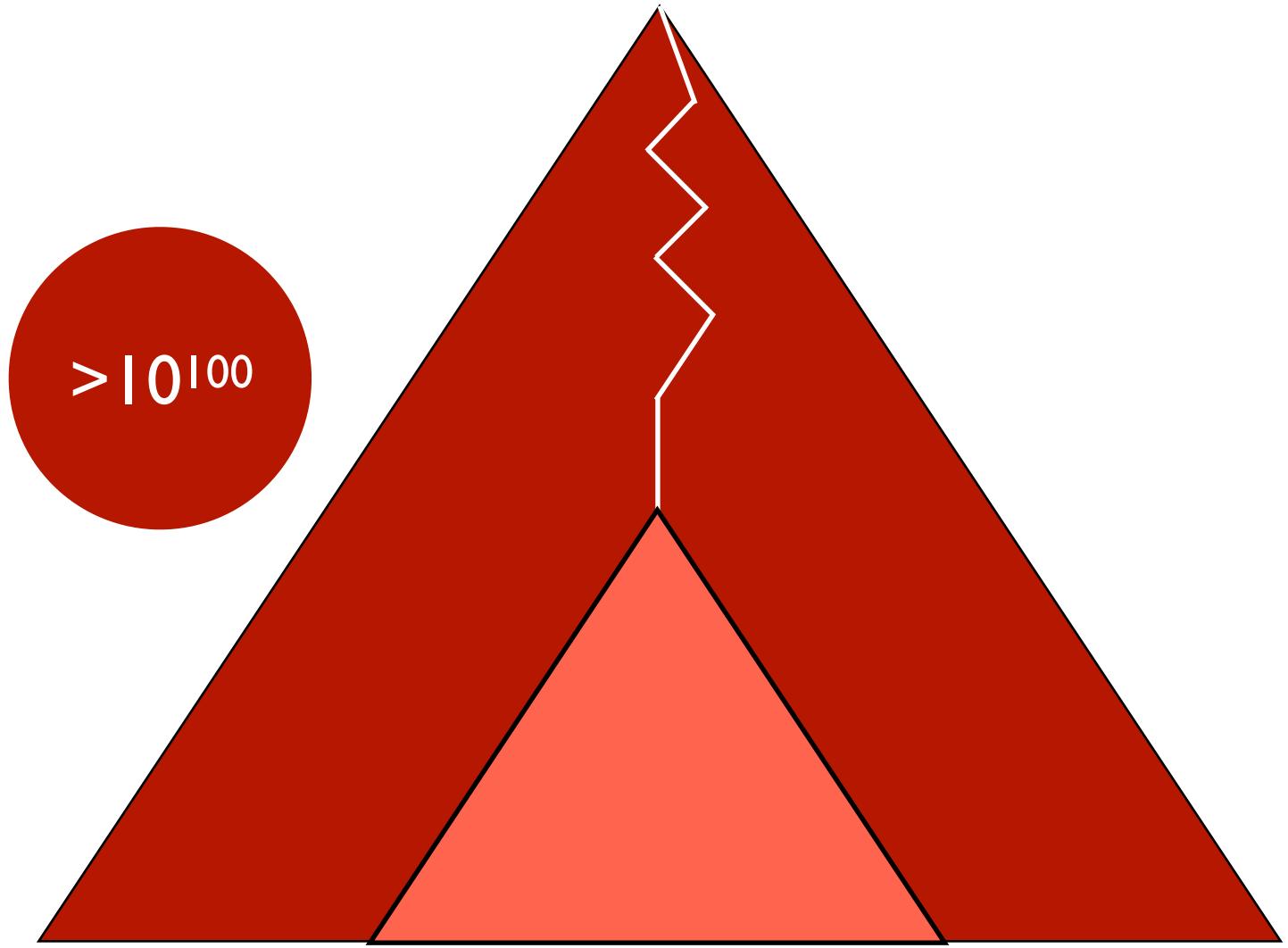
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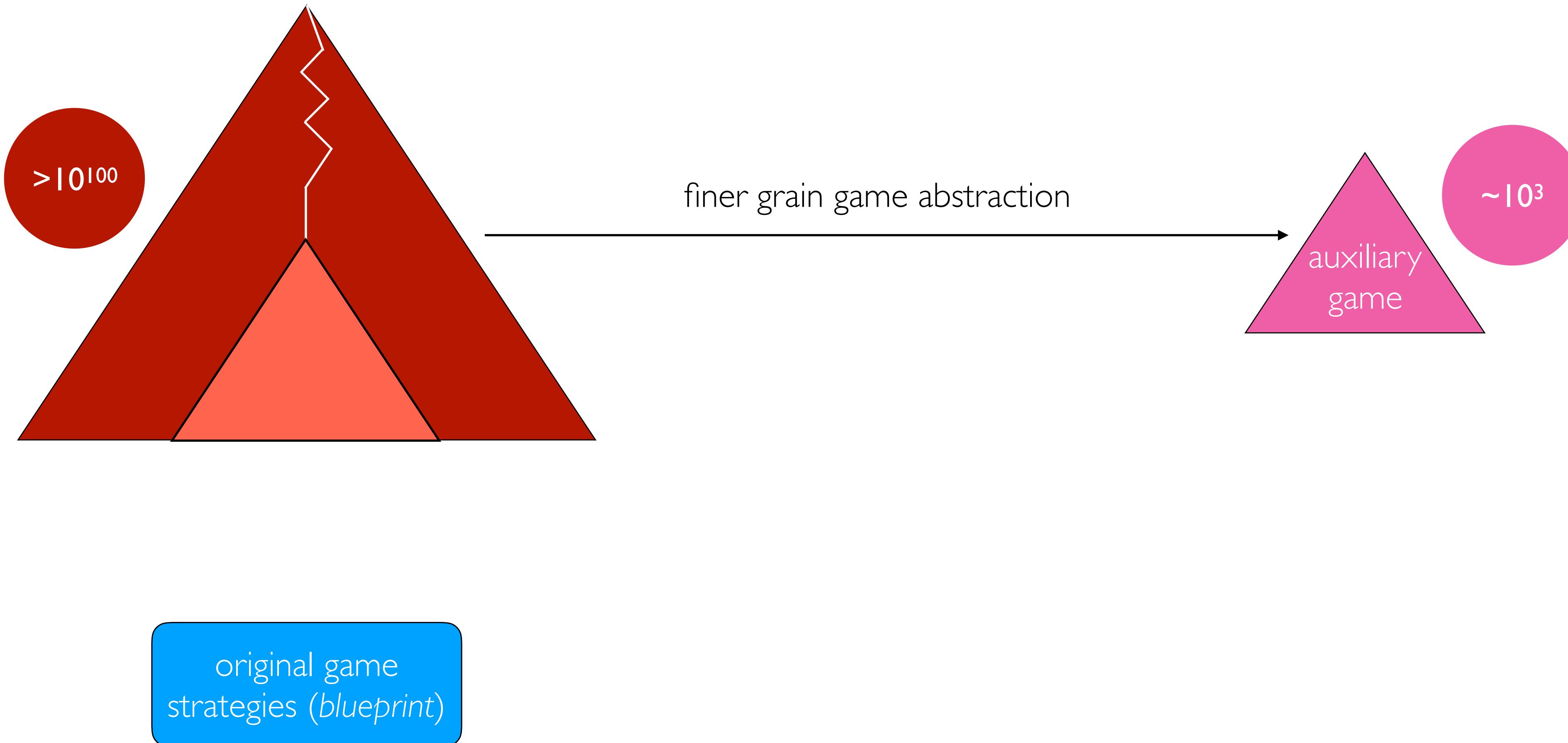


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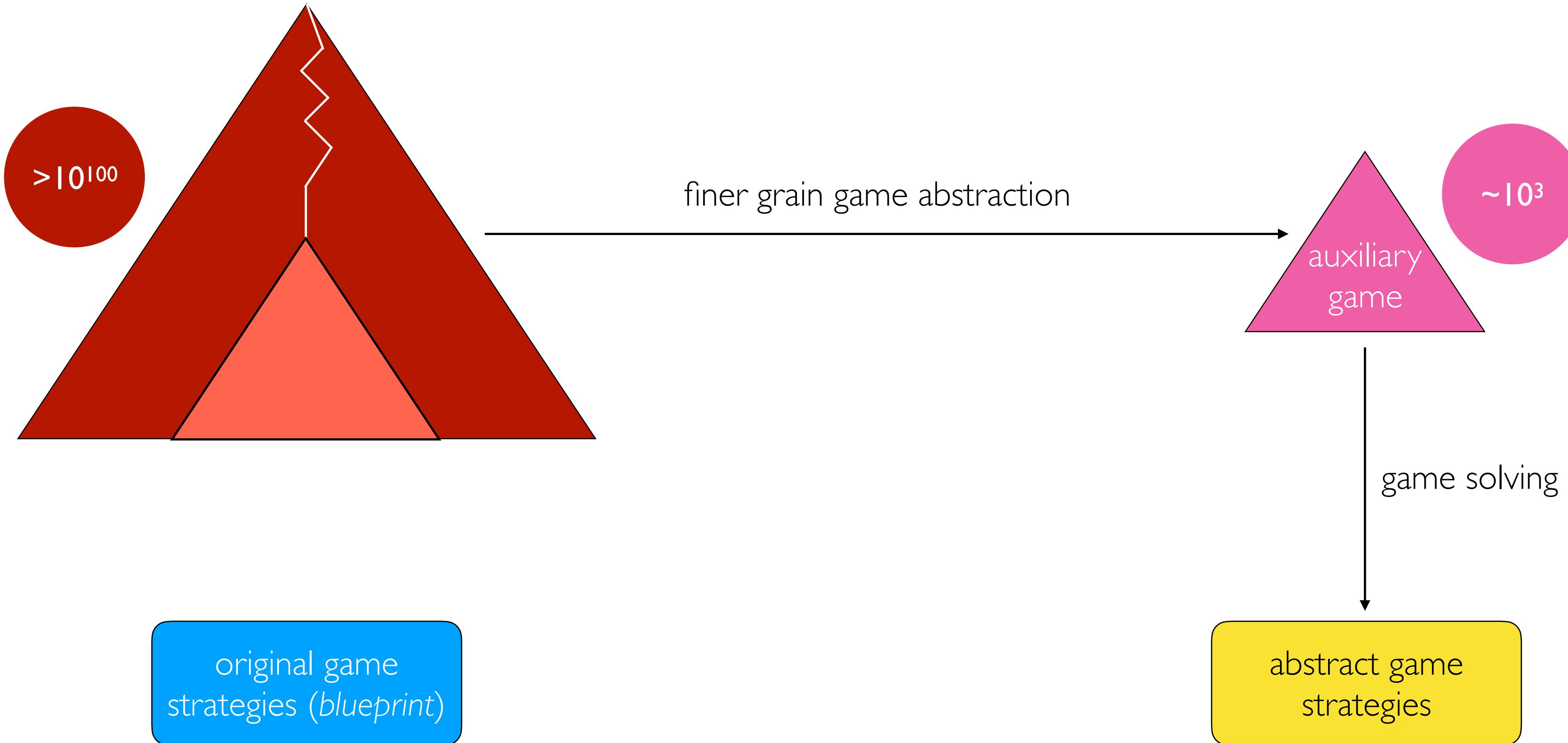


original game
strategies (*blueprint*)

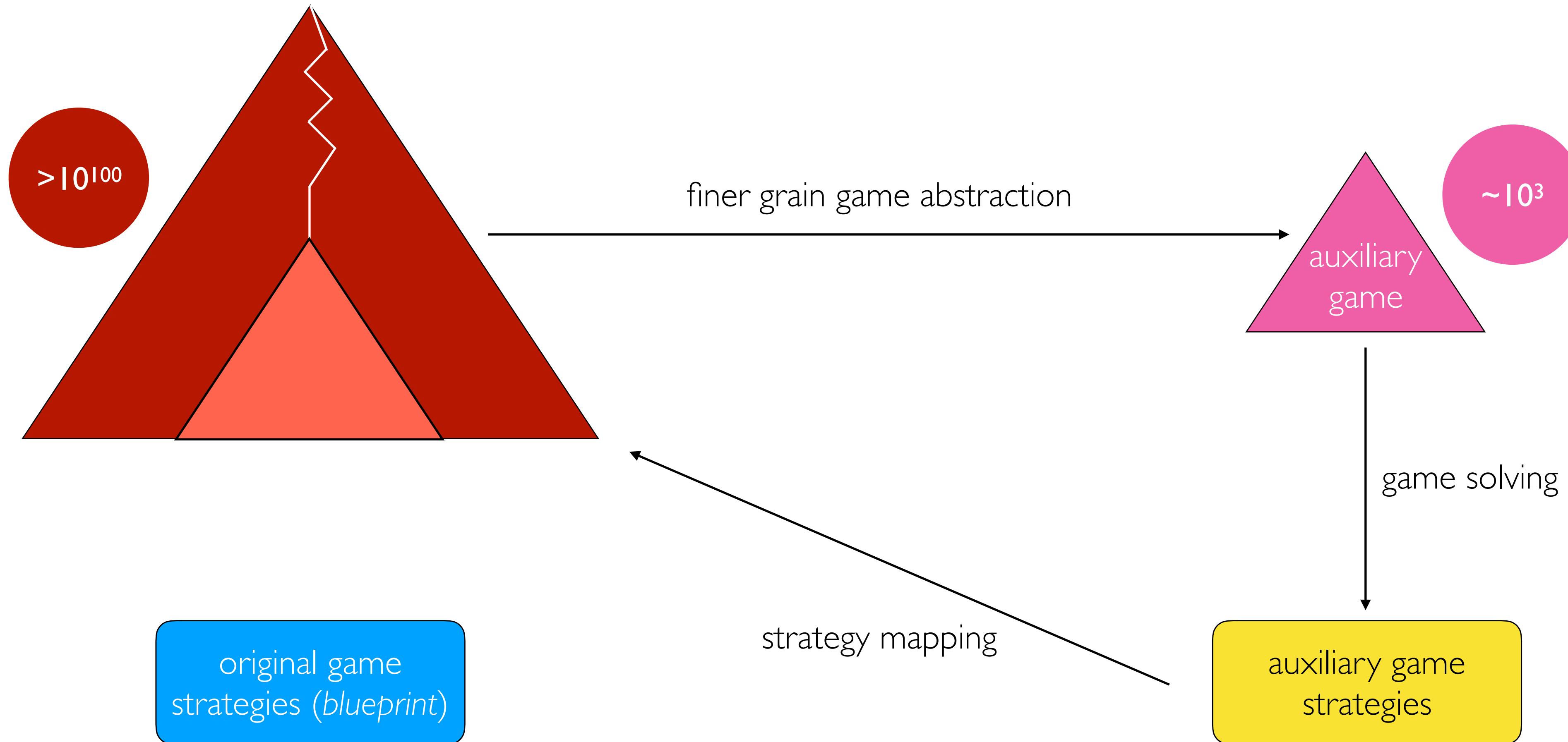
State-of-the-art huge-game solving: *play*



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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 games*, 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



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

Beyond abstractions



Libratus, 2017



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

A screenshot of a news article from Facebook's Artificial Intelligence Research blog. The title reads "Facebook, Carnegie Mellon build first AI that beats pros in 6-player poker". Below the title, it says "July 11, 2019 Written by Noam Brown". On the right side, there is a small image of a poker table showing cards and chips, and social sharing icons for Facebook and Twitter.

Pluribus, 2019



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

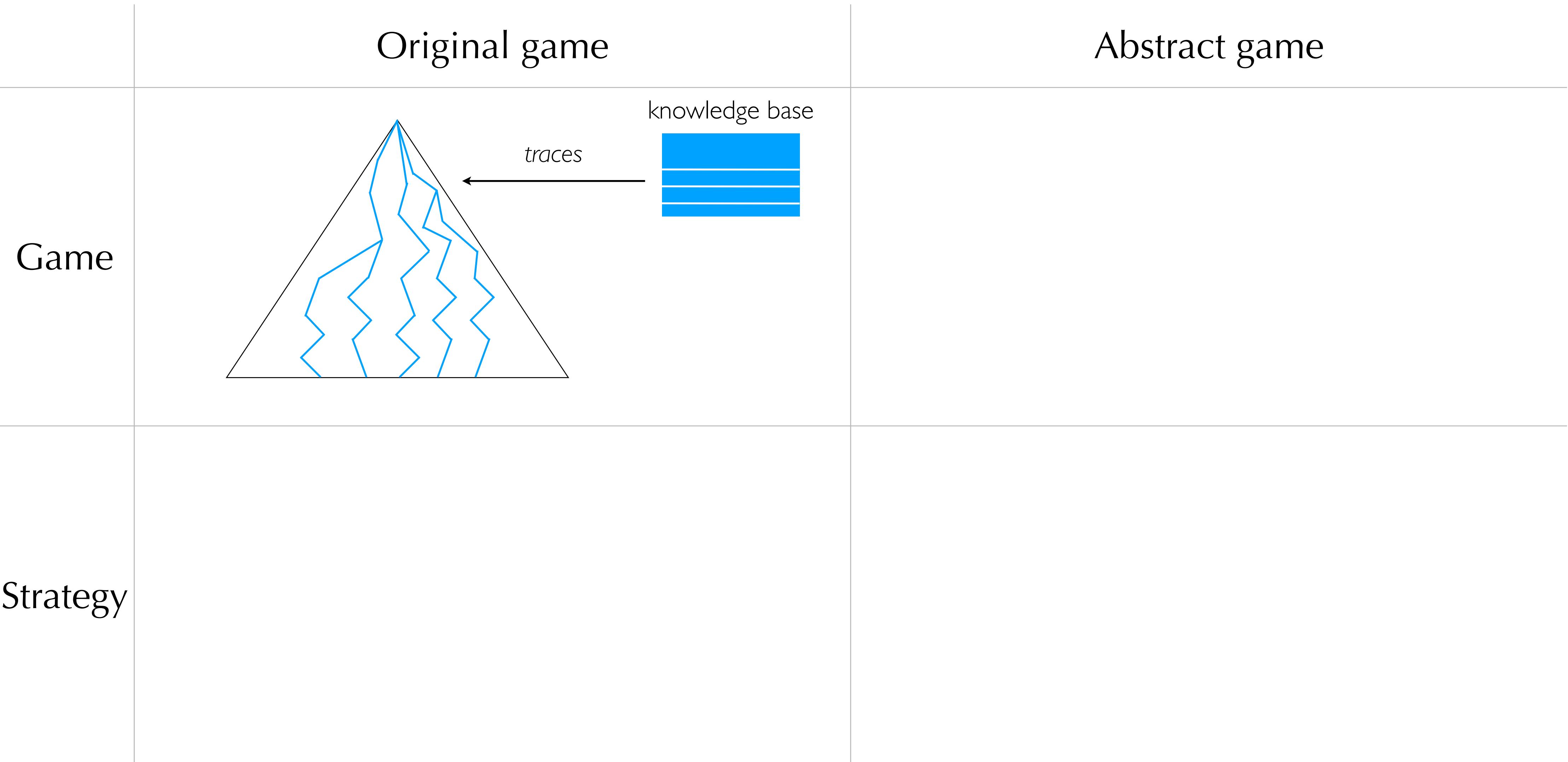
Our goal

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.

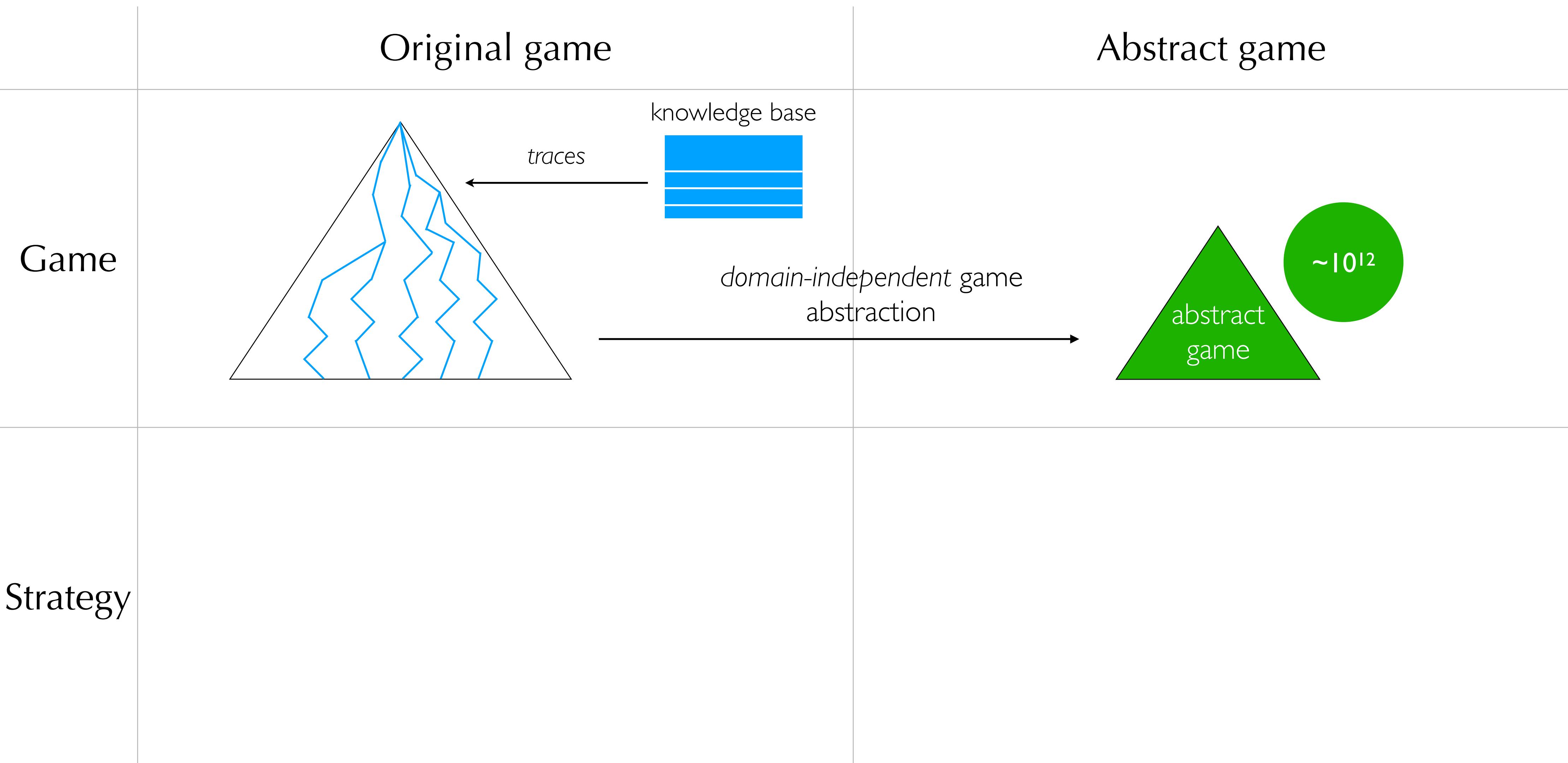
Idea



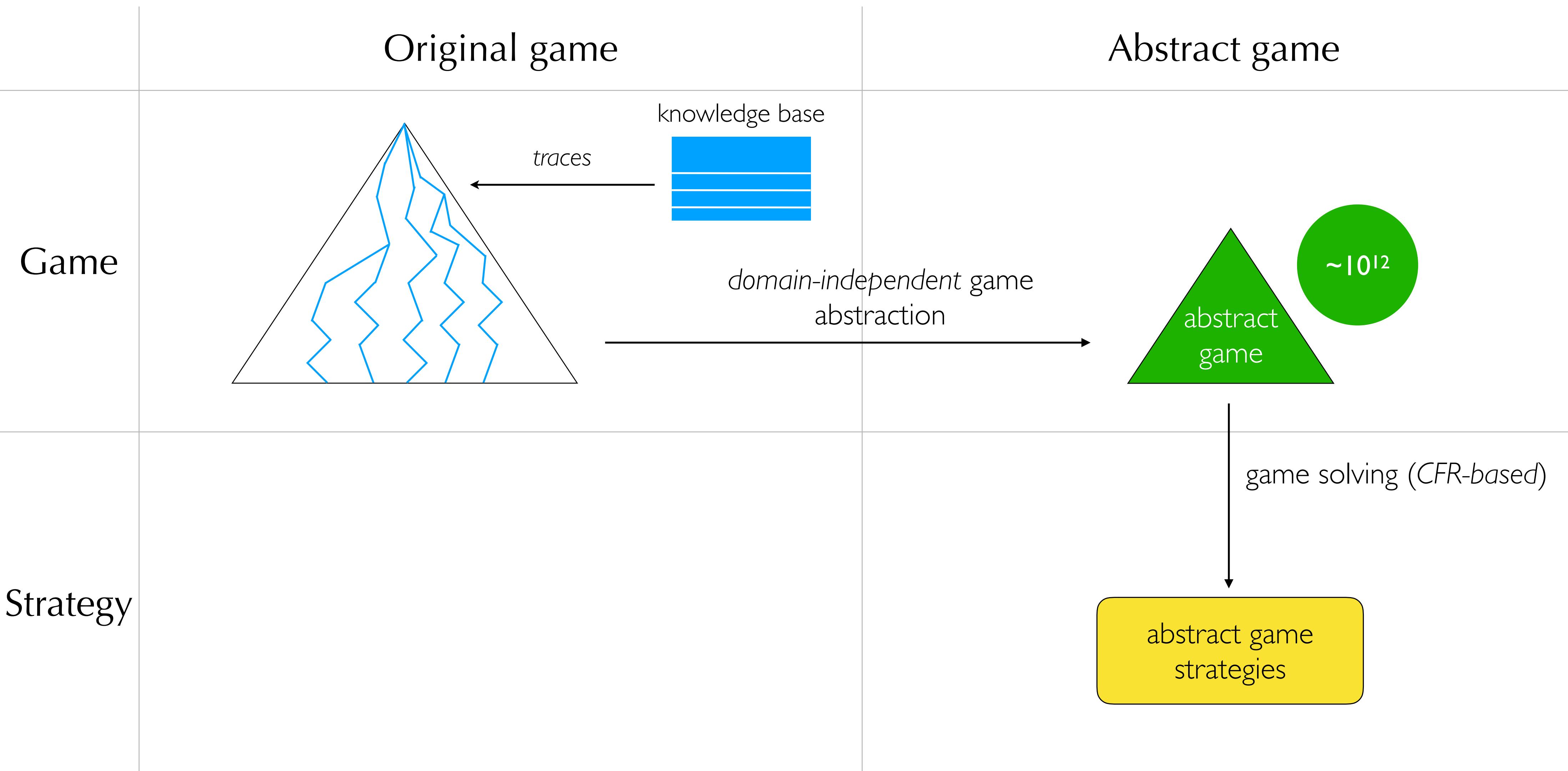
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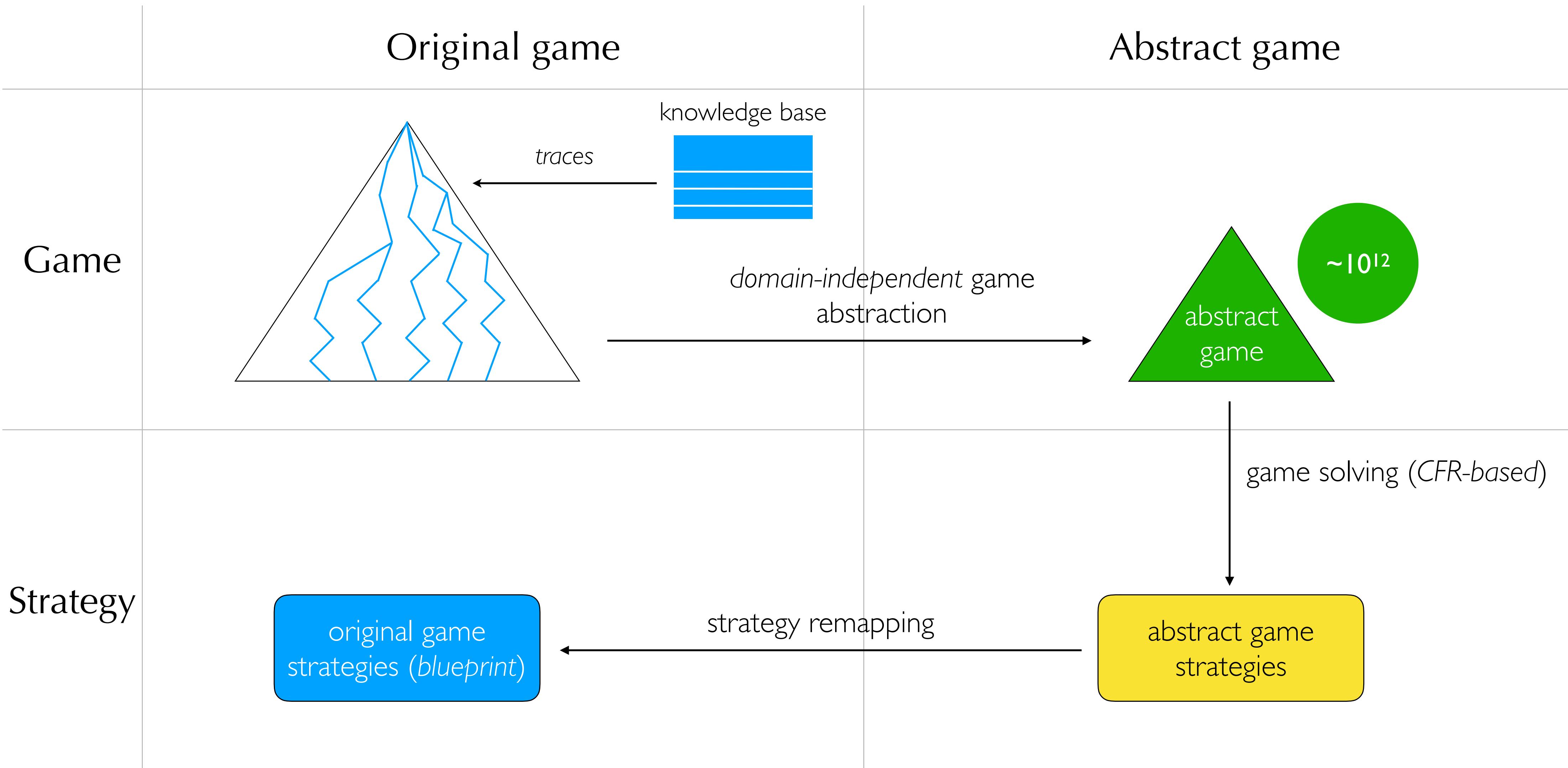
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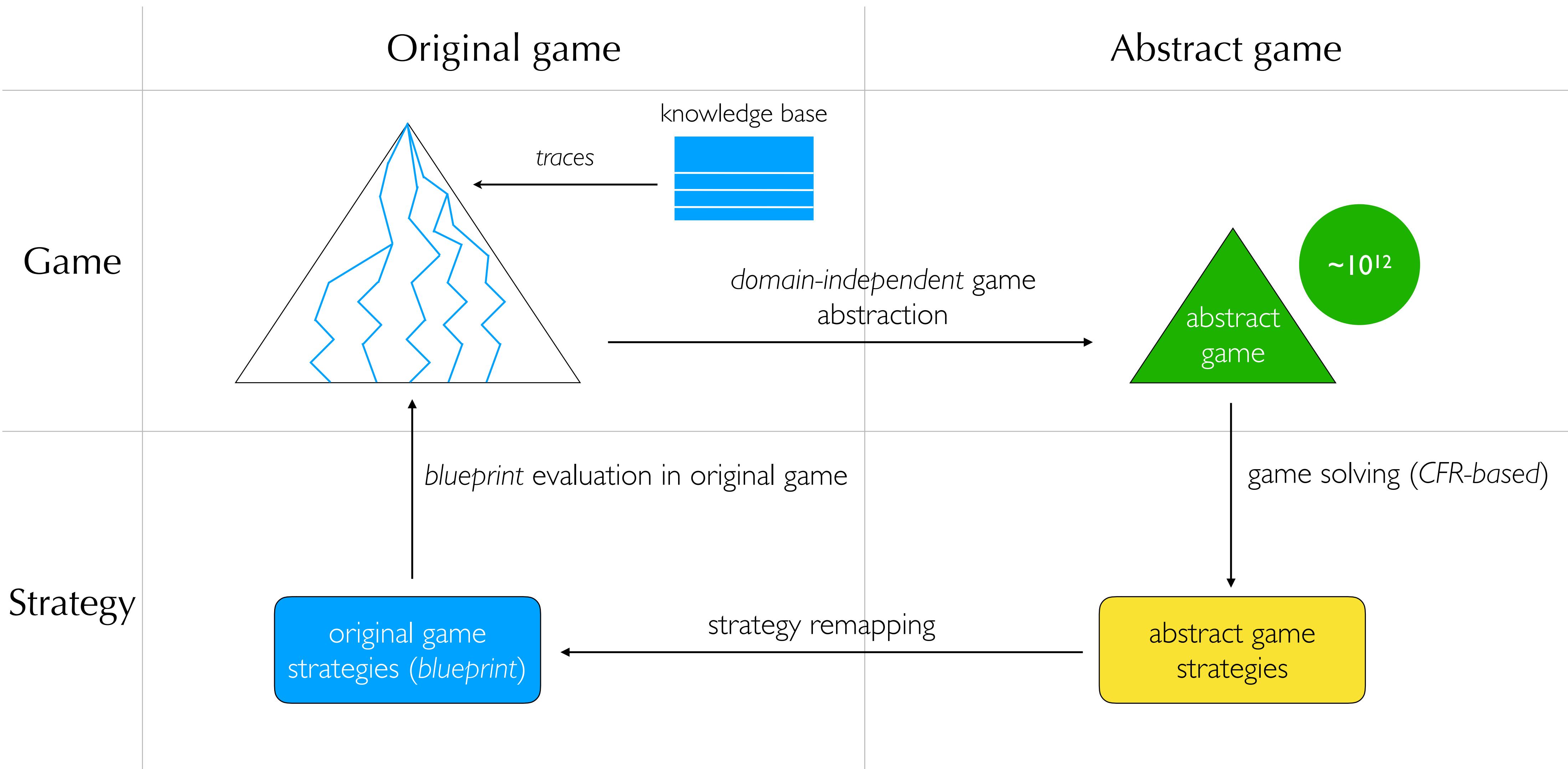
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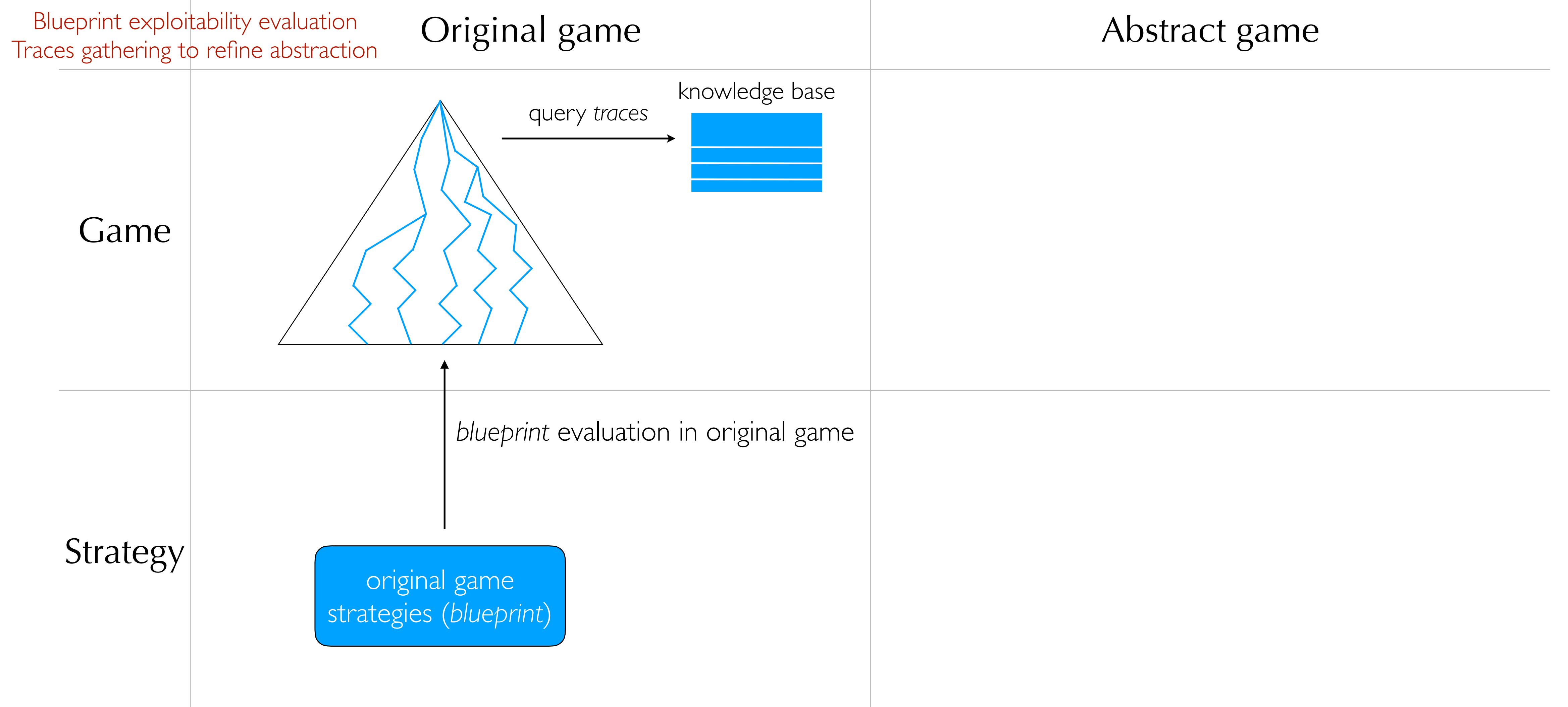
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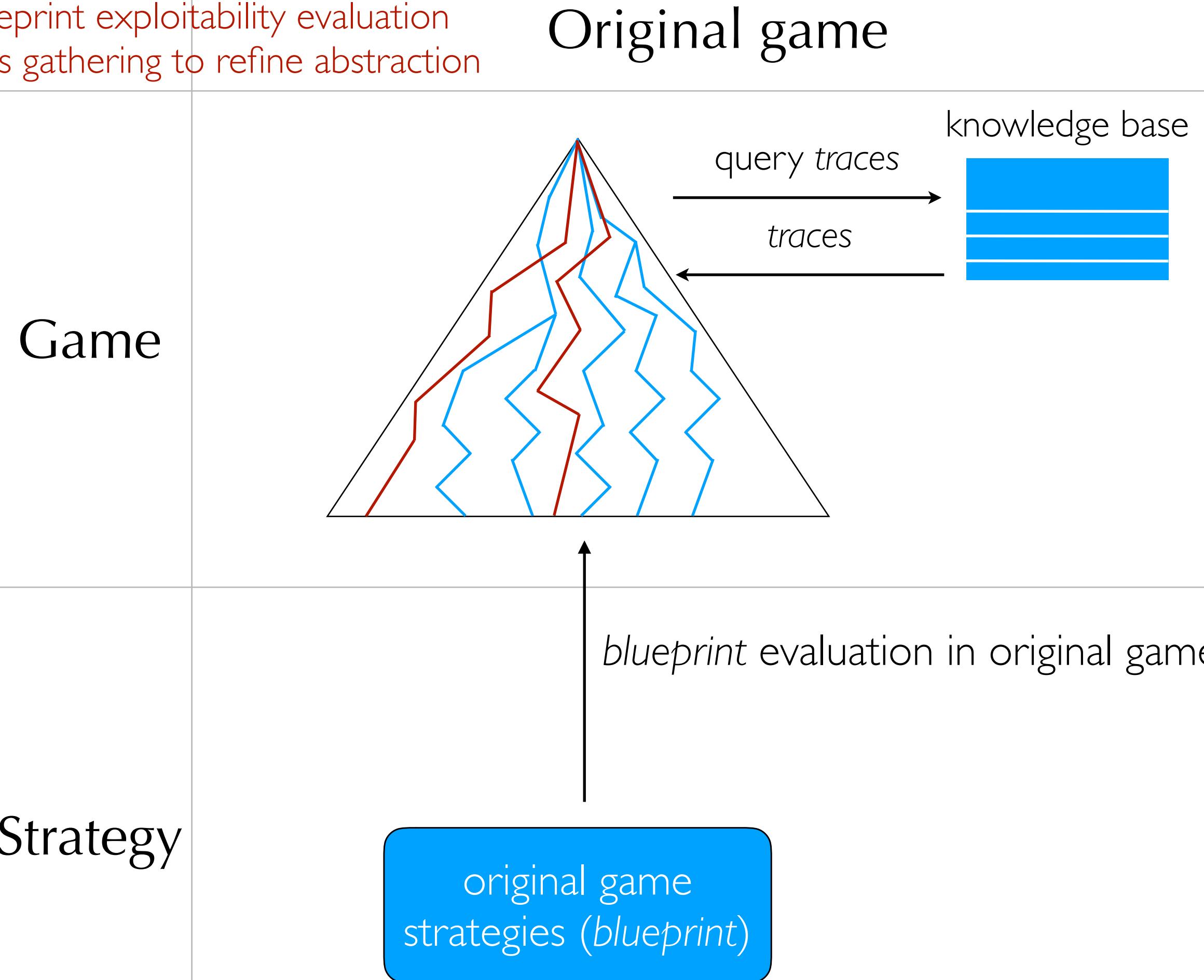


Idea



Idea

Blueprint exploitability evaluation
Traces gathering to refine abstraction



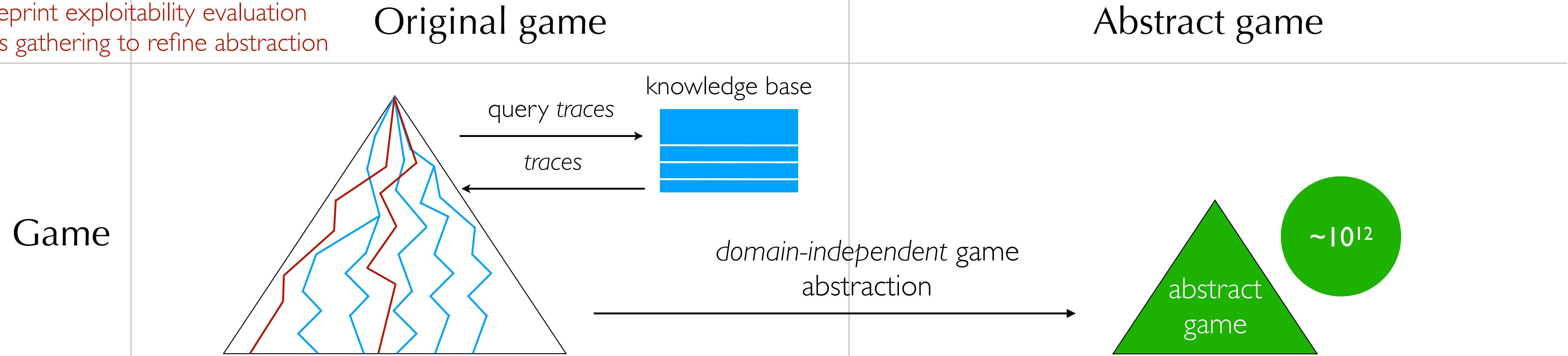
Abstract game

Game

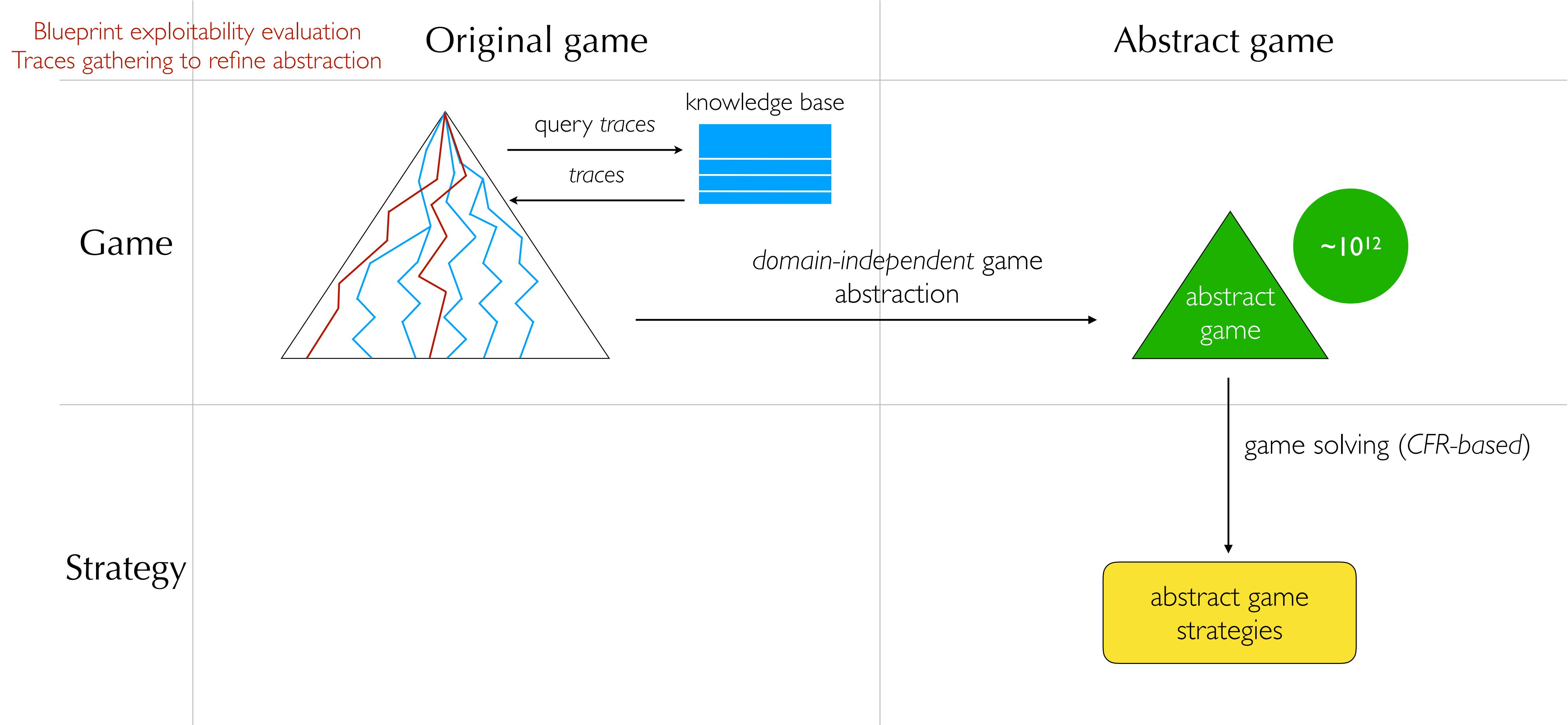
Strategy

Idea

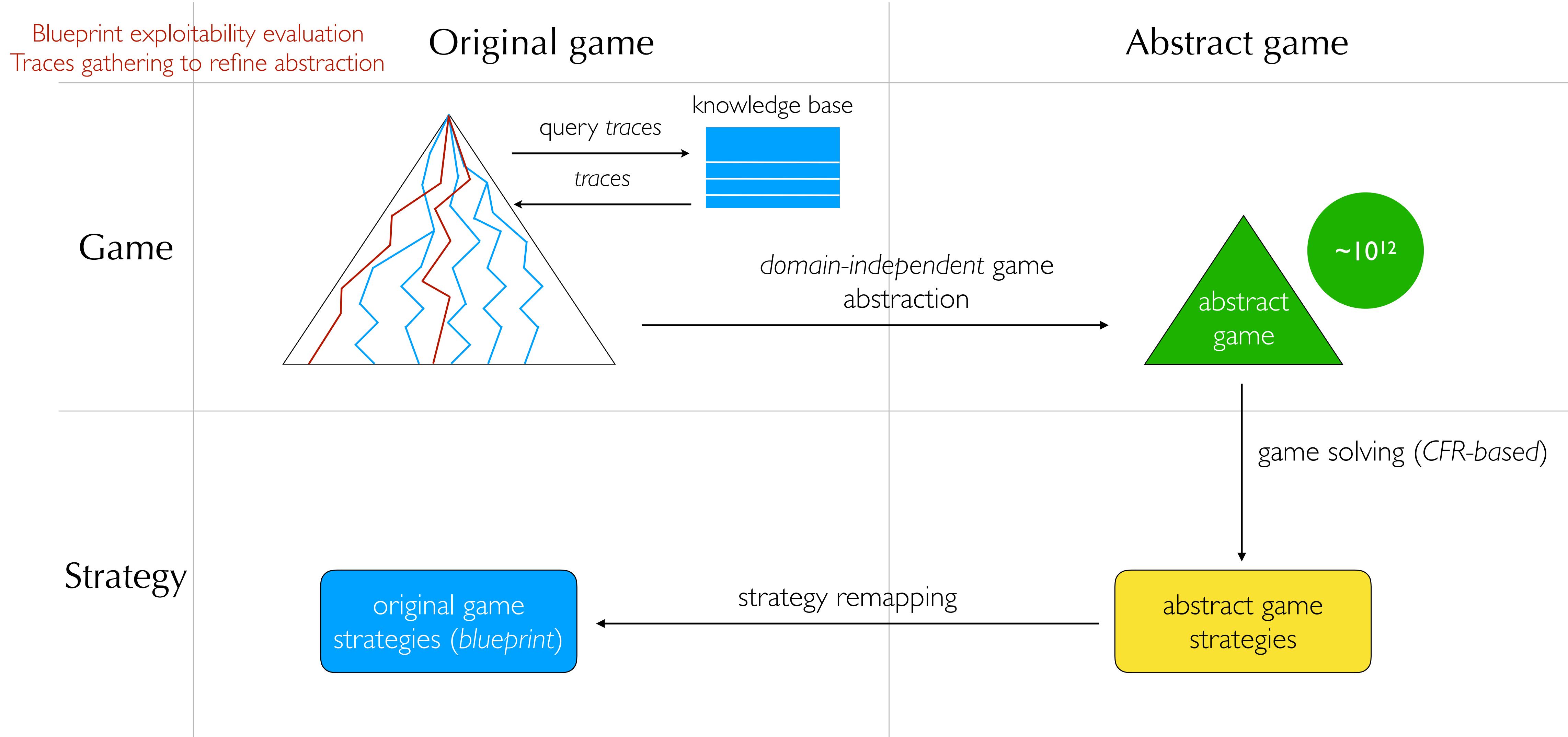
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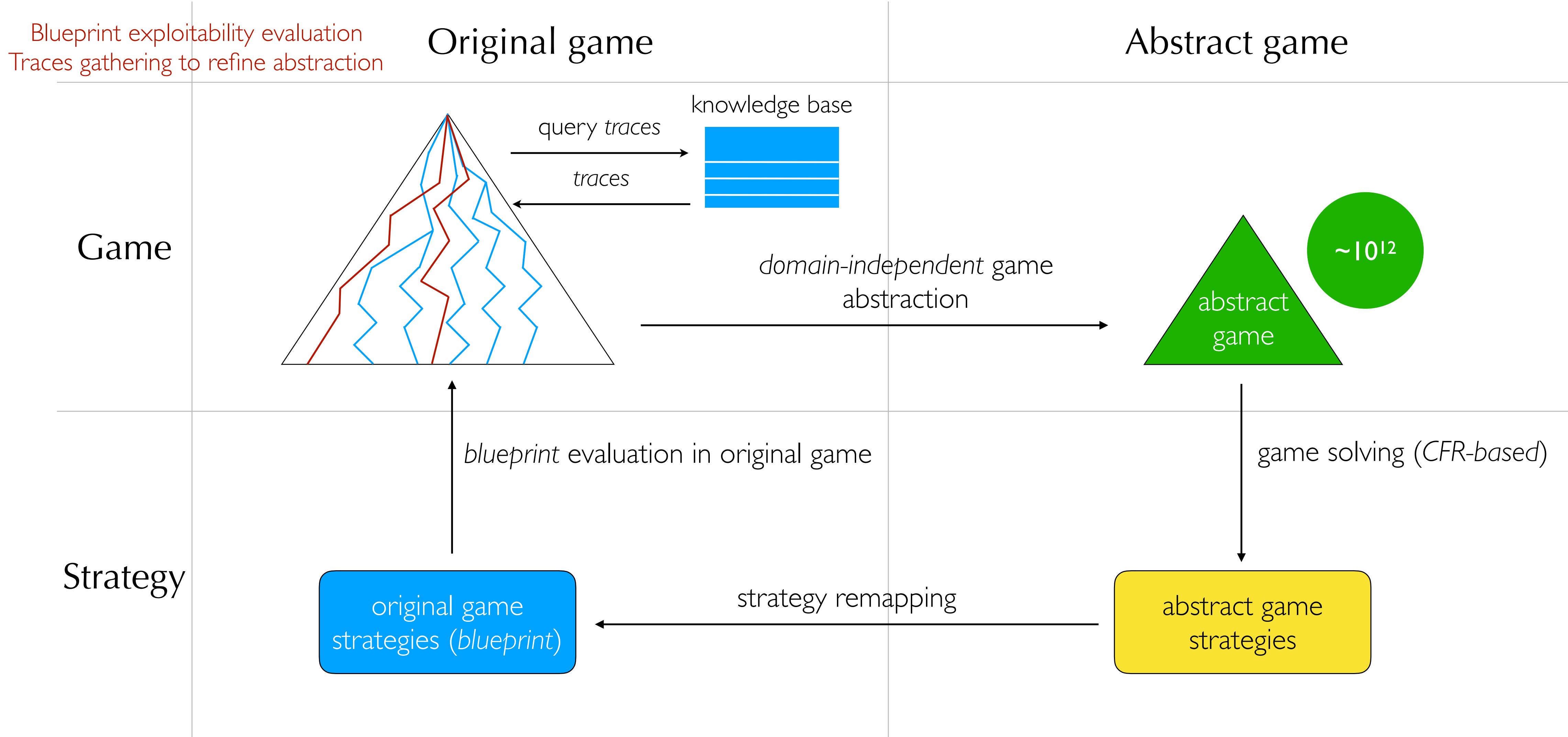
Idea



Idea



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

