

More Realistic Adversarial Settings

Virginia Tech CS4804
Introduction to Artificial Intelligence

Outline

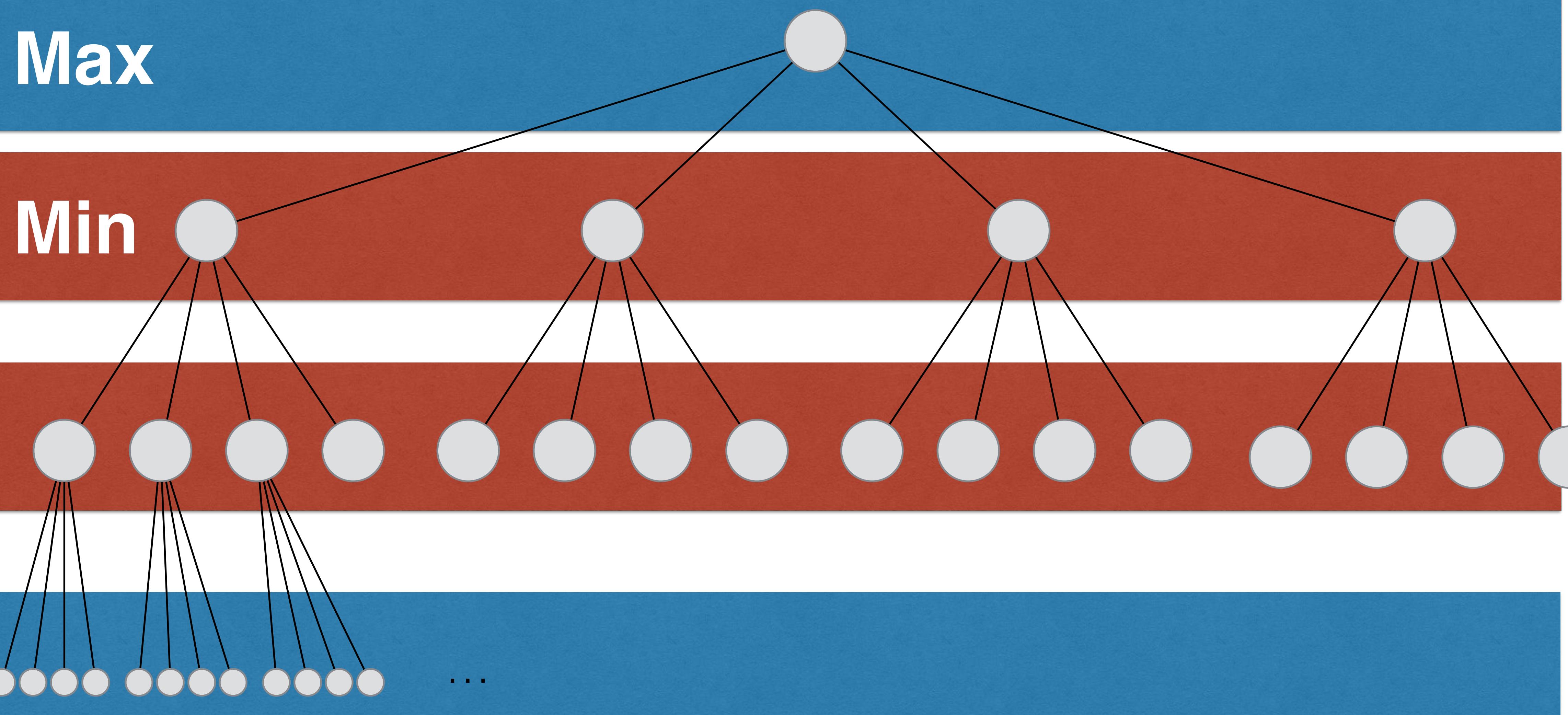
- Move ordering
- Stochastic games
- (Partially-observable games)



Max



Min



```
MINIMAX(s) =  
    if TERMINAL-TEST(s) then UTILITY(s)  
    if PLAYER(s) = MAX then  
        max of MINIMAX(RESULT(s,a)) for a in ACTIONS(s)  
    if PLAYER(s) = MIN then  
        min of MINIMAX(RESULT(s,a)) for a in ACTIONS(s)
```

function MINIMAX-DECISION(*state*) **returns** an action
 return $\arg \max_{a \in \text{ACTIONS}(s)} \text{MIN-VALUE}(\text{RESULT}(s, a))$

function MAX-VALUE(*state*) **returns** a utility value
 if TERMINAL-TEST(*state*) **then return** UTILITY(*state*)
 $v \leftarrow -\infty$
 for each *a* **in** ACTIONS(*state*) **do**
 $v \leftarrow \text{MAX}(v, \text{MIN-VALUE}(\text{RESULT}(s, a)))$
 return *v*

function MIN-VALUE(*state*) **returns** a utility value
 if TERMINAL-TEST(*state*) **then return** UTILITY(*state*)
 $v \leftarrow \infty$
 for each *a* **in** ACTIONS(*state*) **do**
 $v \leftarrow \text{MIN}(v, \text{MAX-VALUE}(\text{RESULT}(s, a)))$
 return *v*

function ALPHA-BETA-SEARCH(*state*) **returns** an action
 $v \leftarrow \text{MAX-VALUE}(state, -\infty, +\infty)$
 return the *action* in ACTIONS(*state*) with value v

function MAX-VALUE(*state*, α , β) **returns** a utility value
 if TERMINAL-TEST(*state*) **then return** UTILITY(*state*)
 $v \leftarrow -\infty$
 for each *a* **in** ACTIONS(*state*) **do**
 $v \leftarrow \text{MAX}(v, \text{MIN-VALUE}(\text{RESULT}(s, a), \alpha, \beta))$
 if $v \geq \beta$ **then return** v
 $\alpha \leftarrow \text{MAX}(\alpha, v)$
 return v

function MIN-VALUE(*state*, α , β) **returns** a utility value
 if TERMINAL-TEST(*state*) **then return** UTILITY(*state*)
 $v \leftarrow +\infty$
 for each *a* **in** ACTIONS(*state*) **do**
 $v \leftarrow \text{MIN}(v, \text{MAX-VALUE}(\text{RESULT}(s, a), \alpha, \beta))$
 if $v \leq \alpha$ **then return** v
 $\beta \leftarrow \text{MIN}(\beta, v)$
 return v

Game theorists crack poker

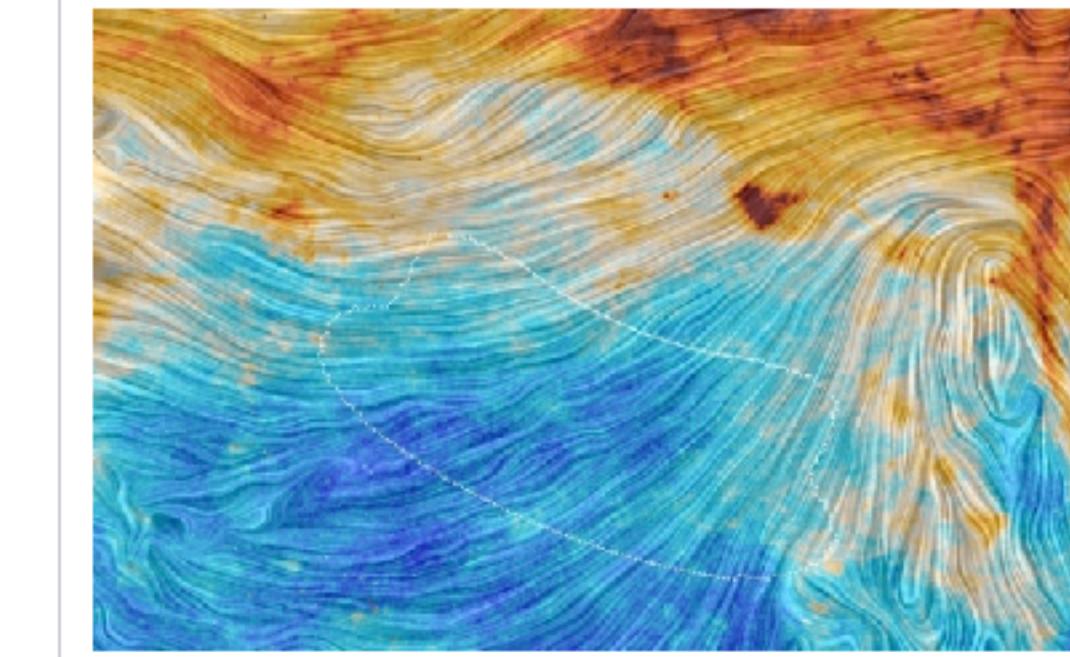
An 'essentially unbeatable' algorithm for the popular card game points to strategies for solving real-life problems without having complete information.

Philip Ball

08 January 2015

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Top story



Gravitational waves discovery officially dead

Galactic dust confirmed as the source of the signal that researchers thought was evidence for gravitational waves from the early Universe.

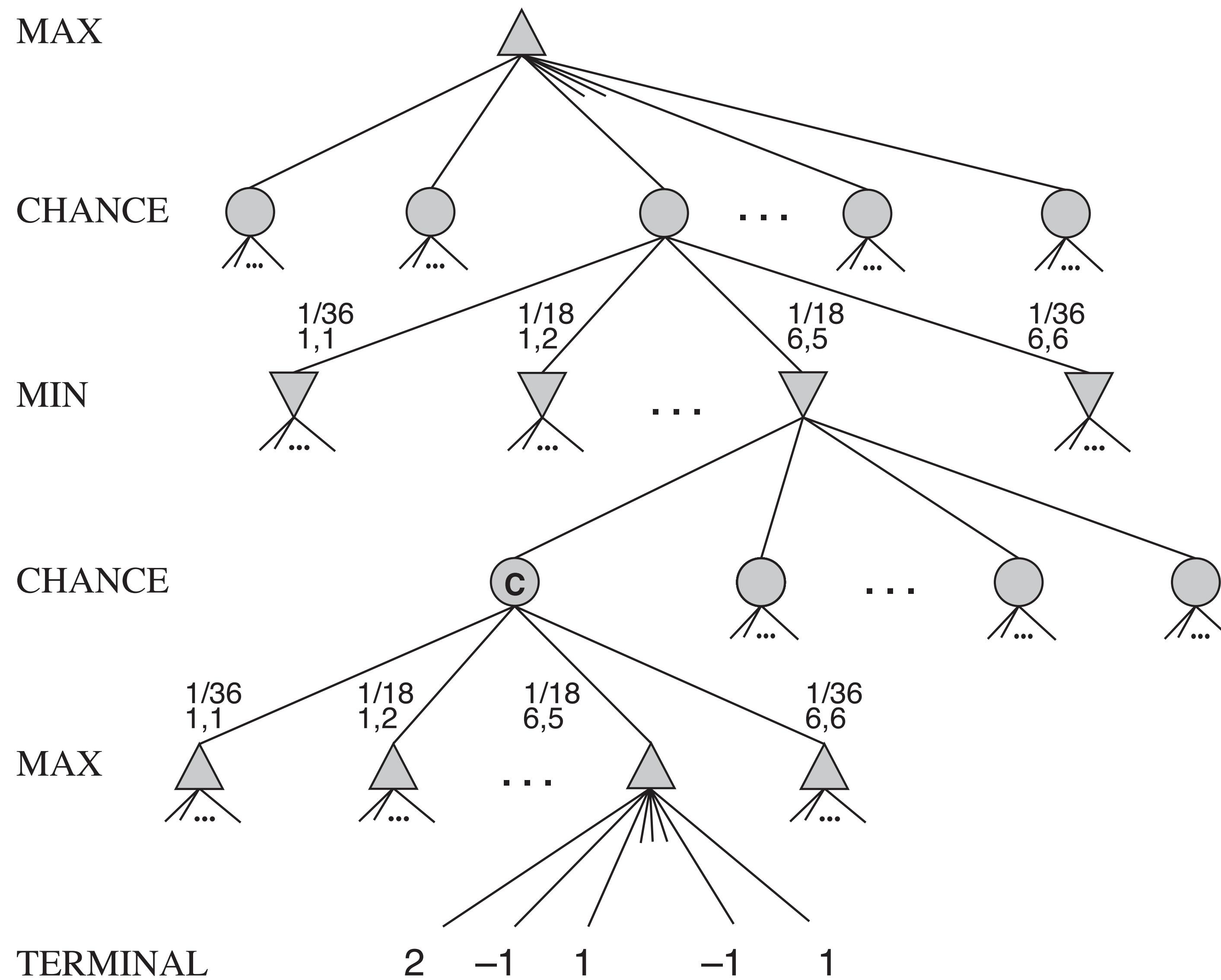
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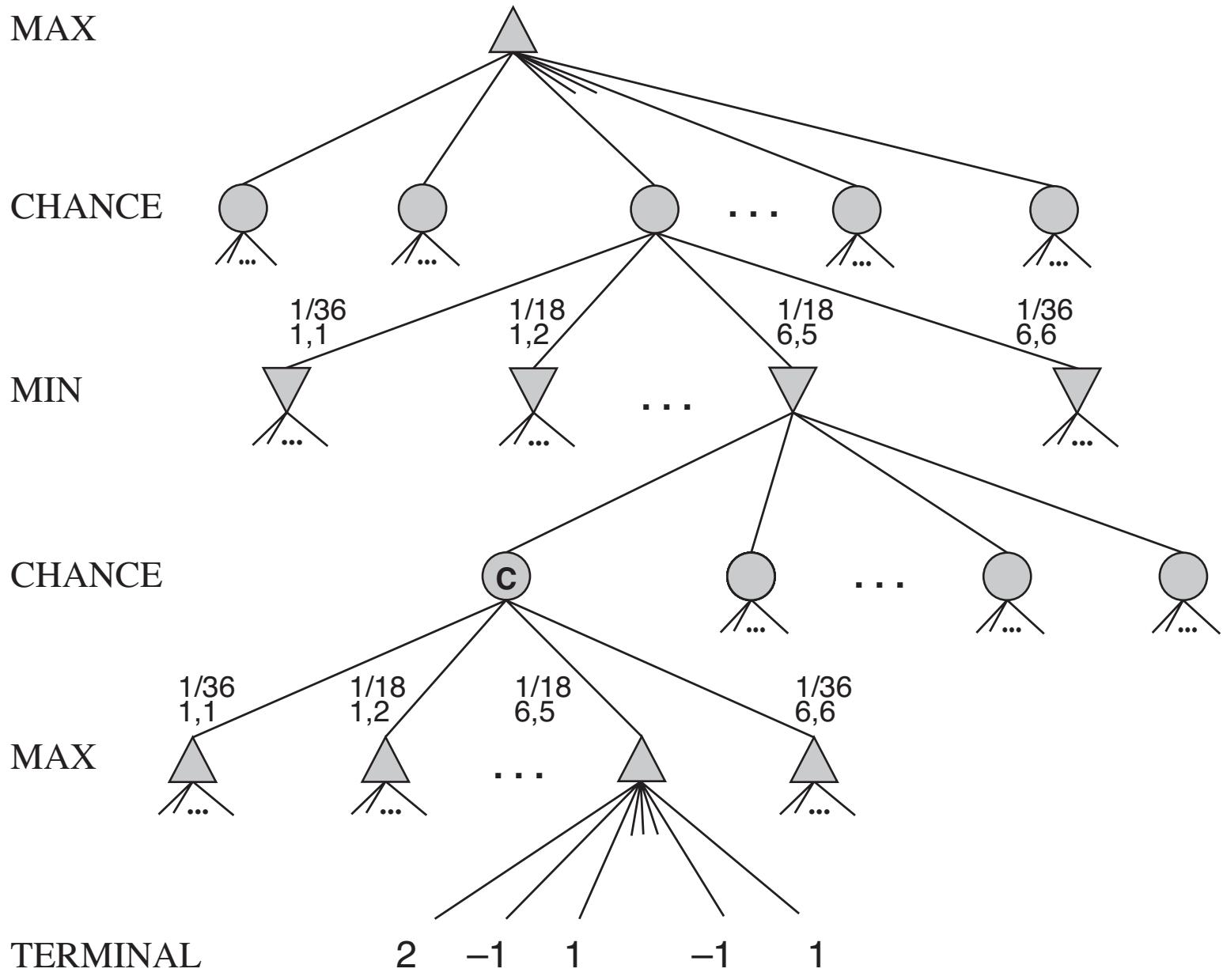
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Expectiminimax

- New player type: Chance

$$\text{EXPECTINIMIMAX}(s) = \begin{cases} \text{UTILITY}(s) & \text{if } \text{TERMINAL-TEST}(s) \\ \max_a \text{EXPECTINIMIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MAX} \\ \min_a \text{EXPECTINIMIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MIN} \\ \sum_r \Pr(r) \text{EXPECTINIMIMAX}(\text{RESULT}(s, r)) & \text{if } \text{PLAYER}(s) = \text{CHANCE} \end{cases}$$





$\text{EXPECTINIMIMAX}(s) =$

$$\begin{cases} \text{UTILITY}(s) & \text{if } \text{TERMINAL-TEST}(s) \\ \max_a \text{EXPECTINIMIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MAX} \\ \min_a \text{EXPECTINIMIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MIN} \\ \sum_r \Pr(r) \text{EXPECTINIMIMAX}(\text{RESULT}(s, r)) & \text{if } \text{PLAYER}(s) = \text{CHANCE} \end{cases}$$

Pruning Chance Nodes

- Bounds on true utility function -> bounds on expectation
- $-10 \leq \text{Utility} \leq 10$
- uniform probability die:

?	?	?	?	?	?
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Pruning Chance Nodes

- Bounds on true utility function -> bounds on expectation
- $-10 \leq \text{Utility} \leq 10$
- uniform probability die:

10	?	?	?	?	?
----	---	---	---	---	---

Best case

10	10	10	10	10	10
----	----	----	----	----	----

expectation 10

Worst case

10	-10	-10	-10	-10	-10
----	-----	-----	-----	-----	-----

expectation -6.67

Pruning Chance Nodes

- Bounds on true utility function -> bounds on expectation
- $-10 \leq \text{Utility} \leq 10$
- uniform probability die:

10	10	?	?	?	?
----	----	---	---	---	---

Best case

10	10	10	10	10	10
----	----	----	----	----	----

expectation 10

Worst case

10	10	-10	-10	-10	-10
----	----	-----	-----	-----	-----

expectation -3.33

Pruning Chance Nodes

- Bounds on true utility function -> bounds on expectation
- $-10 \leq \text{Utility} \leq 10$
- uniform probability die:

10	10	-10	?	?	?
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Best case

10	10	-10	10	10	10
----	----	-----	----	----	----

expectation 6.67

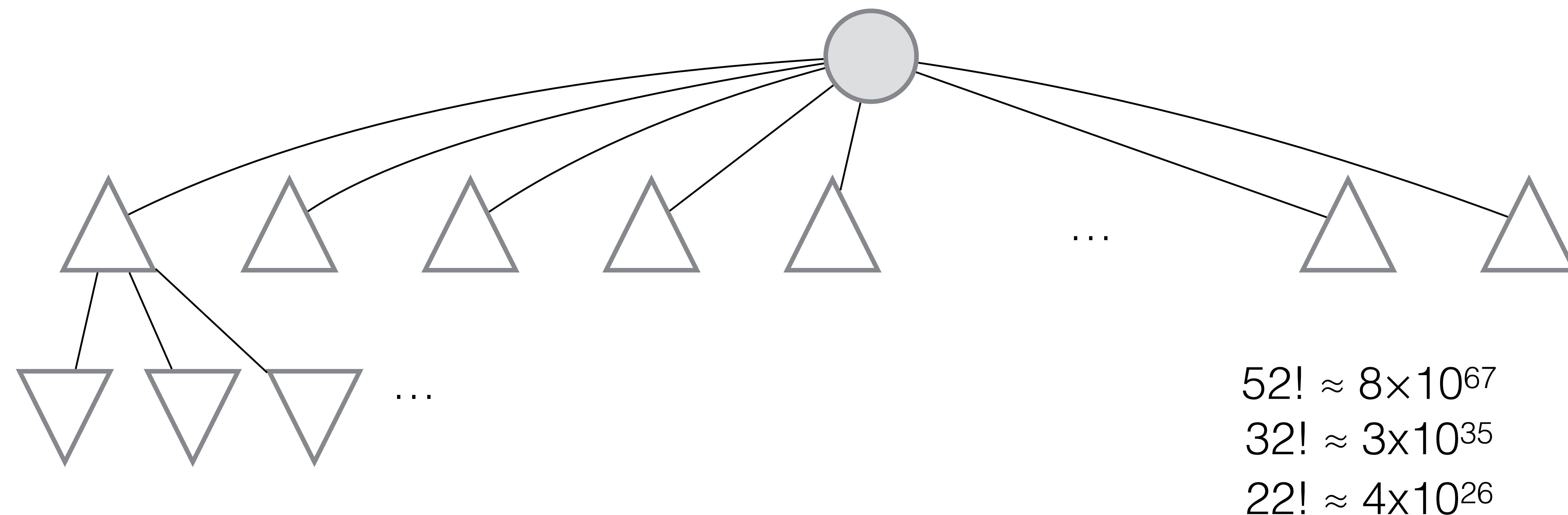
Worst case

10	10	-10	-10	-10	-10
----	----	-----	-----	-----	-----

expectation -3.33

Partial Observations

- One approach: simulate perfect information with Chance nodes



Summary

- Minimax logic works for any move ordering
- Expectiminimax adds Chance “player” and uses expected value
- Strategy for handling partial information