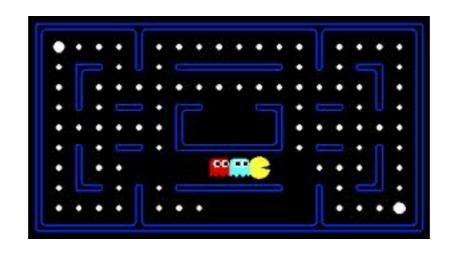


# Games: game evaluation



### **Policies**

Deterministic policies:  $\pi_p(s) \in \mathsf{Actions}(s)$ 

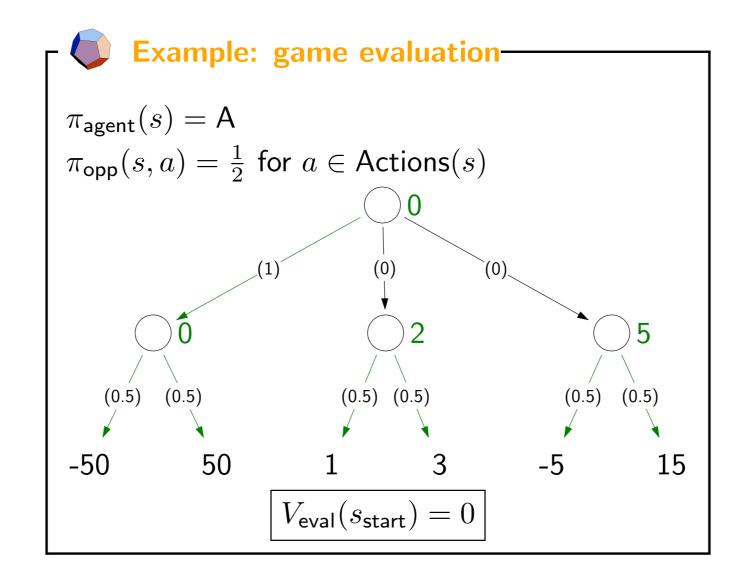
action that player p takes in state s

Stochastic policies  $\pi_p(s, a) \in [0, 1]$ :

probability of player p taking action a in state s

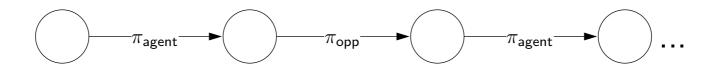
[semi-live solution: humanPolicy]

## Game evaluation example



### Game evaluation recurrence

Analogy: recurrence for policy evaluation in MDPs

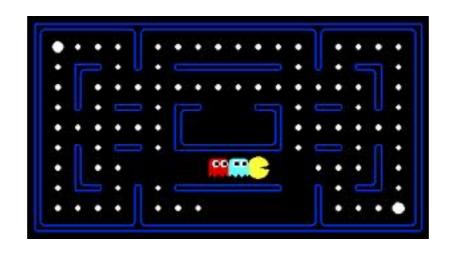


#### Value of the game:

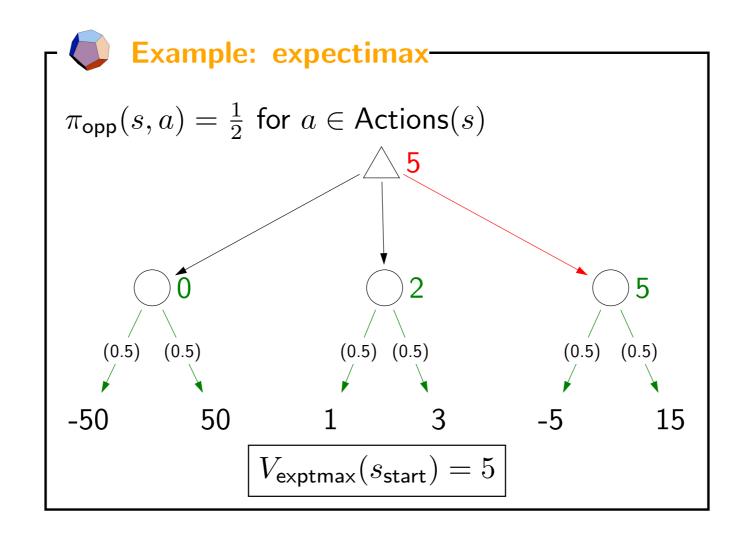
$$V_{\mathsf{eval}}(s) = \begin{cases} \mathsf{Utility}(s) & \mathsf{IsEnd}(s) \\ \sum_{a \in \mathsf{Actions}(s)} \pi_{\mathsf{agent}}(s, a) V_{\mathsf{eval}}(\mathsf{Succ}(s, a)) & \mathsf{Player}(s) = \mathsf{agent} \\ \sum_{a \in \mathsf{Actions}(s)} \pi_{\mathsf{opp}}(s, a) V_{\mathsf{eval}}(\mathsf{Succ}(s, a)) & \mathsf{Player}(s) = \mathsf{opp} \end{cases}$$



# Games: expectimax

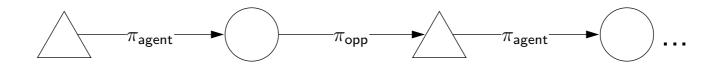


### Expectimax example



## Expectimax recurrence

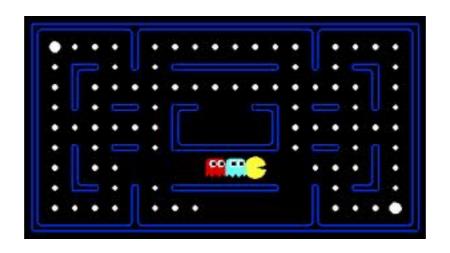
Analogy: recurrence for value iteration in MDPs



$$V_{\mathsf{exptmax}}(s) = \begin{cases} \mathsf{Utility}(s) & \mathsf{IsEnd}(s) \\ \max_{a \in \mathsf{Actions}(s)} V_{\mathsf{exptmax}}(\mathsf{Succ}(s, a)) & \mathsf{Player}(s) = \mathsf{agent} \\ \sum_{a \in \mathsf{Actions}(s)} \pi_{\mathsf{opp}}(s, a) V_{\mathsf{exptmax}}(\mathsf{Succ}(s, a)) & \mathsf{Player}(s) = \mathsf{opp} \end{cases}$$



## Games: minimax

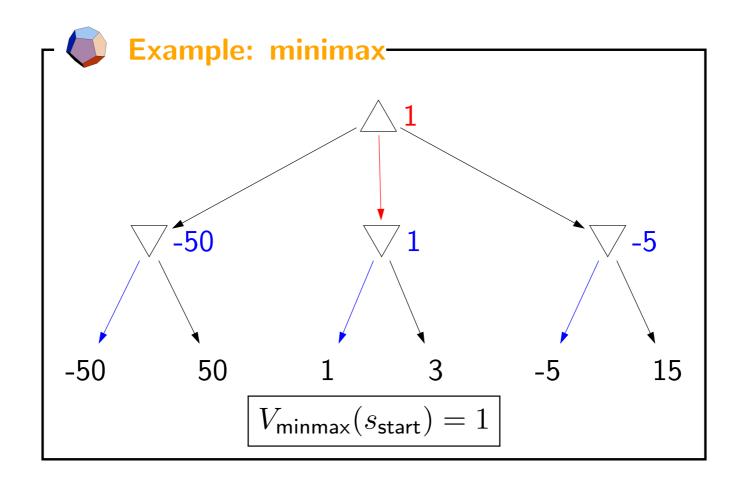


Problem: don't know opponent's policy

Approach: assume the worst case

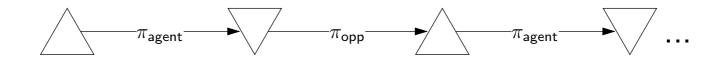


# Minimax example



#### Minimax recurrence

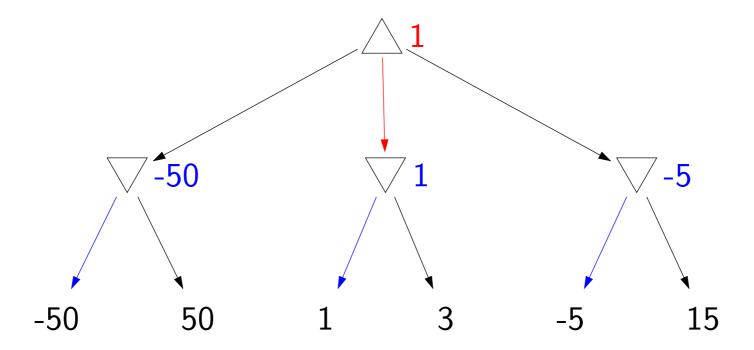
#### No analogy in MDPs:



$$V_{\mathsf{minmax}}(s) = \begin{cases} \mathsf{Utility}(s) & \mathsf{IsEnd}(s) \\ \max_{a \in \mathsf{Actions}(s)} V_{\mathsf{minmax}}(\mathsf{Succ}(s, a)) & \mathsf{Player}(s) = \mathsf{agent} \\ \min_{a \in \mathsf{Actions}(s)} V_{\mathsf{minmax}}(\mathsf{Succ}(s, a)) & \mathsf{Player}(s) = \mathsf{opp} \end{cases}$$

## Extracting minimax policies

$$\begin{split} \pi_{\max}(s) &= \arg\max_{a \in \mathsf{Actions}(s)} V_{\mathsf{minmax}}(\mathsf{Succ}(s, a)) \\ \pi_{\min}(s) &= \arg\min_{a \in \mathsf{Actions}(s)} V_{\mathsf{minmax}}(\mathsf{Succ}(s, a)) \end{split}$$



## The halving game



#### Problem: halving game-

Start with a number N.

Players take turns either decrementing N or replacing it with  $\lfloor \frac{N}{2} \rfloor$ .

The player that is left with 0 wins.

[semi-live solution: minimaxPolicy]

#### Face off

#### Recurrences produces policies:

```
V_{\text{exptmax}} \Rightarrow \pi_{\text{exptmax}(7)}, \pi_7 \text{ (some opponent)}
V_{\text{minmax}} \Rightarrow \pi_{\text{max}}, \pi_{\text{min}}
```

#### Play policies against each other:

```
\pi_{\min} \pi_{7} V(\pi_{\max}, \pi_{\min}) V(\pi_{\max}, \pi_{7}) V(\pi_{\max}, \pi_{7}) V(\pi_{\exp t \max(7)}, \pi_{\min}) V(\pi_{\exp t \max(7)}, \pi_{7})
```

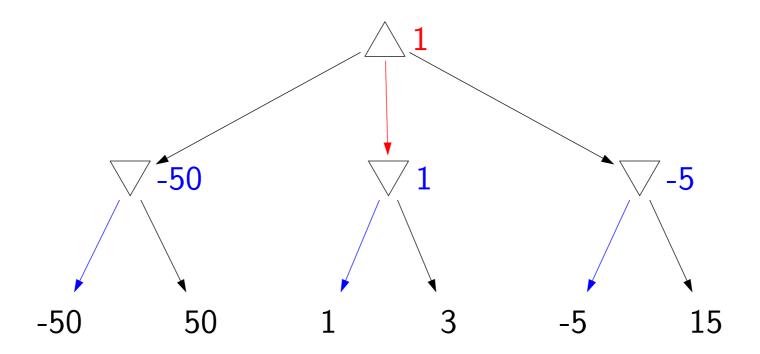
What's the relationship between these values?

## Minimax property 1



#### Proposition: best against minimax opponent-

$$V(\pi_{\max}, \pi_{\min}) \geq V(\pi_{\text{agent}}, \pi_{\min})$$
 for all  $\pi_{\text{agent}}$ 

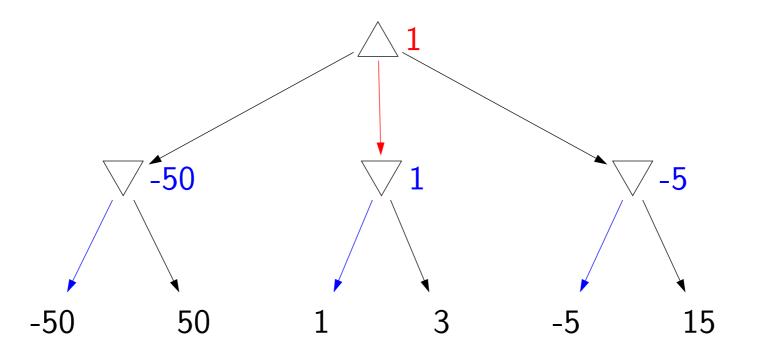


## Minimax property 2



#### Proposition: lower bound against any opponent-

$$V(\pi_{\max}, \pi_{\min}) \leq V(\pi_{\max}, \pi_{\mathsf{opp}})$$
 for all  $\pi_{\mathsf{opp}}$ 

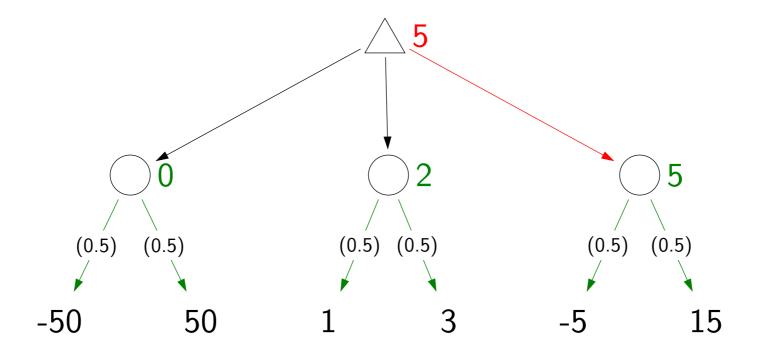


## Minimax property 3



#### Proposition: not optimal if opponent is known-

 $V(\pi_{\max}, \pi_7) \leq V(\pi_{\exp t\max(7)}, \pi_7)$  for opponent  $\pi_7$ 



### Relationship between game values

