

Design of AI for games

Students Survival Pack

IUT de Sénart Fontainebleau
Département Informatique



Different types of games

Perfect Information

Deterministic

checkers, othello, chess, go,
diplomacy

With Chance

backgammon, monopoly, risk

Imperfect Information

Deterministic

battleships, stratego,
diplomacy

With Chance

bridge, poker, scrabble, risk

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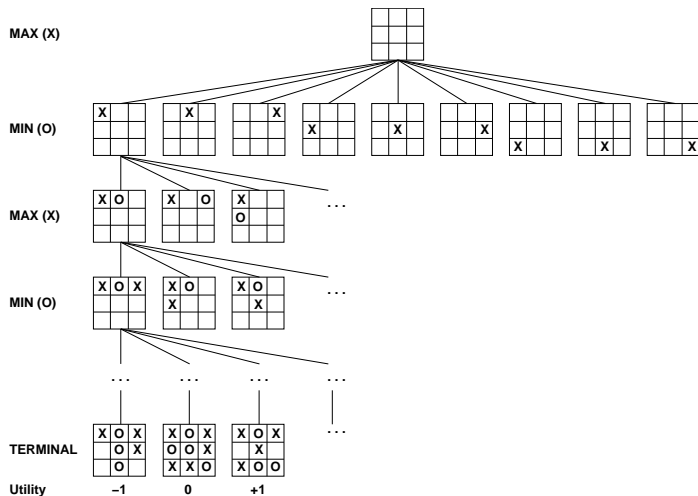
Game tree

- ▶ Nodes: game configuration
- ▶ Successors: configuration reachable in one ply
- ▶ Leafs: are endgame position labelled by a **payoff function**, e.g. Loss= -1 , Draw= 0 and Win= $+1$.

Remark

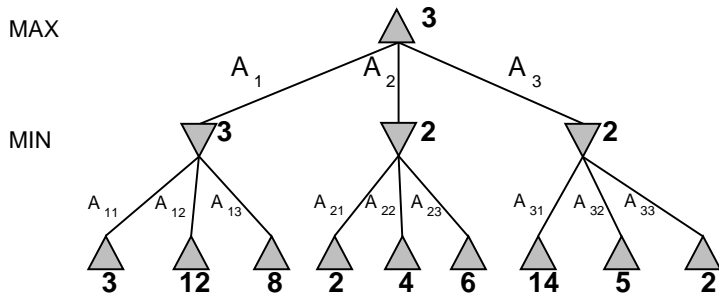
The game tree of a minimax game is finite.

Example: Tic-Tac-Toe



Minimax

- ▶ **Perfect play** for deterministic, perfect-information games.
- ▶ **strategy**: choose move to position with highest **minimax value**
- ▶ It is the **best achievable payoff against best play**.



The algorithm

Minimax(currentState)

```
if currentState is a terminal state then
  return payoff(currentState)
else if I am to move then
  return  $\max_{\text{nextState}} \{\mathbf{Minimax}(\text{nextState})\}$ 
else
  return  $\min_{\text{nextState}} \{\mathbf{Minimax}(\text{nextState})\}$ 
end if
```

With one function for each player

ExploreMax(currentState)

```
if currentState is a terminal state then
  return payoff(currentState)
else
  return  $\max_{\text{nextState}} \{\mathbf{ExploreMin}(\text{nextState})\}$ 
end if
```


With one function for each player

ExploreMin(currentState)

```
if currentState is a terminal state then
  return payoff(currentState)
else
  return minnextState {ExploreMax(nextState)}
end if
```

With one function for each player

First call

ExploreMax(currentState)

ExploreMax

ExploreMin

Remark

- ▶ the two functions are calling each other recursively until a leaf of the game tree is reached
- ▶ the values are propagated upward

Example

- ▶ Try some applets available on the web.
- ▶ For example with `minimax` on a complete tree with nodes of degree 3 and height 2 (beware the variables are called differently in the applet and in the rest of this lecture: their b denotes the degree and their d the depth).
- ▶ You can also enter evaluation for the nodes, try for example

1, 0, 1, 1, -1, 0, 0, -1, 0.

- ▶ Note that, there is no pruning mechanism implemented (e.g. when reaching Win).
- ▶ If you use the algorithm `Minimax alpha beta`, some pruning occurs.

With one function for each player

First call

ExploreMax(currentState)

ExploreMax

ExploreMin

Remark

- ▶ the two functions are calling each other recursively until a leaf of the game tree is reached
- ▶ the values are propagated upward

Further remarks

Warning

- ▶ the tree is never fully stored on the computer
- ▶ some parts of it are stored in RAM
- ▶ Essentially by the call stack of the two functions.

Implementation issues

- ▶ the max and min in these functions are implemented by loops of recursive calls
- ▶ to avoid pointless computing, if we find a maximal value in a max loop (or a minimal value in a min loop) we can stop the loop early.
- ▶ in an **object oriented framework**, the passing of state as arguments will probably not be needed

Cutting off search

Implementation Issues

- ▶ With two functions **ExploreMax** and **ExploreMin**, note that both will need a cut-off test
- ▶ If a cut-off is added, we also need to implement an **evaluation function f** .
- ▶ In a two player game, it is recommended to keep this function f symmetric.
- ▶ In this case, the simplest evaluation function is $+\infty$ for a win, $-\infty$ for a loss and 0 otherwise (essentially every game that has not been simulated until the end is treated as a draw by the AI in its analysis).