

Adversarial Search

CSE 473
University of Washington

Contents

- Board Games
- Minimax Search
- Alpha-Beta Search
- Games with an Element of Chance

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Games Overview

	deterministic	chance
Perfect information	chess, checkers, go, othello	backgammon, monopoly
Imperfect information		bridge, poker, scrabble

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Games & Game Theory

- When there is *more than one agent*, the future is not anymore easily predictable for the agent
- In *competitive* environments (conflicting goals), adversarial search becomes necessary
- In AI, we usually consider special type of games:

board games, which can be characterized as *deterministic, turn-taking, two-player, zero-sum* games with *perfect information*

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Games as Search

- **Components:**
 - **States:**
 - **Initial state:**
 - **Successor function:**
 - **Terminal test:**
 - **Utility function:**

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Games as Search

- **Components:**
 - **States:** board configurations
 - **Initial state:** the board position and which player will move
 - **Successor function:** returns list of *(move, state)* pairs, each indicating a legal move and the resulting state
 - **Terminal test:** determines when the game is over
 - **Utility function:** gives a numeric value in terminal states (eg, -1, 0, +1 in chess for loss, tie, win)

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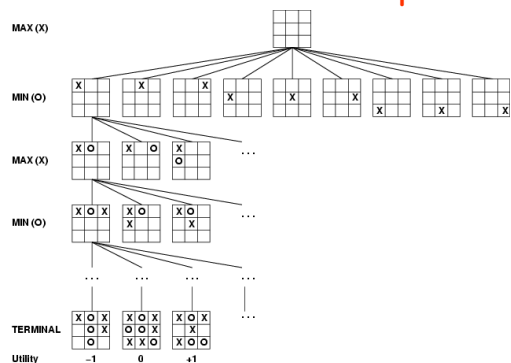
Games as Search

- **Components:**
 - **States:** board configurations
 - **Initial state:** the board position and which player will move
 - **Successor function:** returns list of *(move, state)* pairs, each indicating a legal move and the resulting state
 - **Terminal test:** determines when the game is over
 - **Utility function:** gives a numeric value in terminal states (eg, -1, 0, +1 in chess for loss, tie, win)
- **Convention:** first player is MAX, 2nd player is MIN
- State utility values from MAX's perspective
- Initial state and legal moves define the game tree

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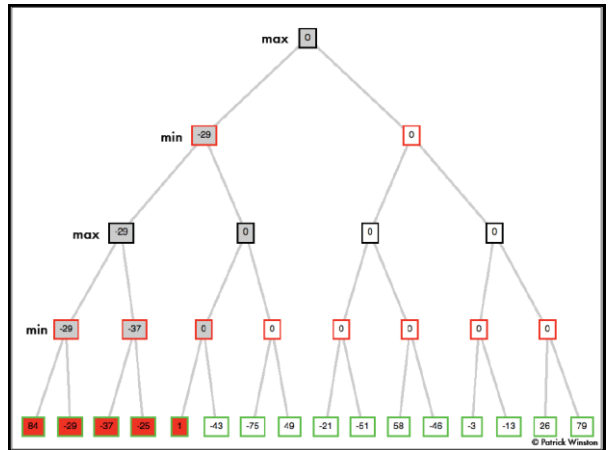
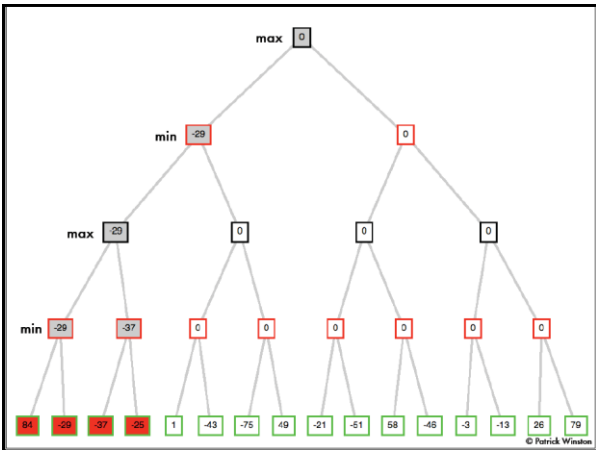
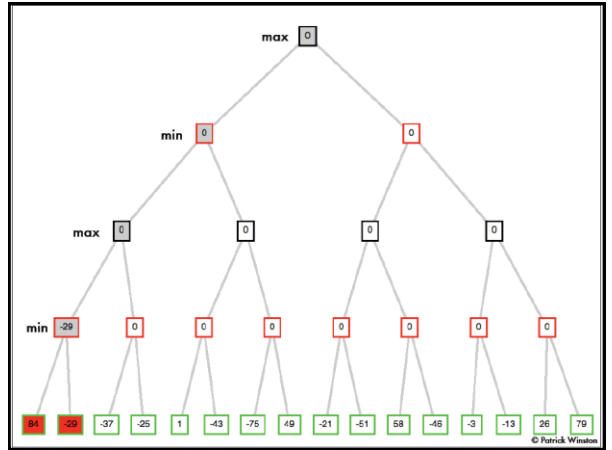
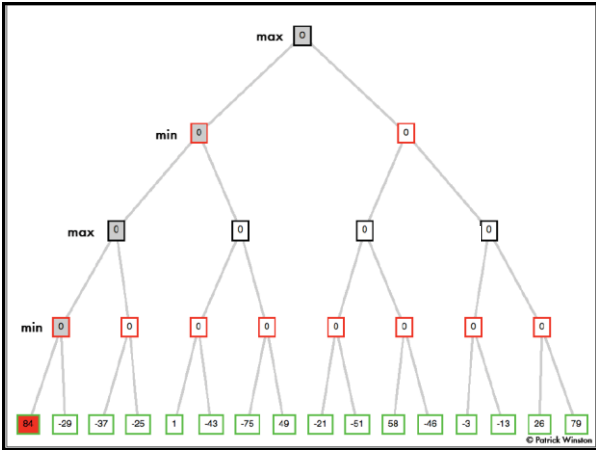
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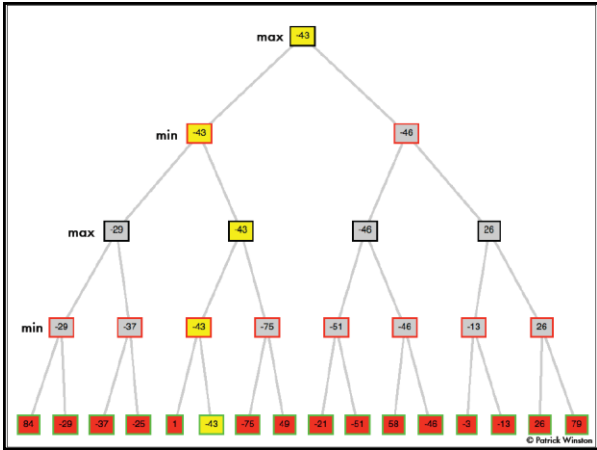
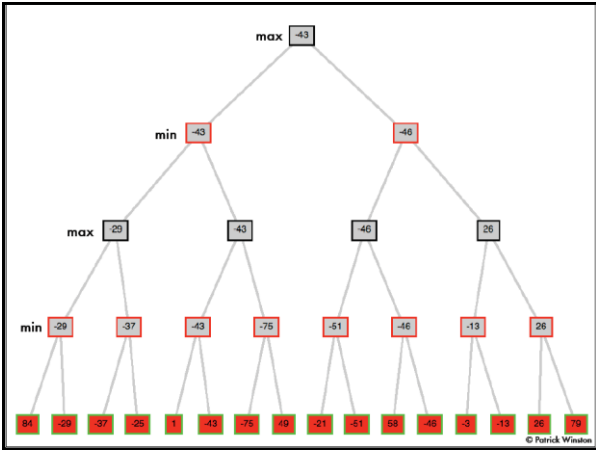
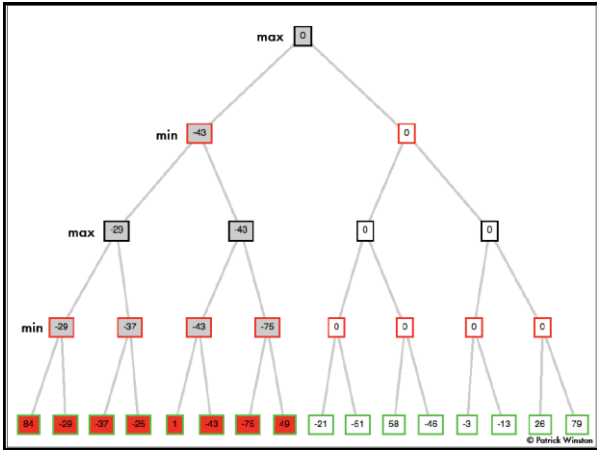
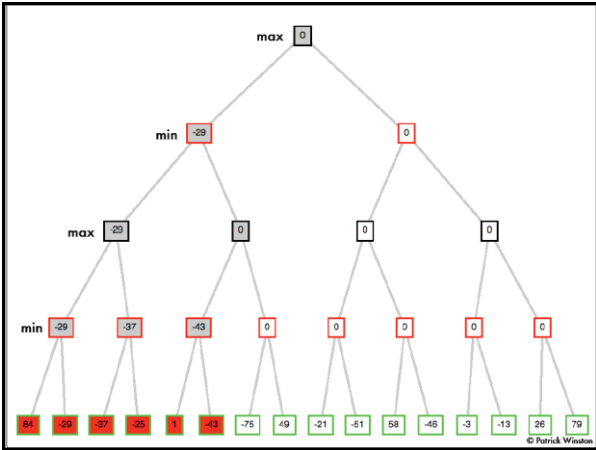
Tic-Tac-Toe Example



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Properties of minimax

- Complete?
- Optimal?
- Time complexity?
- Space complexity?

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Properties of minimax

- Complete? Yes (if tree is finite)
- Optimal? Yes (against an optimal opponent)
- Time complexity? $O(b^m)$
- Space complexity? $O(bm)$ (depth-first exploration)

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Good enough?

- Chess:
 - branching factor $b \approx 35$
 - game length $m \approx 100$
 - search space $b^m \approx 35^{100} \approx 10^{154}$
- The Universe:
 - number of atoms $\approx 10^{78}$
 - age $\approx 10^{21}$ milliseconds

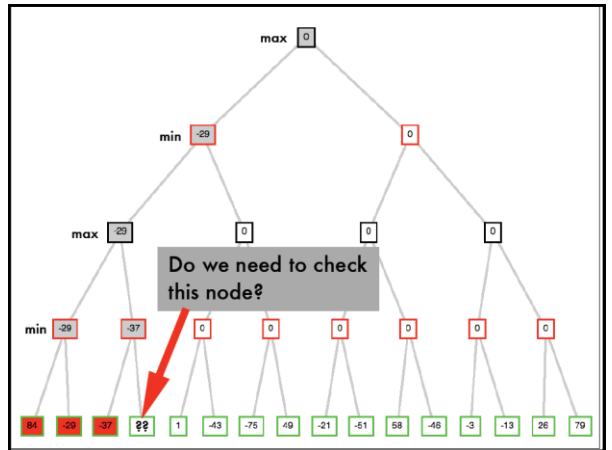
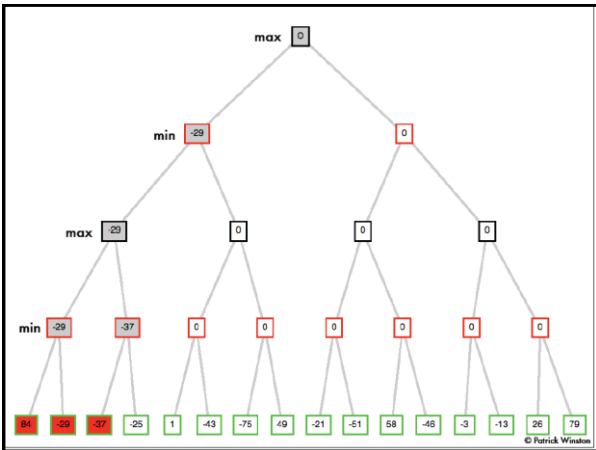
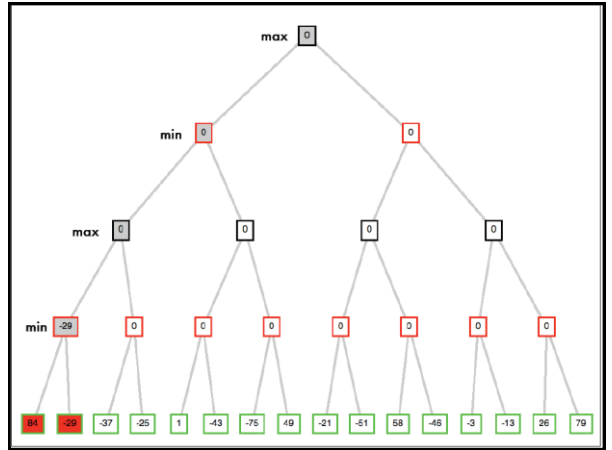
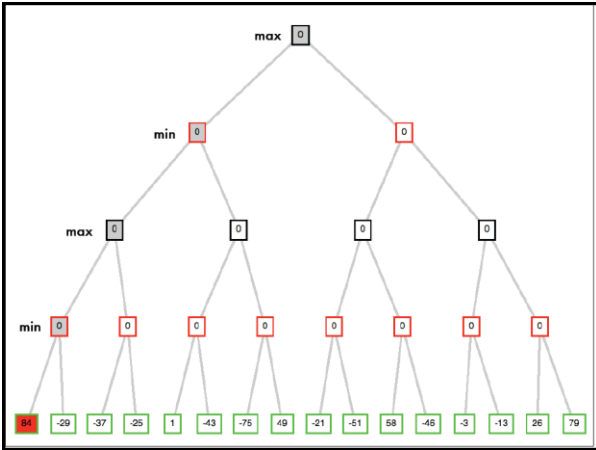
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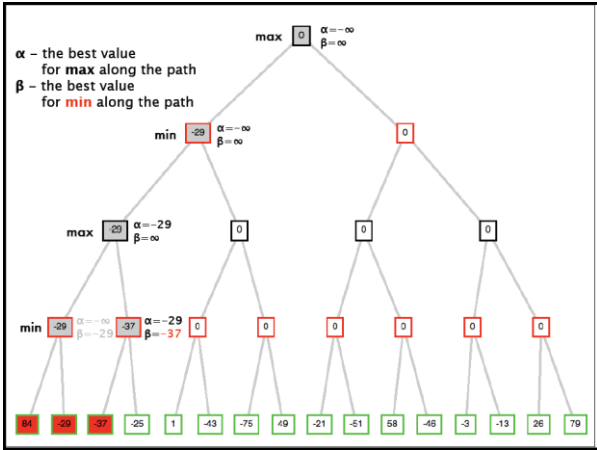
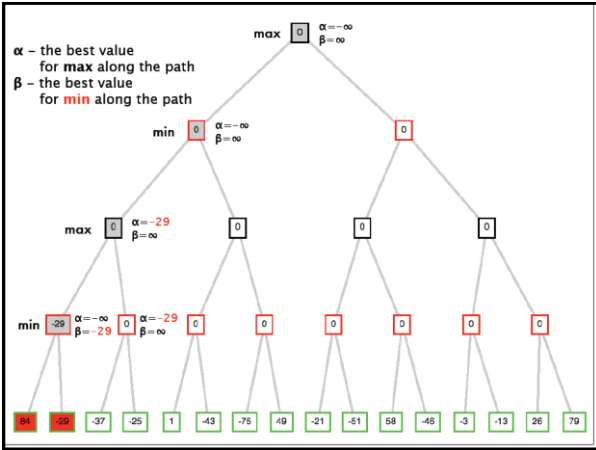
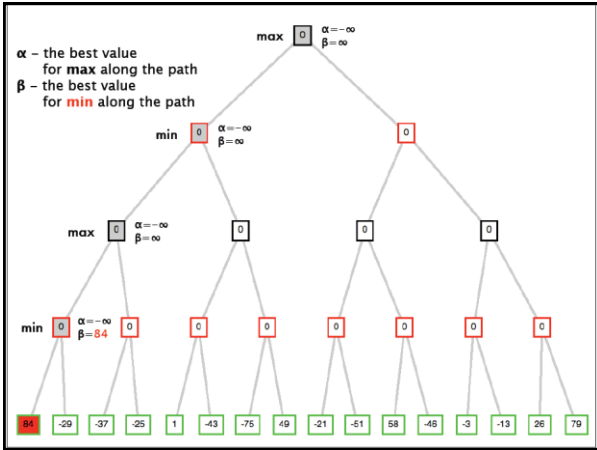
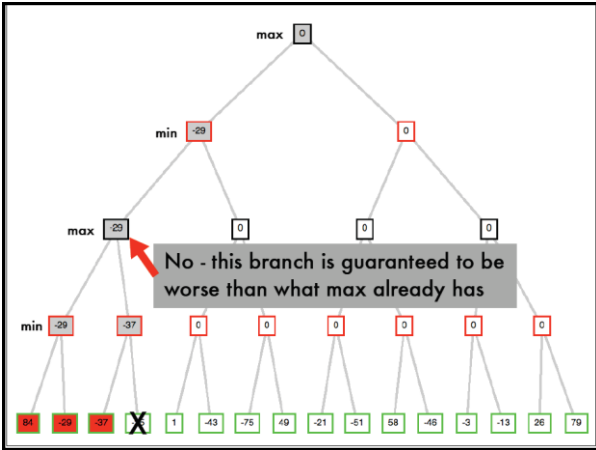
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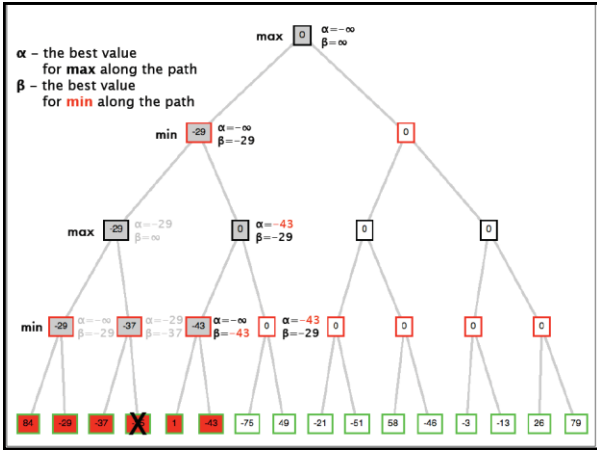
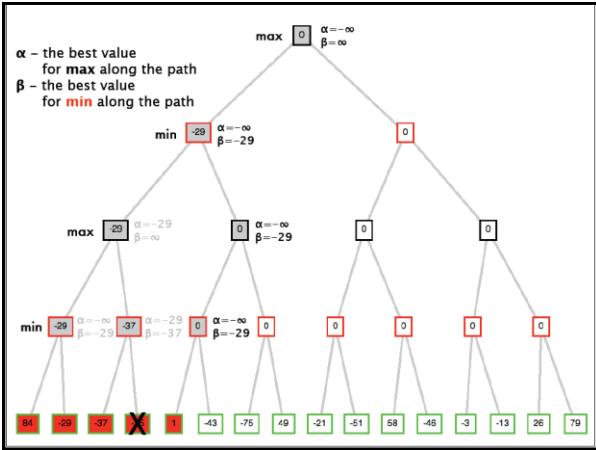
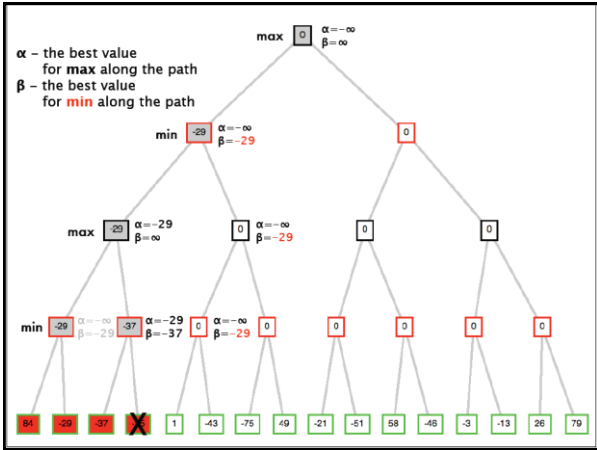
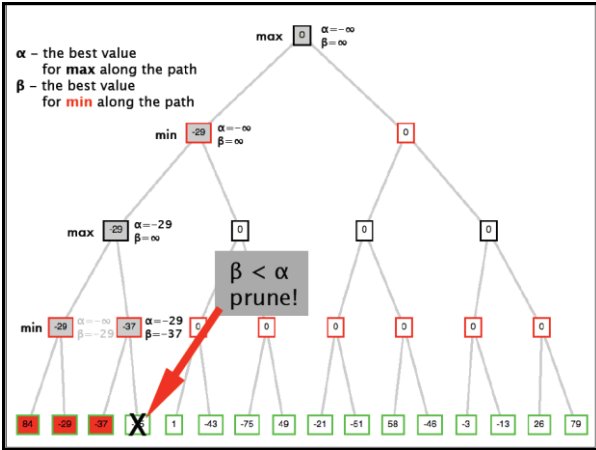
Alpha-Beta Pruning

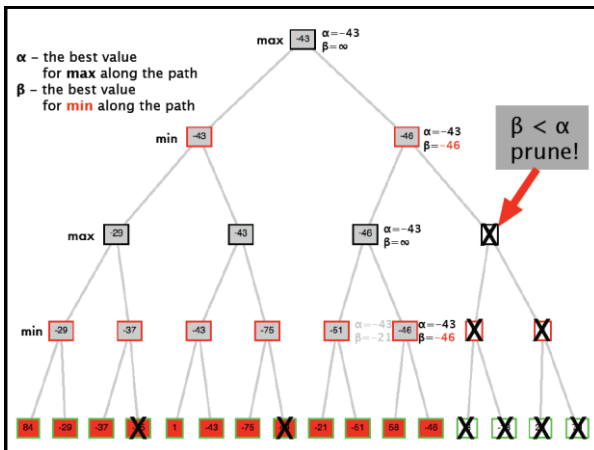
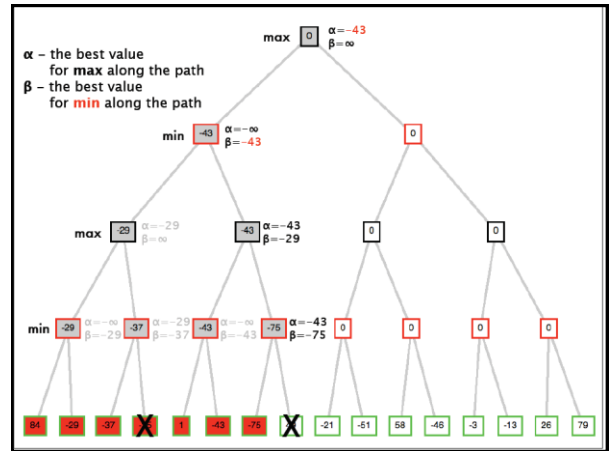
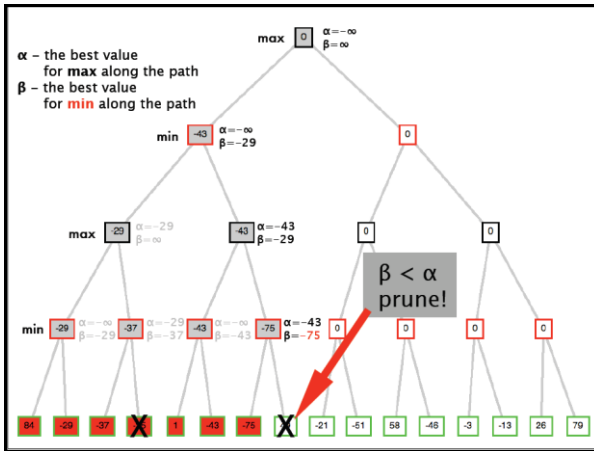
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Alpha-Beta

```

MaxVal(state, alpha, beta){
  if (terminal(state)) return utility(state);
  for (s in successors(state)){
    child = MinVal(s,alpha,beta);
    alpha = max(alpha,child);
    if (alpha>=beta) return child;
  }
  return alpha;
}
  
```

α = the highest (best) value for MAX along path
 β = the lowest (best) value for MIN along path

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Alpha-Beta

```
MinVal(state, alpha, beta){
  if (terminal(state)) return utility(state);
  for (s in successors(state)){
    child = MaxVal(s,alpha,beta);
    beta = min(beta,child);
    if (beta <=alpha) return child;
  }
  return beta;
}
```

alpha = the highest (best) value for MAX along path
beta = the lowest (best) value for MIN along path

Properties of α - β

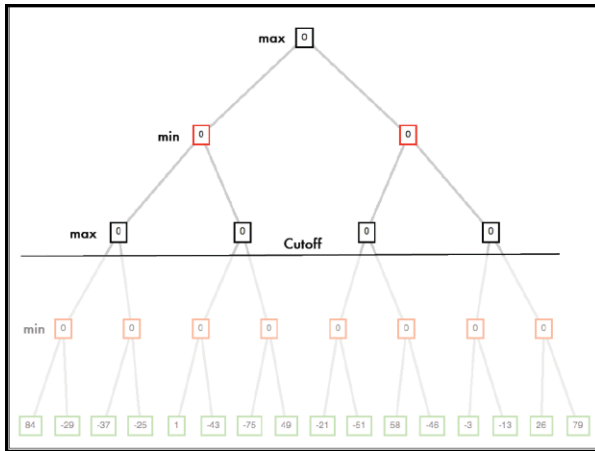
- Still optimal, pruning **does not** affect final result
- Good move ordering improves effectiveness of pruning
- With "perfect ordering," time complexity = $O(b^{m/2})$
→ **doubles** depth of search
- A simple example of the value of reasoning about which computations are relevant (a form of **metareasoning**)

Good enough?

- **Chess:**
 - branching factor $b \approx 35$
 - game length $m \approx 100$
 - search space $b^{m/2} \approx 35^{50} \approx 10^{77}$
- **The Universe:**
 - number of atoms $\approx 10^{78}$
 - age $\approx 10^{21}$ milliseconds

Partial State Spaces

- **Strategies:**
 - search to a fixed depth
 - iterative deepening (most common)
 - stop only at 'quiescent' nodes



Evaluation Function

- When search space is too large, create game tree up to a certain depth only.
- Art is to evaluate positions that are not terminal states.
- Example of simple evaluation criteria in chess:
 - Material worth: pawn=1, knight =3, rook=5, queen=9.
 - Other: king safety, good pawn structure
 - Rule of thumb: 3-point advantage = certain victory

$eval(s) =$

$$\begin{aligned}
 &w1 * material(s) + \\
 &w2 * mobility(s) + \\
 &w3 * king\ safety(s) + \\
 &w4 * center\ control(s) + \dots
 \end{aligned}$$

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Cutting off search

- Does it work in practice?
 - $b^m = 10^6, b=35 \rightarrow m=4$
- 4-ply lookahead is a hopeless chess player!
 - 4-ply \approx human novice
 - 8-ply \approx typical PC, human master
 - 12-ply \approx Deep Blue, Kasparov

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Transposition Tables

- Game trees contain repeated states
- In chess, e.g., the game tree may have 35^{100} nodes, but there are only 10^{40} different board positions
- Similar to closed list in search, maintain a transposition table
 - Got its name from the fact that the same state is reached by a transposition of moves.

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Game Playing in Practice

- **Checkers:** Solved! It has been shown that there is no strategy to beat the computer. The best you can get is a draw.
- **Chess:** Deep Blue defeated human world champion Gary Kasparov in a 6 game match in 1997. Deep Blue searches 200 million positions per second, uses very sophisticated evaluation, and undisclosed methods for extending some lines of search up to 40 ply
- **Othello:** human champions refuse to play against computers because **software is too good**
- **Go:** human champions refuse to play against computers because **software is too bad**

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Summary of Deterministic Games

- Basic idea: minimax -- too slow for most games
- Alpha-Beta pruning can increase max depth by factor up to 2
- Limited depth search may be necessary
- Static evaluation functions necessary for limited depth search and help alpha-beta
- Opening game and End game databases can help
- Computers can beat humans in some games (checkers, chess, othello) but not in others (Go)

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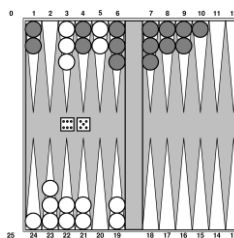
Other Games

	deterministic	chance
Perfect information	chess, checkers, go, othello	backgammon, monopoly
Imperfect information		bridge, poker, scrabble

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Games that Include an Element of Chance



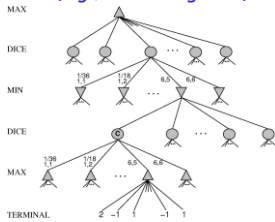
White has just rolled 6-5 and has 4 legal moves.

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Game Tree for Games with an Element of Chance

- In addition to MIN- and MAX nodes, we need chance nodes (e.g., for rolling dice).



- Search costs increase: Instead of $O(b^d)$, we get $O((bn)^d)$, where n is the number of chance outcomes.

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Imperfect Information

- E.g. card games, where opponents' initial cards are unknown
- Idea: For all deals consistent with what you can see
 - compute the minimax value of available actions for each of possible deals
 - compute the expected value over all deals

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