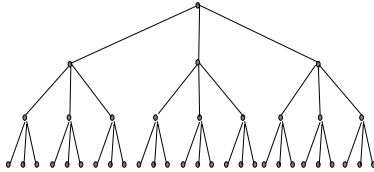


### Search Space Size Reductions

**Worst Case:** In an ordering where worst options evaluated first, all nodes must be examined.

**Best Case:** If nodes ordered so that the best options are evaluated first, then what?



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### The Need for Imperfect Decisions

**Problem:** Minimax assumes the program has time to search to the terminal nodes.

**Solution:** Cut off search earlier and apply a heuristic evaluation function to the leaves.

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### Static Evaluation Functions

Minimax depends on the translation of board quality into a single, summarizing number. Difficult. Expensive.

- Add up values of pieces each player has (weighted by importance of piece).
- Isolated pawns are bad.
- How well protected is your king?
- How much maneuverability to you have?
- Do you control the center of the board?
- Strategies change as the game proceeds.

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### Design Issues for Heuristic Minimax

**Evaluation Function:** Need to be carefully crafted and depends on game! What criteria should an evaluation function fulfill?

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### Linear Evaluation Functions

- $w_1f_1 + w_2f_2 + \dots + w_nf_n$
- This is what most game playing programs use
- Steps in designing an evaluation function:
  1. Pick informative features
  2. Find the weights that make the program play well

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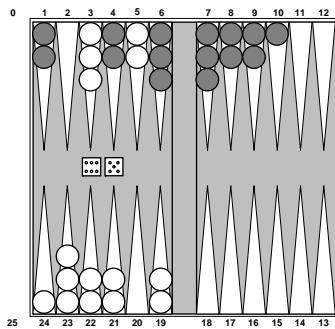
### Design Issues for Heuristic Minimax

**Search:** search to a constant depth

What are problems with constant search depth?

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### Backgammon – Board



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### Backgammon – Rules

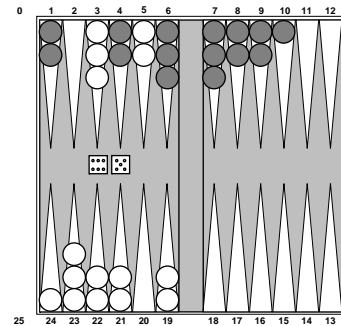
- Goal: move all of your pieces off the board before your opponent does.
- Black moves counterclockwise toward 0.
- White moves clockwise toward 25.
- A piece can move to any position except one where there are two or more of the opponent's pieces.
- If it moves to a position with one opponent piece, that piece is captured and has to start its journey from the beginning.

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### Backgammon – Rules

- If you roll doubles you take 4 moves (example: roll 5,5, make moves 5,5,5,5).
- Moves can be made by one or two pieces (in the case of doubles by 1, 2, 3 or 4 pieces)
- And a few other rules that concern *bearing off* and *forced moves*.

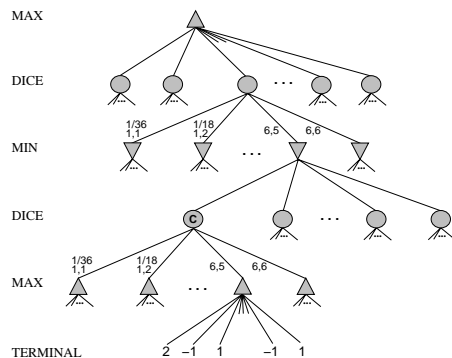
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White has rolled 6-5 and has 4 legal moves: (5-10,5-11), (5-11,19-24), (5-10,10-16) and (5-11,11-16).

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### Game Tree for Backgammon



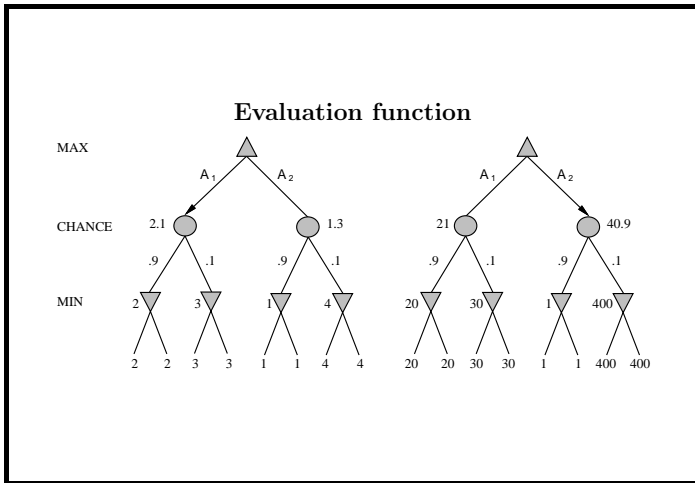
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### Expectiminimax

Expectiminimax (n) =

utility(n)	for n, a terminal state
$\max_{s \in \text{Succ}(n)} \text{expectiminimax}(s)$	for n, a Max node
$\min_{s \in \text{Succ}(n)} \text{expectiminimax}(s)$	for n, a Min node
$\sum_{s \in \text{Succ}(n)} P(s) * \text{expectiminimax}(s)$	for n, a chance node

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### State of the Art in Backgammon

- 1980: *BKG* using two-ply (depth 2) search and lots of luck defeated the human world champion.
- 1992: Tesauro combines Samuel's learning method with neural networks to develop a new evaluation function (search depth 2-3), resulting in a program ranked among the top 3 players in the world.

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### State of the Art in Checkers

- 1952: Samuel developed a checkers program that learned its own evaluation function through self play.
- 1990: *Chinook* (J. Schaeffer) wins the U.S. Open. At the world championship, Marion Tinsley beat *Chinook*.
- 2005: Schaeffer et al. solved checkers for "White Doctor" opening (draw) (about 50 other openings).

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### State of the Art in Go

Large branching factor makes regular search methods inappropriate.

Best computer Go programs ranked only "weak amateur".

Employ pattern recognition techniques and limited search.

\$2,000,000 prize available for first computer program to defeat a top level player.

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### History of Chess in AI

500	legal chess
1200	occasional player
2000	world-ranked
2900	Gary Kasparov

**Early 1950's** Shannon and Turing both had programs that (barely) played legal chess (500 rank).

**1950's** Alex Bernstein's system, (500+ $\epsilon$ ).

**1957** Herb Simon claims that a computer chess program would be world chess champion in 10 years...yeah, right.

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**1966** McCarthy arranges computer chess match, Stanford vs. Russia. Long, drawn-out match. Russia wins.

**1967** Richard Greenblatt, MIT. First of the modern chess programs, *MacHack* (1100 rating).

**1968** McCarthy, Michie, Papert bet Levy (rated 2325) that a computer program would beat him within 10 years.

**1970** ACM started running chess tournaments. Chess 3.0-6 (rated 1400).

**1973** By 1973...Slate: "It had become too painful even to look at Chess 3.6 any more, let alone work on it."

**1973** Chess 4.0: smart plausible-move generator rather than

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speeding up the search. Improved rapidly when put on faster machines.

**1976** Chess 4.5: ranking of 2070.

**1977** Chess 4.5 vs. Levy. Levy wins.

**1980's** Programs depend on search speed rather than knowledge (2300 range).

**1993** DEEP THOUGHT: Sophisticated special-purpose computer;  $\alpha - \beta$  search; searches 10-ply; singular extensions; rated about 2600.

**1995** DEEP BLUE: searches 14-ply; iterative deepening  $\alpha - \beta$  search; considers 100–200 billion positions per

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move; regularly reaches depth 14; evaluation function has 8000+ features; singular extensions to 40-ply; opening book of 4000 positions; end-game database for 5-6 pieces.

**1997** DEEP BLUE: first match won against world-champion (Kasparov).

**2002** IBM declines re-match. FRITZ played world champion Vladimir Kramnik. 8 games. Ended in a draw.

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