

## CS61B Lecture #22

**Today:** Backtracking searches, game trees (*DSIJ*, Section 6.)

## Searching by “Generate and Test”

- We've been considering the problem of searching a set of data in some kind of data structure: “Is  $x \in S$ ?”
- But suppose we *don't* have a set  $S$ , but know how to recognize what we're after if we find it: “Is there an  $x$  such that  $P(x)$ ?”
- If we know how to enumerate all possible candidates, can we approach of *Generate and Test*: test all possibilities in turn.
- Can sometimes be more clever: avoid trying things that won't work for example.
- What happens if the set of possible candidates is infinite?

## Backtracking Search

- Backtracking search is one way to enumerate all possibilities
- Example: *Knight's Tour*. Find all paths a knight can travel on a board such that it touches every square exactly once and one knight move from where it started.
- In the example below, the numbers indicate position numbers (starts at 0).
- Here, knight (N) is stuck; how to handle this?

6							
		5					
4	7						
	10		2				
8	3	0					
N		9		1			

## General Recursive Algorithm

```
/** Append to PATH a sequence of knight moves starting at ROW,
 * that avoids all squares that have been hit already and
 * that ends up one square away from ENDROW, ENDCOL. B[i][j] :
 * true iff row i and column j have been hit on PATH so far.
 * Returns true if it succeeds, else false (with no change to
 * Call initially with PATH containing the starting square, and
 * the starting square (only) marked in B. */
```

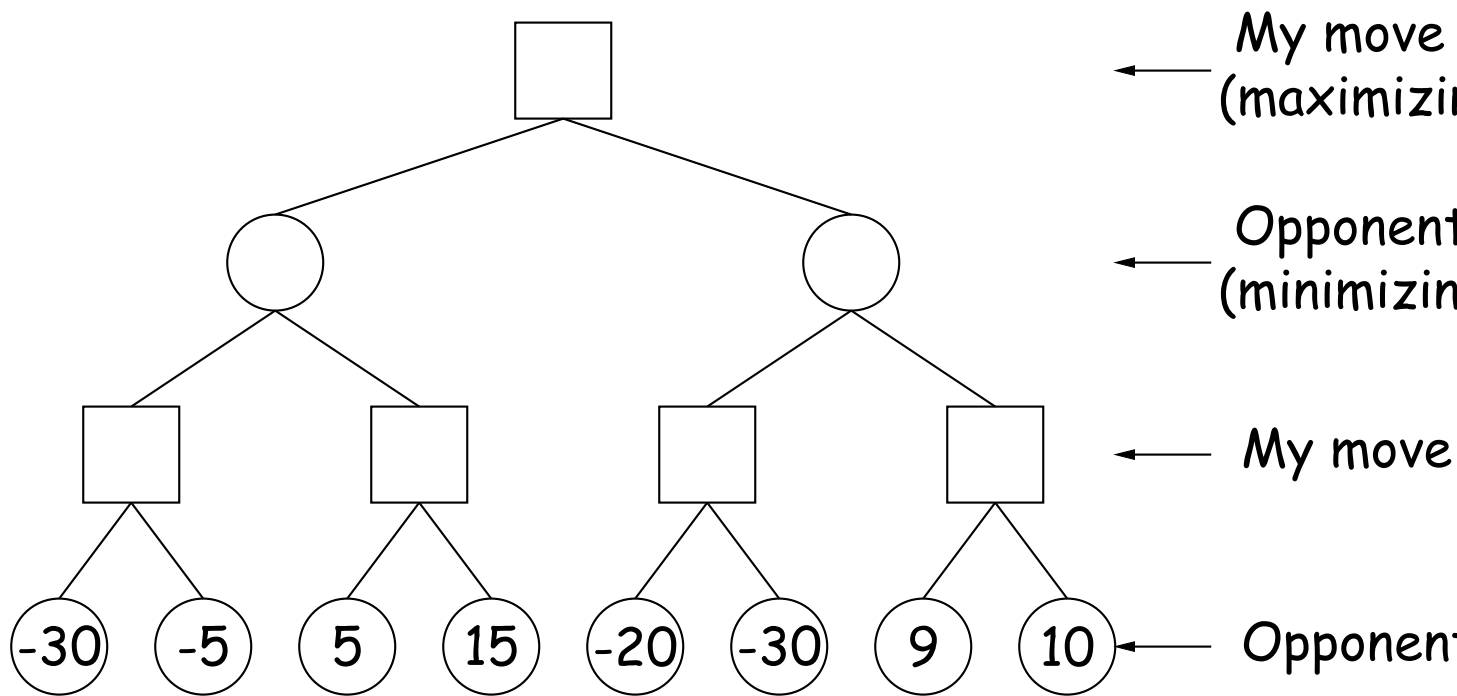
```
boolean findPath(boolean[][] b, int row, int col,
                  int endRow, int endCol, List path) {
    if (path.size() == 64) return isKnightMove(row, col, endRow, endCol);
    for (r, c = all possible moves from (row, col)) {
        if (!b[r][c]) {
            b[r][c] = true; // Mark the square
            path.add(new Move(r, c));
            if (findPath(b, r, c, endRow, endCol, path)) return true;
            b[r][c] = false; // Backtrack out of the move.
            path.remove(path.size()-1);
        }
    }
    return false;
}
```

## Another Kind of Search: Best Move

- Consider the problem of finding the *best* move in a two-person game.
- One way: assign a *heuristic value* to each possible move and choose the highest (aka *static evaluation*). Examples:
  - number of black pieces – number of white pieces in check
  - weighted sum of white piece values – weighted sum of black piece values (Queen=9, Rook=5, etc.)
  - Nearness of pieces to strategic areas (center of board).
- But this is misleading. A move might give us more pieces, but also a devastating response from the opponent.
- So, for each move, look at *opponent's* possible moves, assume the opponent picks the best one for him, and use that as the value.
- But what if *you* have a great response to his response?
- How do we organize this sensibly?

# Game Trees

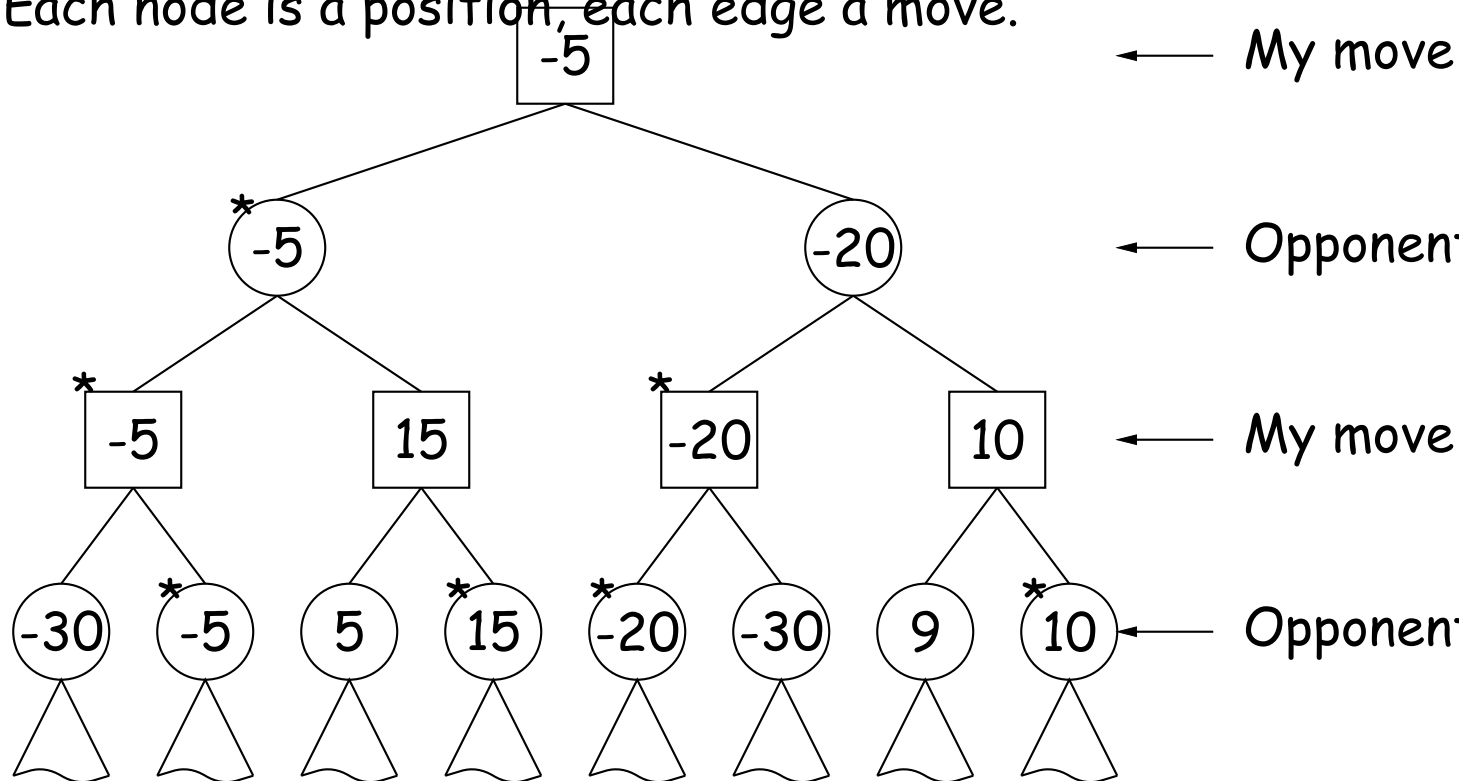
- Think of the space of possible continuations of the game as
- Each node is a position, each edge a move.



- Suppose numbers at the bottom are the values of those final positions *to me*. Smaller numbers are of more value to *my opponent*.
- What should I move? What value can I get if my opponent moves as well as possible?

## Game Trees, Minimax

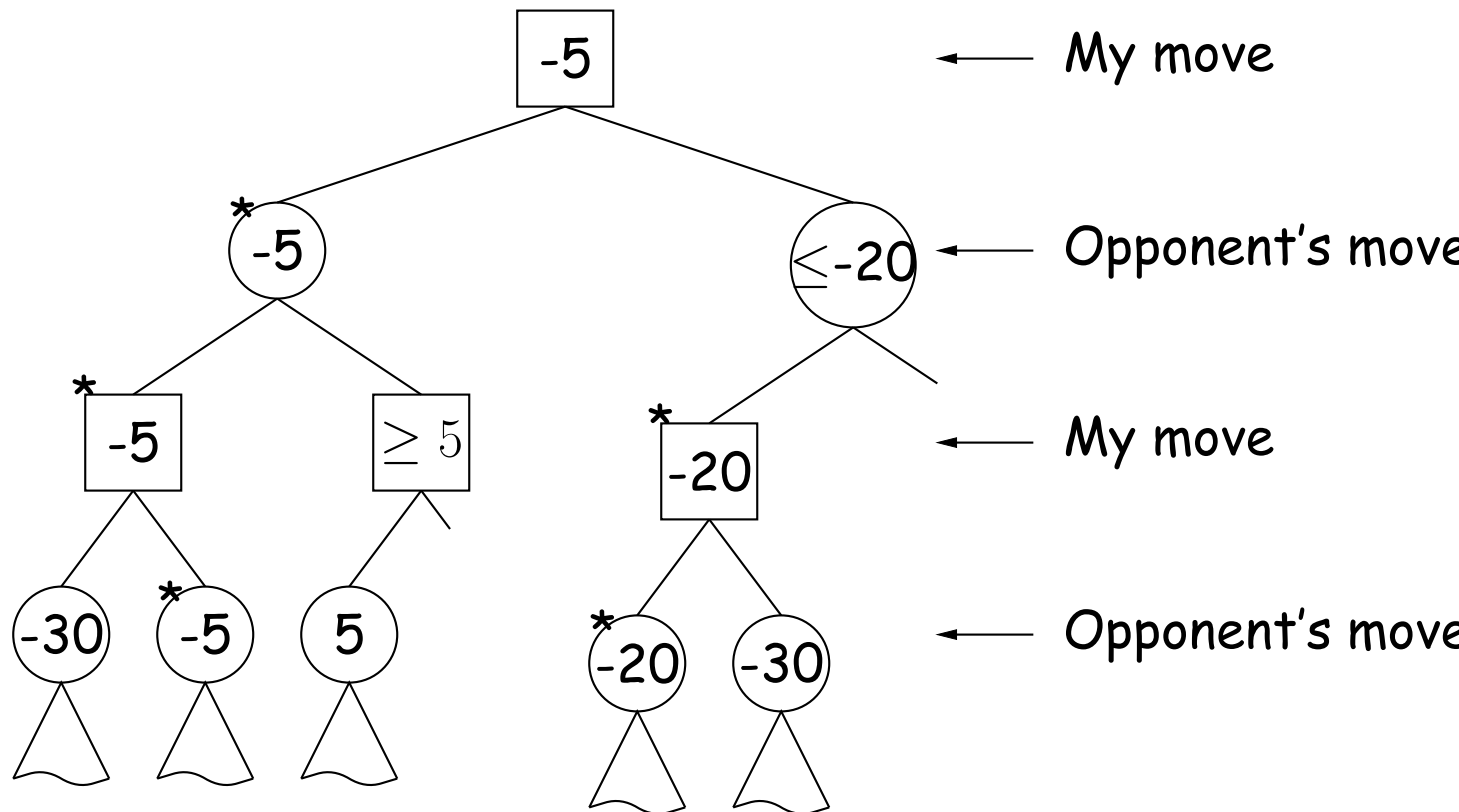
- Think of the space of possible continuations of the game as
- Each node is a position, each edge a move.



- Numbers are the values we guess for the positions (larger is better for me). Starred nodes would be chosen.
- I always choose child (next position) with maximum value; opponent chooses minimum value ("Minimax algorithm")

# Alpha-Beta Pruning

- We can *prune* this tree as we search it.



- At the ' $\geq 5$ ' position, I know that the opponent will not choose to move here (since he already has a  $-5$  move).
- At the ' $\leq -20$ ' position, my opponent knows that I will never choose to move here (since I already have a  $-5$  move).



## Cutting off the Search

- If you could traverse game tree to the bottom, you'd be able to force a win (if it's possible).
- Sometimes possible near the end of a game.
- Unfortunately, game trees tend to be either infinite or impossibly large.
- So, we choose a maximum *depth*, and use a heuristic value computed on the position alone (called a *static valuation*) as the value at the maximum depth.
- Or we might use *iterative deepening*, repeating the search at increasing depths until time is up.
- Much more sophisticated searches are possible, however (take a look at the next slide).

## Overall Search Algorithm

- Depending on whose move it is (maximizing player or minimizing player) we'll search for a move estimated to be optimal in one direction or the other.
- Search will be exhaustive down to a particular depth in the search tree; below that, we guess values.
- Also pass  $\alpha$  and  $\beta$  limits:
  - High player does not care about exploring a position further if he knows its value is larger than what the minimizing player can get ( $\beta$ ), because the minimizing player will never allow a position to come about.
  - Likewise, minimizing player won't explore a position whose value is less than what the maximizing player knows he can get.
- To start, a maximizing player will find a move with

`findMax(current position, search depth,  $-\infty$ ,  $+\infty$ )`

- minimizing player:

`findMin(current position, search depth,  $-\infty$ ,  $+\infty$ )`

## Some Pseudocode for Searching (One Level)

- The most basic kind of game-tree search is to assign some heuristic value to any given position, looking at just the next possible

```
Move simpleFindMax(Position posn, double alpha, double beta)
    if (posn.maxPlayerWon())
        return artificial "Move" with value  $+\infty$ ;
    else if (posn.minPlayerWon())
        return artificial "Move" with value  $-\infty$ ;
    Move bestSoFar = artificial "Move" with value  $-\infty$ ;
    for (each M = a legal move for maximizing player from posn)
        Position next = posn.makeMove(M);
        next.setValue(heuristicEstimate(next));
        if (next.value() >= bestSoFar.value()) {
            bestSoFar = next;
            alpha = max(alpha, next.value());
            if (beta <= alpha) break;
        }
    }
    return bestSoFar;
}
```

## One-Level Search for Minimizing Player

```
Move simpleFindMin(Position posn, double alpha, double beta)
{
    if (posn.maxPlayerWon())
        return artificial "Move" with value  $+\infty$ ;
    else if (posn.minPlayerWon())
        return artificial "Move" with value  $-\infty$ ;
    Move bestSoFar = artificial "Move" with value  $+\infty$ ;
    for (each M = a legal move for minimizing player from posn)
    {
        Position next = posn.makeMove(M);
        next.setValue(heuristicEstimate(next));
        if (next.value() <= bestSoFar.value()) {
            bestSoFar = next;
            beta = min(beta, next.value());
            if (beta <= alpha) break;
        }
    }
    return bestSoFar;
}
```

## Some Pseudocode for Searching (Maximizing Player)

```
/** Return a best move for maximizing player from POSN
 * to depth DEPTH. Any move with value >= BETA is already
 * "good enough". */
Move findMax(Position posn, int depth, double alpha, double beta) {
    if (depth == 0 || gameOver(posn))
        return simpleFindMax(posn, alpha, beta);
    Move bestSoFar = artificial "Move" with value  $-\infty$ ;
    for (each M = a legal move for maximizing player from posn) {
        Position next = posn.makeMove(M);
        Move response = findMin(next, depth-1, alpha, beta);
        if (response.value() >= bestSoFar.value()) {
            bestSoFar = next;
            next.setValue(response.value());
            alpha = max(alpha, response.value());
            if (beta <= alpha) break;
        }
    }
    return bestSoFar;
}
```

## Some Pseudocode for Searching (Minimizing Player)

```
/** Return a best move for minimizing player from POSN
 * to depth DEPTH. Any move with value <= ALPHA is a
 * "good enough". */
Move findMin(Position posn, int depth, double alpha, double beta) {
    if (depth == 0 || gameOver(posn))
        return simpleFindMin(posn, alpha, beta);
    Move bestSoFar = artificial "Move" with value  $+\infty$ ;
    for (each M = a legal move for minimizing player from posn) {
        Position next = posn.makeMove(M);
        Move response = findMax(next, depth-1, alpha, beta);
        if (response.value() <= bestSoFar.value()) {
            bestSoFar = next;
            next.setValue(response.value());
            beta = min(beta, response.value());
            if (beta <= alpha) break;
        }
    }
    return bestSoFar;
}
```