

Game Trees

- Suppose that we have a two-player game in which the players take turns making moves.
 - Further, we assume that the game will always end in either (1) a victory by one player over the other or (2) a draw.
- We can model the game as a *multiway* tree.
 - Each node in the *game tree* corresponds to a possible configuration of the “game board” .
 - The initial configuration of the “game board” is the root node.
 - A leaf node corresponds to an end-of-game configuration: a win, a loss, or a draw.

Game Trees: An Example (cont.)

- Assume the players are John and Mary.
- When John is to make a move, he will move to a game node that is best for him.
 - What is best for John is based on some numerical value that he associates with each of the possible next moves.
 - * How are the value of each of the game node calculated?
 - What follows is one such possible computation.

Min-Max Algorithm

- For each *leaf* node X , John can assign a value of 1 to X if he wins, 0 if he ties, and -1 if he loses.

– Conceptually, John defines a “pay-off” function P as follows:

$$P(X) = \begin{array}{ll} 1 & \text{if John wins} \\ 0 & \text{if John ties} \\ -1 & \text{if John loses} \end{array}$$

where X is a leaf .

Min-Max Algorithm (cont.)

- John assigns a value to a *non-leaf* node X , as follows:

$$V(X) = \begin{array}{l} \max \{ V(c) \mid c \text{ is a child node of } X \} \\ \text{if John moves next} \\ \\ \min \{ V(c) \mid c \text{ is a child node of } X \} \\ \text{if Mary moves next} \end{array}$$

where X is not a leaf.

- The idea is that John would move to a node that has the maximum value for him, and Mary would do her best by moving to a node that has the minimum value (from John's perspective).

Min-Max Algorithm (cont.)

- In general, it is not possible to examine all leaf nodes of a non-trivial game.
 - At best, you can examine the game tree up to certain depth.

Modified Min-Max Algorithm

- Instead of flip-flopping between max and min as described above, we can reformulate the min-max strategy based on the simple mathematical formula:

$$\max(a, b) = -\min(-a, -b)$$

Modified Min-Max Algorithm (cont.)

- Let

$E(n)$ be the pay-off function that John uses to evaluate a game node n .

- Let

$e(n) = E(n)$ if n is node from which John is to make the next move

$-E(n)$ if Mary is to make the next move.

Modified Min-Max Algorithm (cont.)

- Let

$\text{ModMinMax}(x) = e(x)$, if x is a leaf of the game subtree
 $\max(-\text{ModMinMax}(c))$, if x is not a leaf
of the game subtree and c ranges over all
of the (immediate) children of x .

Modified Min-Max Algorithm (cont.)

- Given x , a game tree node (i.e., a game board configuration), and d , the number of lookahead moves from x , compute the value of x based on the min-max formula:

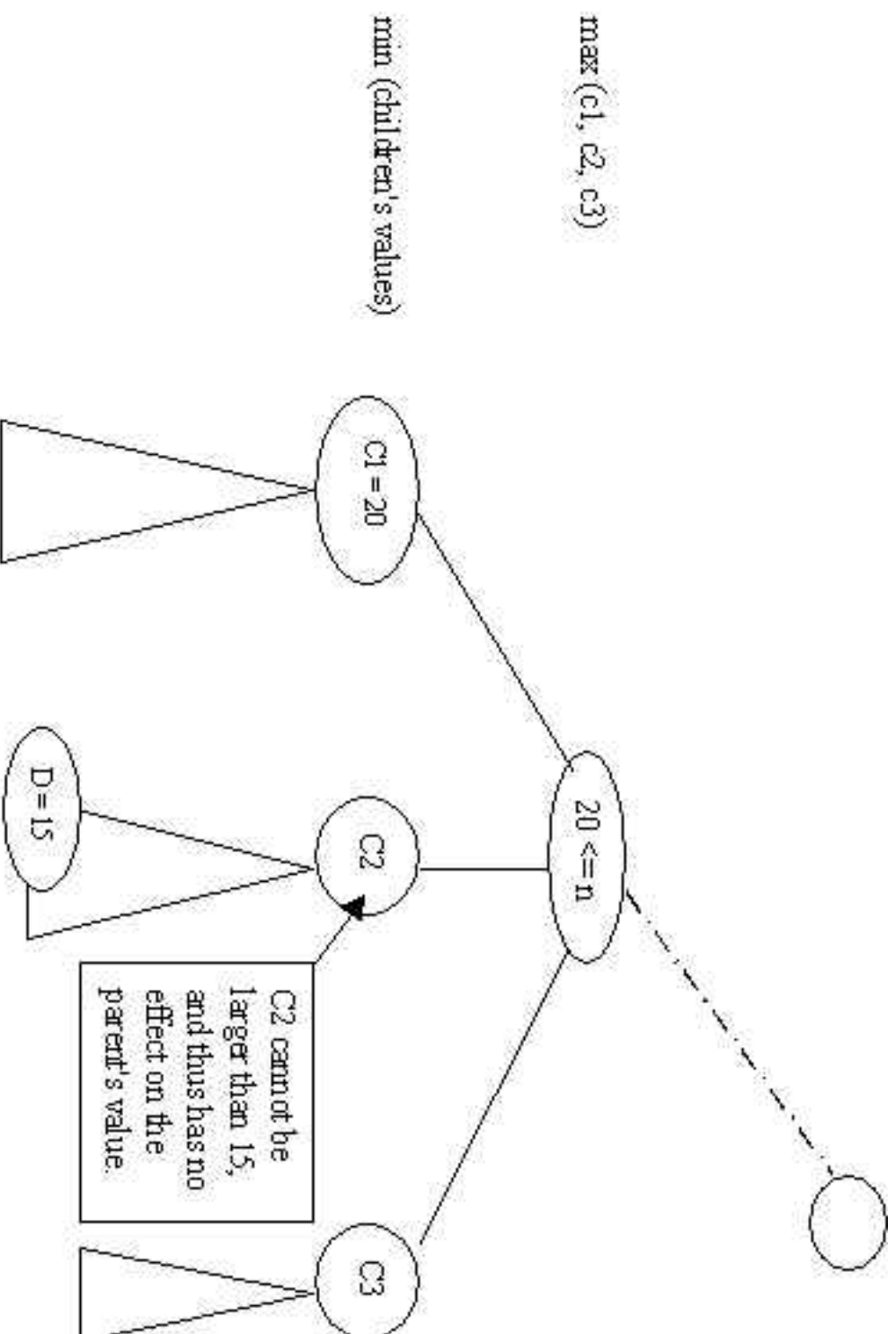
```
int ModMinMax(GameNode x, int d)
    if x is a leaf node or d == 0
        then return e(x)
    else // c[0] is the first child node of x.
        ans = -ModMinMax(c[0], d - 1)
        for (int i = 1; i < number of children of x; i++)
            temp = -ModMinMax(c[i], d - 1);
            if ans < temp
                then ans = temp;
        return ans;
```

Alpha-Beta Pruning

- In computing the min-max value of a game tree node, we can skip (“prune”) the evaluation of some of the node’s children.
- Let alpha be a lower bound for the value of a max node A , and let B be a child node of A .
 - If the value v of a child of B is less or equal to alpha, then we can use v as a value for B and skip the rest of the children of B .
 - * This is called “alpha pruning”.

Alpha Pruning

- In the figure below, $\alpha = 20$, and we can prune the rest of the children of C2, once the value of D is (recursively) computed.



Beta Pruning

- Let beta be an upper bound for the value of a min node B, and let C be a child node of B.
 - If the value v of a child of C is greater or equal to beta, then we can use v as a value for C and skip the rest of the children of C.
 - * This is called "beta pruning".