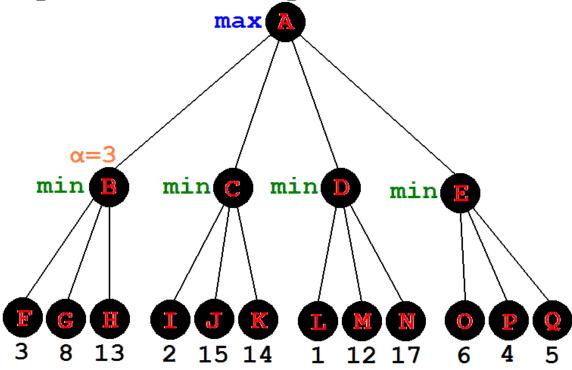
**Alpha-Beta Pruning** 



To understand alpha-beta pruning, consider the situation above. The game is in state A. The computer is to make the next move and it has a choice of going to states B, C, D, or E. For the computer this means a maximization of state value (of course, because the computer is trying to make the best move).

First, the computer examines state B. In state B the computer must simulate, or anticipate, its opponent's next move. This is a minimization problem. Since it will look no deeper than B's children in this example, the computer determines a value of 3 for state B (the min of 3.8.13). Since 0 always tracks the best max value so far, it is set to 3.

Next, the computer examines state C. The first thing it sees is that state I has value I. Consider for a moment what this means for the game. If the computer were to move to state I the opponent would move to state I, with value I. If the computer were to move to state I, the opponent could choose state I or worse (remember, we haven't seen states I or I0 or I1. So, from the

computer's point of view, there is no point in exploring subtrees J or K at all. State B is already better than C, so move on and see if D or E might be better still. The C value allows you to recall what you've seen before, and prune away parts of the tree that won't matter.

The situation with  $\beta$  is simply the mirror image of this.

With the above in mind, the pseudocode should become easier to understand:

```
function MIN_VALUE(state, alpha, beta)
                     utility value for state
 returns:
 inputs:
                     current state of game
           state
                     best max value along path to state
           alpha
                     best min value along path to state
           beta
 If depth-limit-reached return utility(state)
 v := infinity
for all successors, s, of state do
      v := min(v,MAX VALUE(s,alpha,beta))
      If v <= alpha return v
      beta := min(beta, v)
 return v
```