

## Problem 1: solution

MAX(X)

MIN

MAX

X   O   X	X   O   X	X   O   X
X   X	X   X   0	X   X   X
O   O   O	O   O	O   O   O

-1	0	0	0	0	-1
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MIN

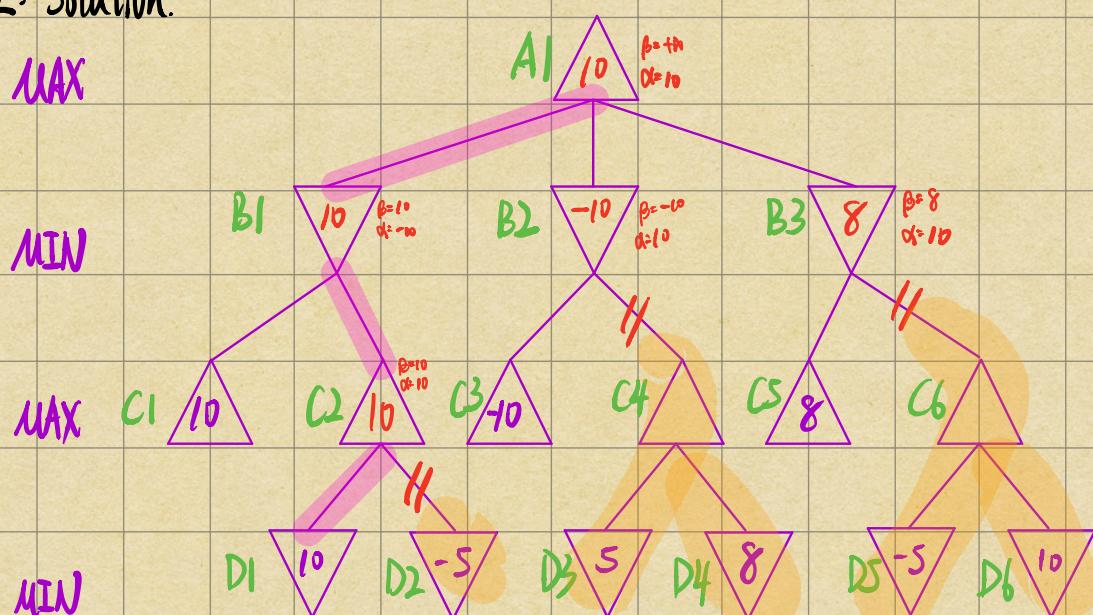
X   O   X	X   O   X	X   O   X	X   O   X
X   X   0	O   X   X	X   X   0	O   X   X
O   X   0	O   X   0	O   X   0	O   X   0

0	0	0	0
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The optimal action for the MAX player(X) to play is green path

## Problem 2: Solution.

a. MAX

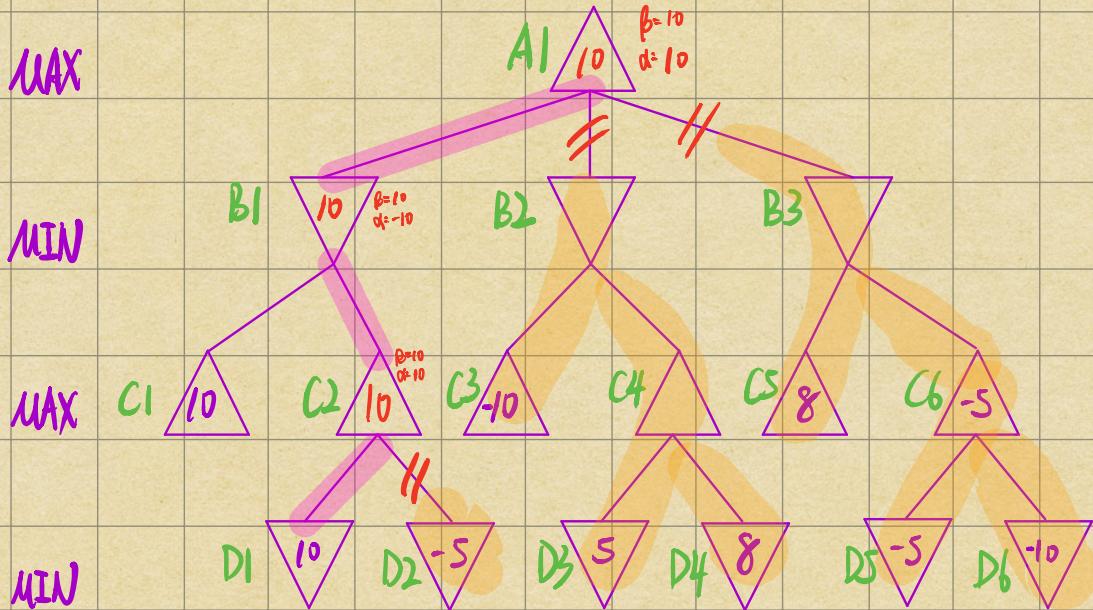


Thus, D2, C4, D3, D4, C6, D5, D6 will be pruned.

The estimated utility values for the rest of the nodes are

A1:10, B1:10, B2:8, B3:8, C2:10

b. Given additional knowledge, we obtain  $\alpha = -10$ ,  $\beta = 10$  as prior condition.

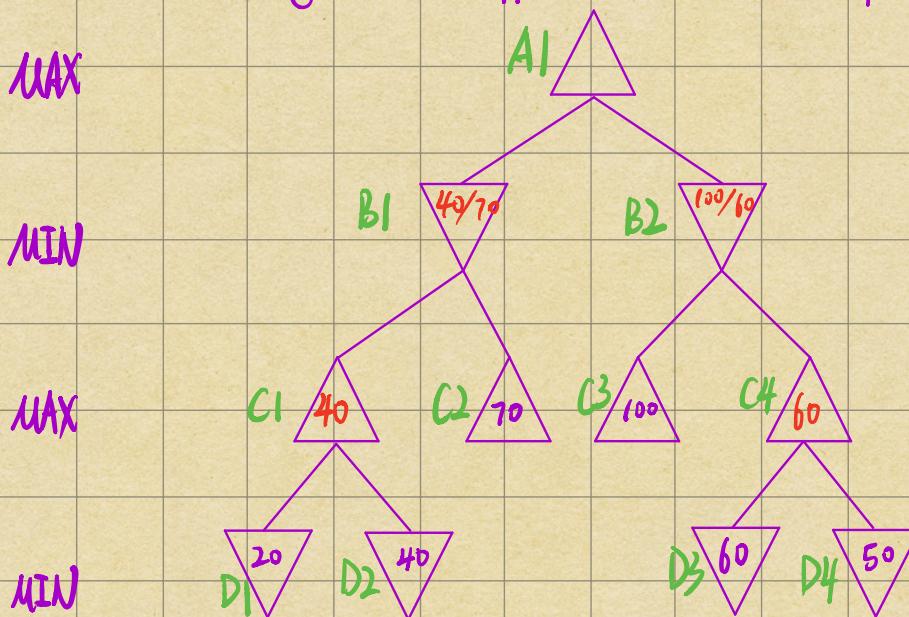


As the pruned tree above, we find that this knowledge can further improve the efficiency.

The pruned nodes are : B2, C3, C4, D3, D4, B3, C5, C6, D5, D6,

### Problem 3 : Solution.

Because we don't know algorithm the opponents uses, we have four possible second level nodes:  
**(MIN player)**

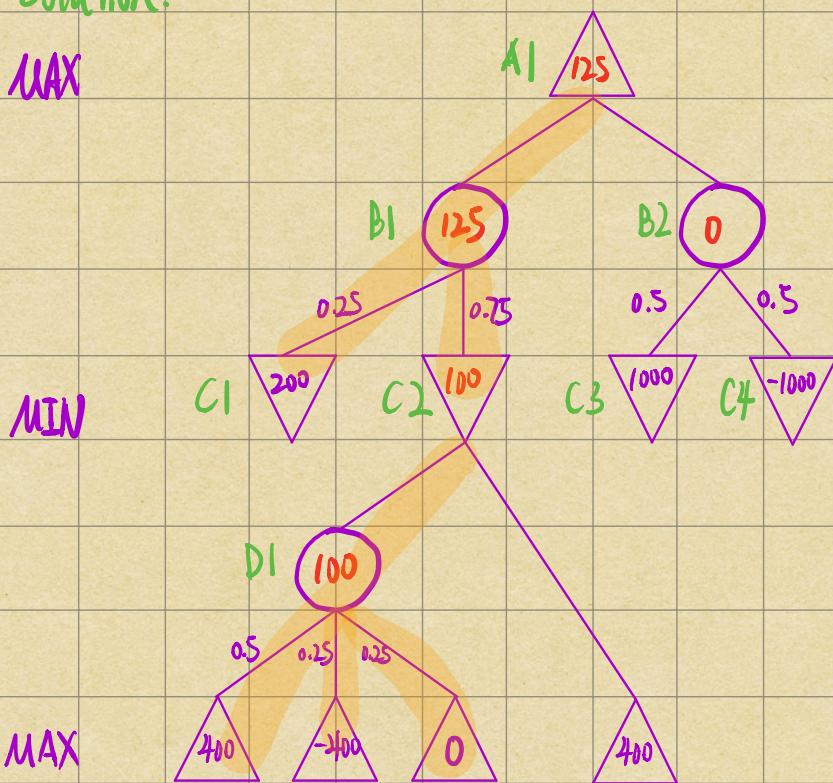


$\{B1:40, B2:100\}$ ,  $\{B1:40, B2:60\}$ ,  $\{B1:70, B2:100\}$ ,  $\{B1:70, B2:60\}$

For each case, we obtain 4 possible A : { $A_{11} : 100$ ,  $A_{12} : 60$ ,  $A_{13} : 100$ ,  $A_{14} : 70$ }.

Thus, the best possible outcome is 100, the worst possible outcome is 60.

Problem 4: solution.



1' The value of non-terminal nodes are  $A_1: 125$ ,  $B_1: 125$ ,  $B_2: 0$ ,  $C_2: 100$ ,  $D_1: 100$

2' The strategy will choose left subpath from  $A_1$ ,

3' The Minmax value obtained by the root node represent the expectation of the Minimax Algorithm. of MAX player.

4' Maximum payoff is 400, minimum payoff is -400 if MIN plays the optimal strategy.

5' If MIN plays a random strategy,  $C_2$  could have {100, 400}.

but the Maximum payoff is also 400, the minum payoff is also -400.

Problem 5:

We can just modify the MINMAX algorithm to solve this problem.

1' Adjust the minValue(state) function through replace Successor(state) function to DeepGreenMove(state).

function beatDeepGreen(state) returns an action

inputs: state, current state in game

returns: the  $a$  in ACTIONS(state) maximizing minValue(result(a, state))

function maxValue(state) return a utility value

If terminalTest(state) then return utility(state)

$v \leftarrow -\infty$

for  $a, s$  in successors(state) do  $v \leftarrow \max(v, minValue(s))$

return  $v$

function minValue(state) return a utility value

If terminalTest(state) then return utility(state)

$v \leftarrow \infty$

for  $a, s$  in DeepGreenMove(state) do  $v \leftarrow \min(v, maxValue(s))$

return  $v$