

# COMP3121: Assignment 3 – Q2

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Given the values for  $R_a$ ,  $R_b$ ,  $P_a$ ,  $P_b$ ,  $S_a$  and  $S_b$ , we will first give a general algorithm that minimises the number of losses, which in turn maximises the number of points won, and finally *show* that this produces an optimal solution. That is, *no other algorithm can be any better than our current algorithm*.

## General algorithm

Let  $N$  be the number of rounds played between Player  $A$  and Player  $B$ . The idea is that we want to *minimise* the number of losses. We do this by maximising the number of wins and then playing draws if possible. As such, we reserve as many paper for  $R_a$ , scissors for  $P_a$  and rock for  $S_a$ . In each round, Player  $B$  will play their most optimal hand possible given the plays that  $A$  has made.  $B$  applies a greedy search to find a play that wins the round. For example, if  $A$  plays a rock, then  $B$  will apply a greedy search that looks for a paper in their hand. If that search returns a match, then  $B$  will play that hand which wins the round. However, if  $B$  doesn't find a match, then the next best option is for  $B$  to draw so that they do not lose any points.  $B$  will apply another greedy search to find a rock. If that search returns a match, then  $B$  will play that hand which draws the round. If this search fails, then the only option is to lose that round in which case,  $B$  will play scissors. We will be guaranteed at least one scissors because we attain the equality  $R_b + P_b + S_b = N$ . So if there are no more beneficial plays, there will *always* be a losing hand to play. This will be the basis for our optimality proof.

## Proof of optimality

We will now prove that this algorithm produces an optimal solution. It should be noted that this algorithm may not be *the only* optimal algorithm but no other strategy will exceed this algorithm. We proceed with a **proof by contradiction**.

Define our algorithm to be algorithm  $X$  played with  $N$  rounds. Assume that there exists a strategy  $X^*$  that is *more* optimal than  $X$ . Then, at some round  $k > 0$ , a loss in  $X$  must have been either a win or a draw in  $X^*$ , or a draw in  $X$  must have been a win in  $X^*$ . We will define our solution space produced by  $X$  to be  $S_N = \{s_1, s_2, \dots, s_k, \dots, s_N\}$ . Let  $k$  be the index of the first iteration where  $X$  finds the suboptimal solution. Then define the solution space produced by  $X^*$  to be  $T_N = \{s_1, s_2, \dots, s_{k-1}, s'_k, s'_{k+1}, \dots, s'_N\}$  since this means that our algorithm produced optimal choices for the first  $(k - 1)$  iterations. We'll consider the two different scenarios where  $X^*$  is more optimal than  $X$ .

- **Case 1:**  $X$  draws and  $X^*$  wins. Then there must have been a round  $0 < n < k$  that we didn't play our most optimal hand because at the  $k$ th round, it must have been possible to win and our greedy algorithm didn't pick up. However, the previous  $(k - 1)$  hands have been optimal as per our algorithm. This is a contradiction and so in this case,  $X^*$  is only as good as  $X$ .
- **Case 2:**  $X$  loses while  $X^*$  either draws or wins. Again, this assumes that there has been no optimal play

possible which yielded in a loss in  $X$ . But this cannot be possible since it follows that  $X^*$  made a more optimal move. Since  $X$  and  $X^*$  played optimal hands, then it is that  $X^*$  would have traded its win and draw now for a loss later on which ultimately yields in a net change of zero. This works because the sum of all of  $R_b$ ,  $S_b$  and  $P_b$  is  $N$ . As such, a win or draw now would lead to a loss later on. This contradicts the optimality of  $X^*$  so in both cases,  $X^*$  is *only as good* as  $X$ .

In both of these cases, we see that  $X^*$  is only as good as  $X$ . Since  $X^*$  is optimal, then it follow that  $X$  is also optimal.