

The algorithm

Iterative-Compute-Opt

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M[0] = 0;
S[0] = empty; # array that holds the current optimal solution (pairs made so far) at each step
for i = 1, 2, ...,  $\binom{n}{k}$ : # each cell of M represents a potential pair, in increasing order by the first
# student of each pair's index, then by the second student's index. E.g. it would look like
# [(1,2), (1,3), (1,4) ... (5,6), (5,7) ...].
    let (j,k) = the pair that i represents # e.g. i=8 represents the pair (2,5) when n=6
    if (vi + p(j,k,n)) > M[i-1]:
        M[i] = vi + p(j,k,n)
        S[i] = S[p(j,k,n)].add(i) # for solution at step i, add student i onto solution at step p(i)
    else:
        M[i] = M[i-1]
        S[i] = S[i-1] # for solution at step i, keep the previous step's solution
    endif
endfor
return S[ $\binom{n}{k}$ ] as the optimal set of all pairs
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p(j,k,n) is a function that, given there are n total students, returns the last pair before a pair of students (j,k) where both students are different from j and k. For example, p(j=3, k=5, n=6) is (2,6); p(j=5, k=6, n=6) is (3,4). If that doesn't exist (the input is (1,x), for example, so all earlier pairs also have student 1), return 0.

v_i is 1 if the pair (j,k) that i represents overlap by at least t, 0 otherwise

Proof of Correctness

Observe that for an optimal solution O, pair (n-1, n) either belongs or doesn't belong to O. If (n-1, n) ∈ O, then O *must* include an optimal solution to the problem consisting of potential pairs {(1,2) ... (n-3, n-2)}. On the other hand, if n ∉ O, then O simply equals the optimal solution to the problem consisting of potential pairs {(1,2) ... (n-2, n)}. We summarize this in a formula that essentially says ∀ potential pairs (j,k) (compacted into one number i so the arrays can access things), either i ∈ O_i, in which case Most_Pairs = v_i + Most_Pairs(p(i)), or i ∉ O_i, in which case Most_Pairs = Most_Pairs(i-1). So

$$\text{Most_Pairs}(i) = \max(v_i + \text{Most_Pairs}(p(i)), \text{Most_Pairs}(i-1))$$

We will now prove by **strong induction** that the algorithm above returns the optimal answer.

Base Case

We want to prove $M[1]$ is the optimal/maximum pairs if there were only one student, and $S[1]$ is empty. By definition,

$$\begin{aligned}M[1] &= \max(v_1 + M[p(1)], M[1-1]) \\ \text{\# the pseudocode essentially does a max function when using the comparison operator} \\ &= \max(0 + M[0], M[0]) \\ &= \max(0, 0) \\ &= 0 \\ \\ S[0] &= \text{empty} \\ S[1] &= S[0] \\ (\text{because } M[1] \text{ is not strictly greater than } M[0]) \\ S[1] &= \text{empty}\end{aligned}$$

Which is correct because there can't be any pairs if there's only one student.

Inductive Case

We need to prove $M[i] = \text{Most_Pairs}[i]$.

Since we're using strong induction, we can assume that $\forall i < j$, $M[i]$ is the maximum number of pairs there can be if there were only the first i possible pairs, and that $S[i]$ holds said pairs. Thus,

$$\begin{aligned}M[j] &= \max(v_j + M[p(j)], M[j-1]) \\ &= \max(v_j + \text{Most_Pairs}[p(j)], \text{Most_Pairs}[j-1])\end{aligned}$$

Which was our definition given above (before the base case is proved).

S keeps track of all the pairs that the running optimal solution consists of. S is constructed such that whichever side "wins" the max function, then that side's set of pairs involved becomes the current running optimal solution.

Runtime Analysis

Iterating through M is $O(n^2)$ because the formula for $\binom{n}{k}$ is $\frac{n(n-1)}{2}$. Calculating $p(i)$ is $O(1)$. The comparison is $O(1)$. Setting elements of S and M is $O(1)$. In all, it is $O(n^2)$.