The algorithm

```
Iterative-Compute-Opt
M[0] = 0;
S[0] = empty; # array that holds the current optimal solution (pairs made so far) at each step
for i = 1, 2, …, (n choose 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) …].
    set (j,k) = the pair that i represents # e.g. i=8 represents the pair (2,5) when n=6
    if (v_i + p(j,k,n)) > M[i-1]:
        M[i] = v_i + 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[(n choose k)] as the optimal set of all pairs
```

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.

$\mathbf{v_i}$ 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) $\in$ 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 $\notin$ 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 $\forall$ potential pairs (j,k) (compacted into one number i so the arrays can access things), either i $\in O_i$, in which case Most_Pairs = v_i + Most_Pairs(p(i)), or i $\notin O_i$, 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,

$$M[1] = \max(v_1 + M[p(1)], M[1-1])$$

the pseudocode essentially does a max function when using the comparison operator

$$= \max(0 + M[0], M[0])$$
$$= \max(0, 0)$$
$$= 0$$

S[0] = empty
S[1] = S[0]
(because M[1] is not strictly greater than M[0])
S[1] = **empty**

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

Inductive Case

We need to prove M[i] = Most_Pairs[i].
Since we're using strong induction, we can assume that ∀ 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,

$$M[j] = \max(v_j + M[p(j)], M[j-1])$$
$$= \max(v_j + \text{Most\_Pairs}[p(j)], \text{Most\_Pairs}[j-1])$$

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 $O(n^2)$.