

Analysis: Combining Classes

Combining Classes: Analysis

Small dataset

Let us define a function $f(x)$ as the number of students scoring more than or equal to x . We can compute $f(x)$ in $O(N)$ time by iterating through all classes.

Therefore, we can use binary search to find the largest x that satisfies $f(x) \geq K$, which is the score of the K -th highest score. This solution runs in $O(Q \times N \times \log(10^9))$ time.

Large dataset

We can solve the Large dataset by compressing the scores first. That is, we map the existing values of L_i and R_i onto a range $[1, D]$, where D is the number of distinct values in L_i and R_i . After compressing the scores, we can assume that the students' scores are between 1 and $2N$.

For each possible score x , let $\text{cnt}(x)$ be the number of students with score x . We can compute the values for all $\text{cnt}(x)$ in $O(N \log(N))$ using a range-update point-query data structure such as a segment tree. If we compute the suffix sum (similar to [prefix sum](#), but we cumulate the sum backwards) of $\text{cnt}(x)$, we will get the number of students scoring more than or equal to x , which is $f(x)$.

Therefore, just as in our solution for the Small dataset, we can use binary search to find the largest x that satisfies $f(x) \geq K$. However, note that the answer is not exactly x , since we compressed the scores earlier. The exact details of the implementation are left as an exercise.

This solution runs in $O((N + Q) \log N)$ time.