1. perform  $O(\log(nW))$  binary searches on the result value, each check takes O(n) by greedy.

we can also reduce the number of binary searches to  $O(\log n)$ : (there may be a more generalized reference) there are  $O(n^2)$  possible values for the result (sum of interval). if each time we sample a value from the currently remaining possibilities uniformly at random and check that value, the algorithm will terminate in  $O(\log n)$  rounds in expectation. we can use O(n) to sample, by determining the current range by two pointers.

now we reduce the total running time for the binary search part to O(n). after determining the current range in O(n), we can sample in  $O(\log n)$  use this sample range for the next  $\Theta(\log n)$  binary searches, the actual sampling space will shrink, use rejection sampling to repeatedly sample until we get a sample in the actual sample space. if we fail  $O(n^{\epsilon})$  times ( $\epsilon < 1$  is a constant), rebuild the sample space. we need to rebuild O(1) times in expectation.

we can also use the algorithm for selecting the k-th largest element in a sorted matrix in both row and column in deterministic O(n) (but the binary search part in total need  $O(n \log n)$ ). see 378. Kth Smallest Element in a Sorted Matrix.

the running time is  $O(n \log n)$  (independent of m).

- 2. perform each check by exponential search in  $O(m \log \frac{n}{m})$ . the running time is  $O(n + m \log \frac{n}{m} \log n)$ .
- 3. O(n) [1, 2].

## References

- [1] Greg N Frederickson. Optimal algorithms for tree partitioning. In *SODA*, volume 91, pages 168–177, 1991.
- [2] Greg N Frederickson and Samson Zhou. Optimal parametric search for path and tree partitioning. arXiv preprint arXiv:1711.00599, 2017.