

# CSCE 221

## Problem Set 3

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I.

a

Procedure in "printLots.h".

b

Assuming a datastructure that returns `at()` with  $O(1)$ , `printLots()` is  $O(N)$ . `at()` may also iterate through the entire vector up to the needed index in which case `printLots()` is  $O(N^2)$ .

II.

a

Procedure in "intersection.h".

b

Okay.

c

`intersection()` is a for loop of one vector `int*` nested in another for loop for another vector `int*`. Since they are combined in such a way that runs  $N$  times  $N$  times (for each increment of  $N$ , the Procedure will iterate  $N$  more times) the procedure runs at  $O(N^2)$ .

III.

a

Procedure in "josephus.h".

b

The first pass requires  $N$  computations, the second pass requires  $N - \frac{N}{2}$ , the third pass will then require  $(N - \frac{N}{2}) - \frac{N}{4}$  and so on. It seems that the number of computations goes as

$$f(N) = \left( \sum_{i=0}^{I=N} \frac{N}{-2^i} \right)$$

When  $i = N$ , the final element has been found and the  $\Sigma$  terminates. It can be seen that this series is geometric and convergent  $\forall N \cdot M\epsilon$ .  $f(N)$  is a linear combination of computations, the volume of which decays as  $\frac{1}{2^i}$ . Higher order terms are quickly dominated by the first term and as such,  $O(f(N)) = O(N)$ .

c

1) 1: As seen in Fig. 1, garbage collection takes up most of the computation time, however, still grows as  $O(N)$ . Since garbage collection and computation differ only by a constant, as  $N \rightarrow \infty$ , the 3 series are identical. However, all nontrivial datasets are finite and it would still be desirable to minimise the prevalence of the garbage collection step, and complete the computations in a fraction of the time (out most precious resource). An immediate solution would be to move reduce the problem from an iterative problem to a pure math problem, which would bring the computational time down to  $O(1)$  with no need for garbage collection. I suspect that a function of  $N$  and  $M$  will produce the winner without the need for creating a large list. My initial investigation leads me to believe that the winner is chosen with periodicity  $\tau(N, M)$ . I ran the algorithm for this problem to tell me the results and the winner does indeed follow a periodic pattern that grows every reset. Unfortunately, I am unable to follow this to its conclusion due to time constraints, however, the first 10k winners are found in "win.txt".

In every chart, anomalous spikes and noise stem from cpu multitasking and temperature change. Fig. 1 shows an area of low noise and high precision, this is just the result of leaving my computer running at night.

Time per Computation Set

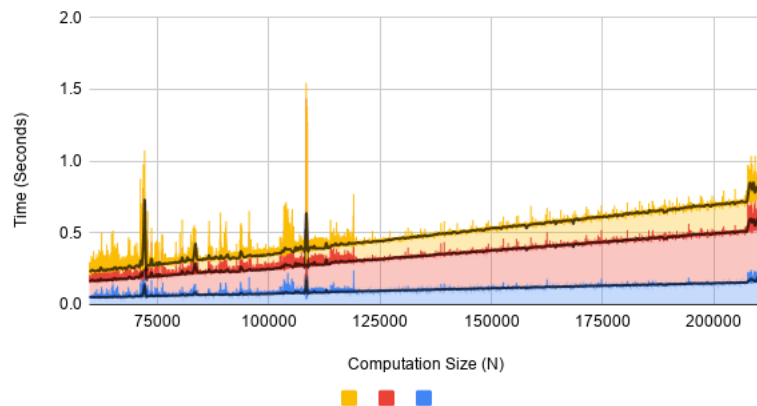


Fig. 1. Yellow: Total computation time, Red: Time taken by data management to allocate and deallocate space for  $N$  people, Blue: Time needed to compute the winner for a given  $N$ .

2) 2: Black trendline represents the moving average of the elimination step, which decays exponentially for a given  $N$ . The sudden spikes are a new list, and each subsequent iteration follows  $f(N)$  as previously discussed. As such, the overall

computation time for each iteration goes as  $O(N)$ . As seen in Fig. 3, the behaviour of the elimination step becomes more apparent for large ( $200k$ ) values of  $N$ . Unfortunately I am not able to plot every collected datapoint ( $7M$ ) with excel or google to show asymptotic behaviour (it seems to be constant but it is suspected that the view is too small to get an accurate depiction). Right now I am trying to figure out an efficient way of plotting such datasets. As it stands, this chart still goes as  $O(N)$ .

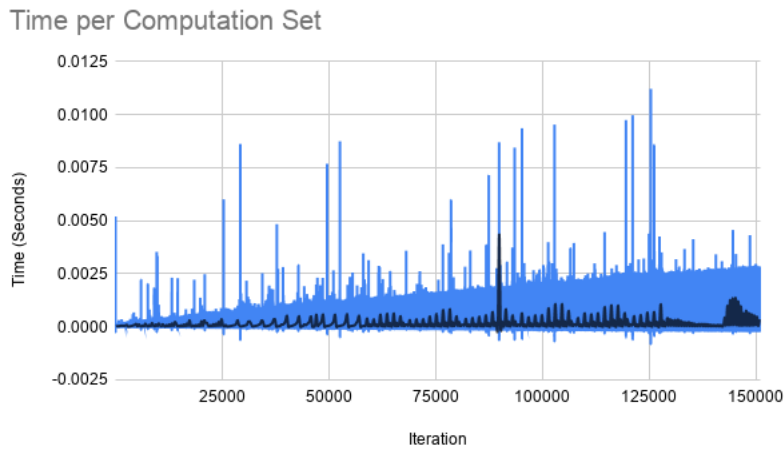


Fig. 2. This chart shows initial growth of iterations as  $N$  grows.

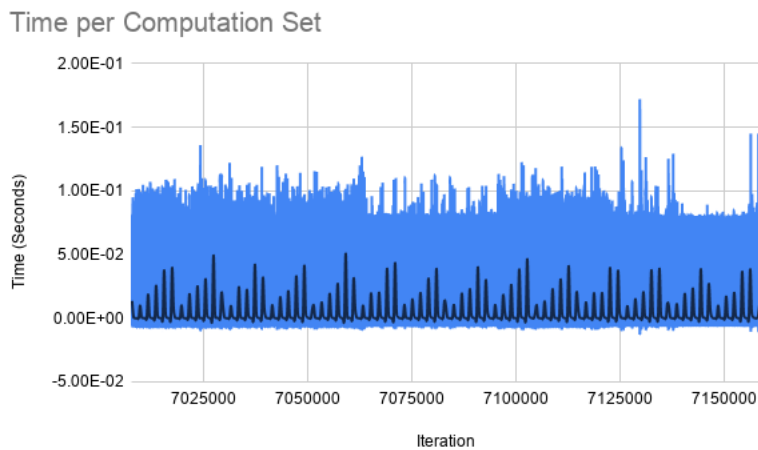


Fig. 3. Elimination time for large values of  $N$  ( $200k$ ).