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The group divided the work up in these ways.

Java File: Benjamin Stewart, Daniel Ty
Algorithms: Benjamin Stewart, Daniel Ty
Graphs: Benjamin Stewart, Joseph Bode
Complexity Analysis: Joseph Bode
Solutions: Daniel Ty, Benjamin Stewart

***DISCLAIMER 1*: We returned the *Values* that make up t instead of the *Indices*.**

Solution: Brute Force

For this algorithm, if the target is 0, we return true and the empty set. If our given array, S , is 0 and our target is nonzero, we return false. We recursively check if $\text{isSumBF}(S, n - 1, \text{target}, \text{indices})$ or $\text{isSumBF}(S, n - 1, \text{target} - S[n - 1], \text{indices})$, return true and indices. Else, we return false.

Solution: Dynamic Programming

Create 2D Boolean Array *subset* of size $[n+1][t+1]$. We then fill out default true and false values. We then traverse through *subset* and check if $\text{subset}[i-1][j] == j \rightarrow \text{subset}[i][j]$. Else we check if $\text{arr}[i-1] > j \rightarrow \text{subset}[i][j] = \text{false}$ else $\rightarrow \text{subset}[i][j] = \text{subset}[i-1][j - \text{arr}[i-1]]$. Finally we return $\text{subset}[n][\text{sum}]$

Solution: Clever

Split the Indices, Compute a table T of all subsets of L that don't exceed t . Return True if any I in $T = t$, else continue. Compute a table W of all subsets of H that yield a subset of S that don't exceed t . Return True if any J in $W = t$, else continue. Sort table W in ascending order. If $S[I] + S[J] = t$, return true, else if $> t$ break to return to the next iteration of outer loop since W is sorted, else continue. If we get through the whole method with no return true, we return false.

Complexity Analysis: Brute Force

In this algorithm, we took every subset of S and computed their sums and then compared it against the target. There are 2^n subsets for every set and each of them hold up to n elements. This is so that each sum evaluation costs $\Theta(n2^n)$ arithmetic operations. Every subset can be represented as a sequence of n bits. This indicates whether each element is or is not in the subset. The sum of each of these fits $\lg t$ bits. This is the space that is required to show t itself. We also know that any sum exceeding this would be disregarded and discarded. Since it is known that

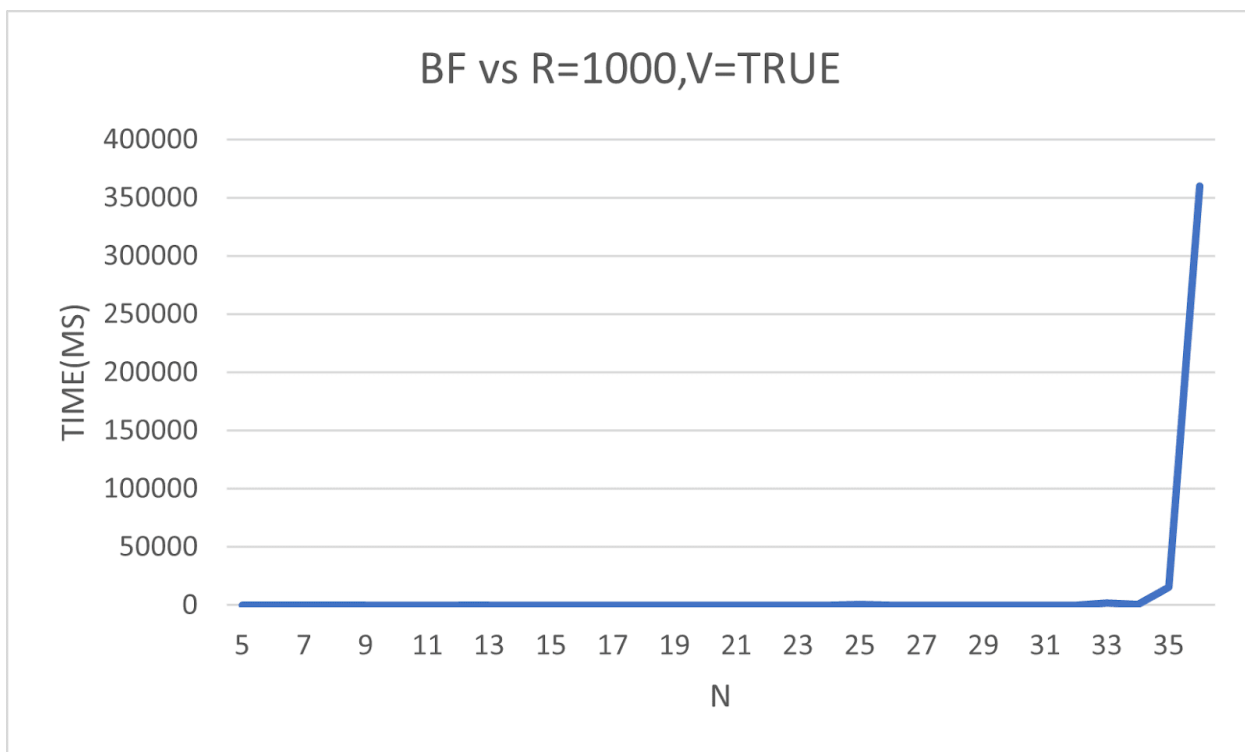
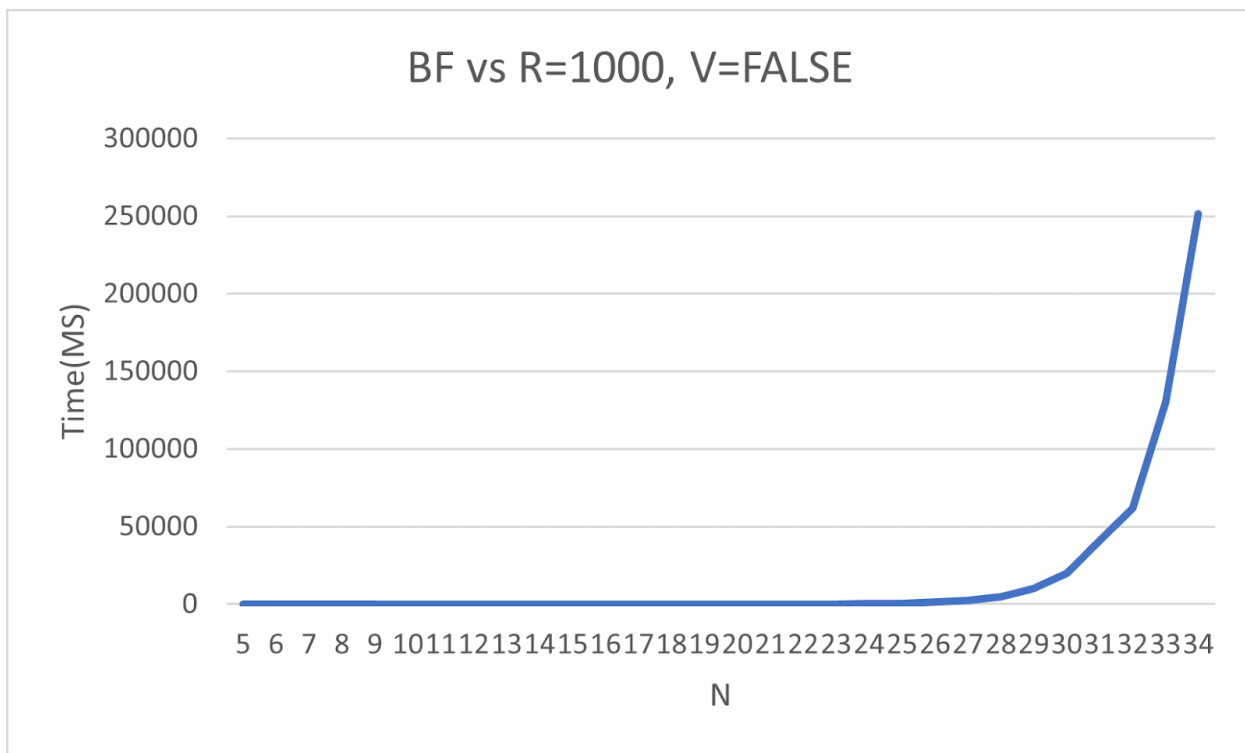
$\lg t \in O(n)$ (the space needed to store t is in the same general area as the space that is needed to store the input), the space complexity is $O(n)$ bits ($O(1)$ elements).

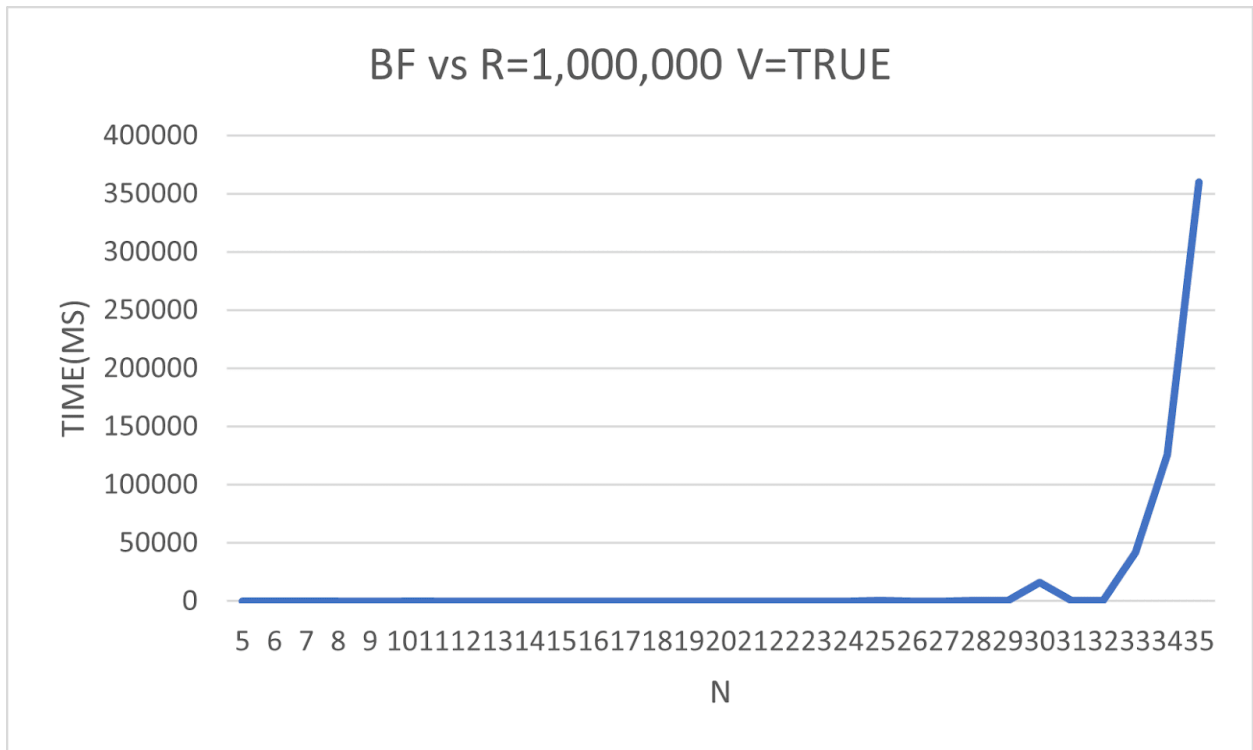
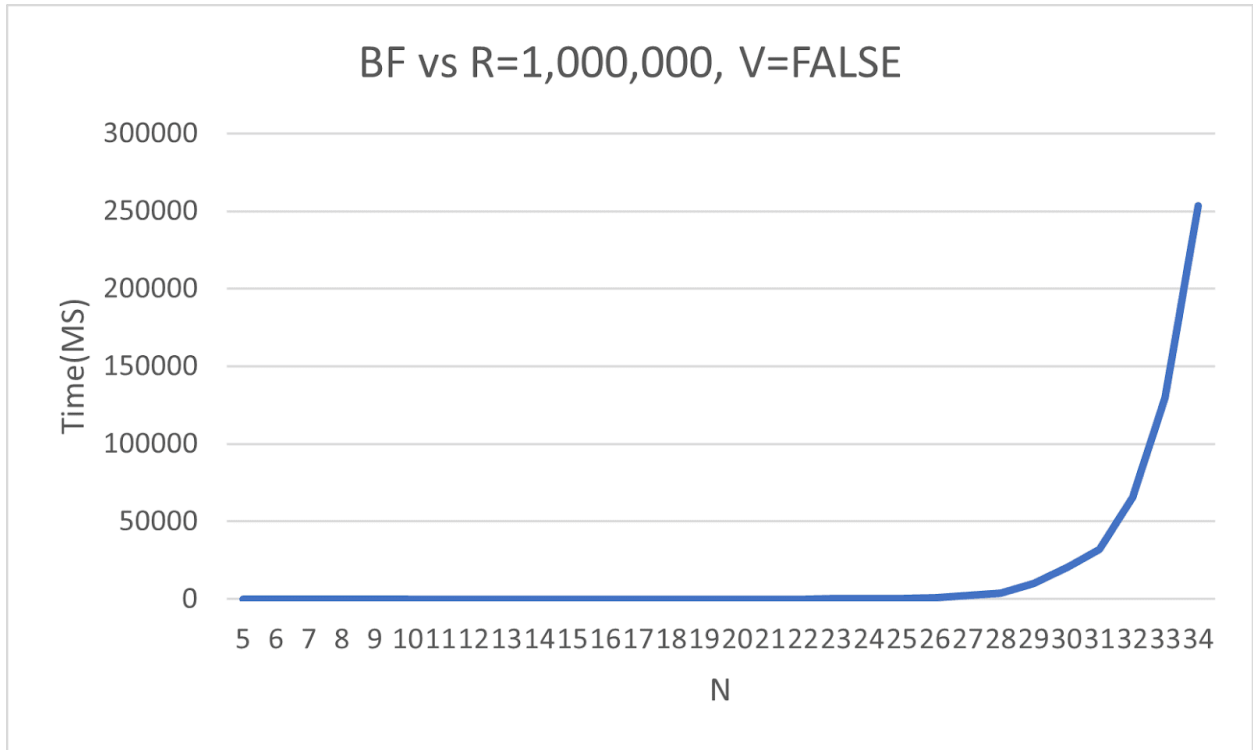
Complexity Analysis: Dynamic Programming

The dynamic programming algorithm is the same thing that was gone over in the course files and lectures. There are two nested loops. One of them being t steps and the other being n steps. The total number of steps (with both of them taking constant time) is nt and the time complexity should be $\Theta(nt)$. We already know from the first analysis that $\lg t \in O(n)$, we can simplify this to $\Theta(n2^n)$ operations. The DP algorithm needs an $n \times t$ matrix of numbers that is large enough to store a value similar to t . That is $O(\lg t) = O(n)$ bits. The final space complexity is $O(nt \lg t) = O(n^2 2^n)$ bits, which is $O(n2^n)$ elements.

Complexity Analysis: Clever Algorithms

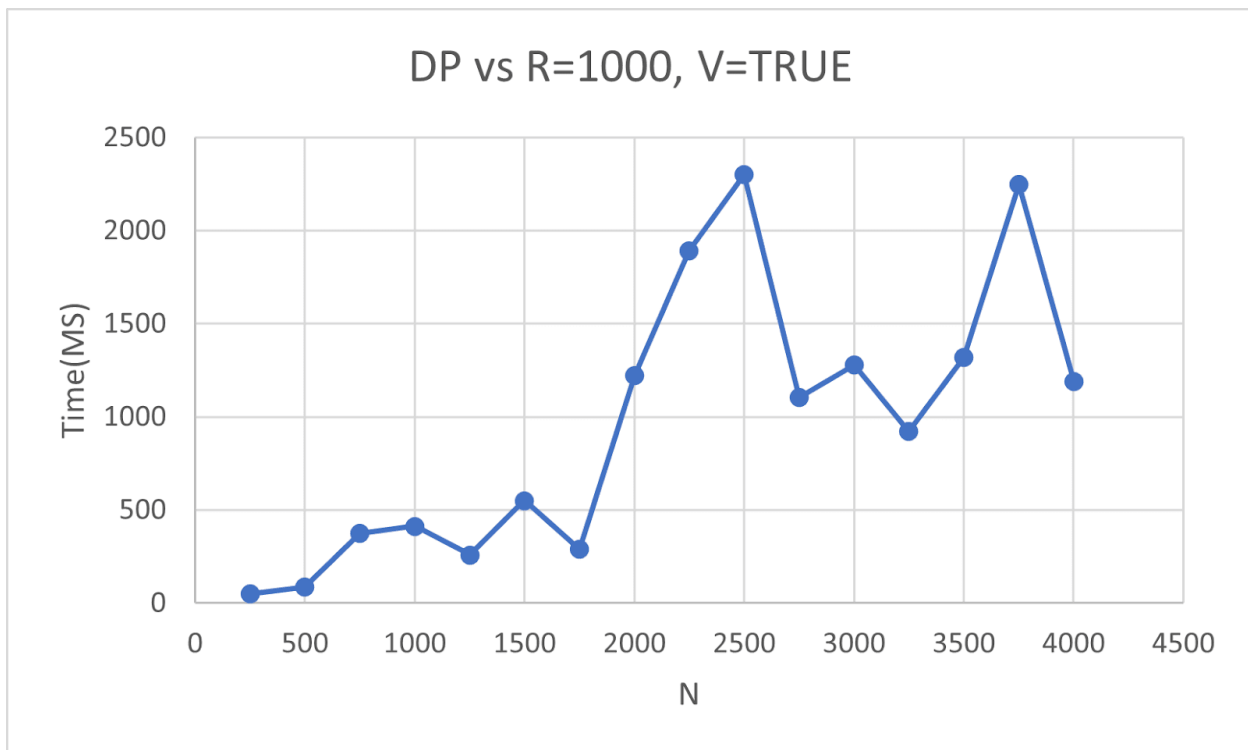
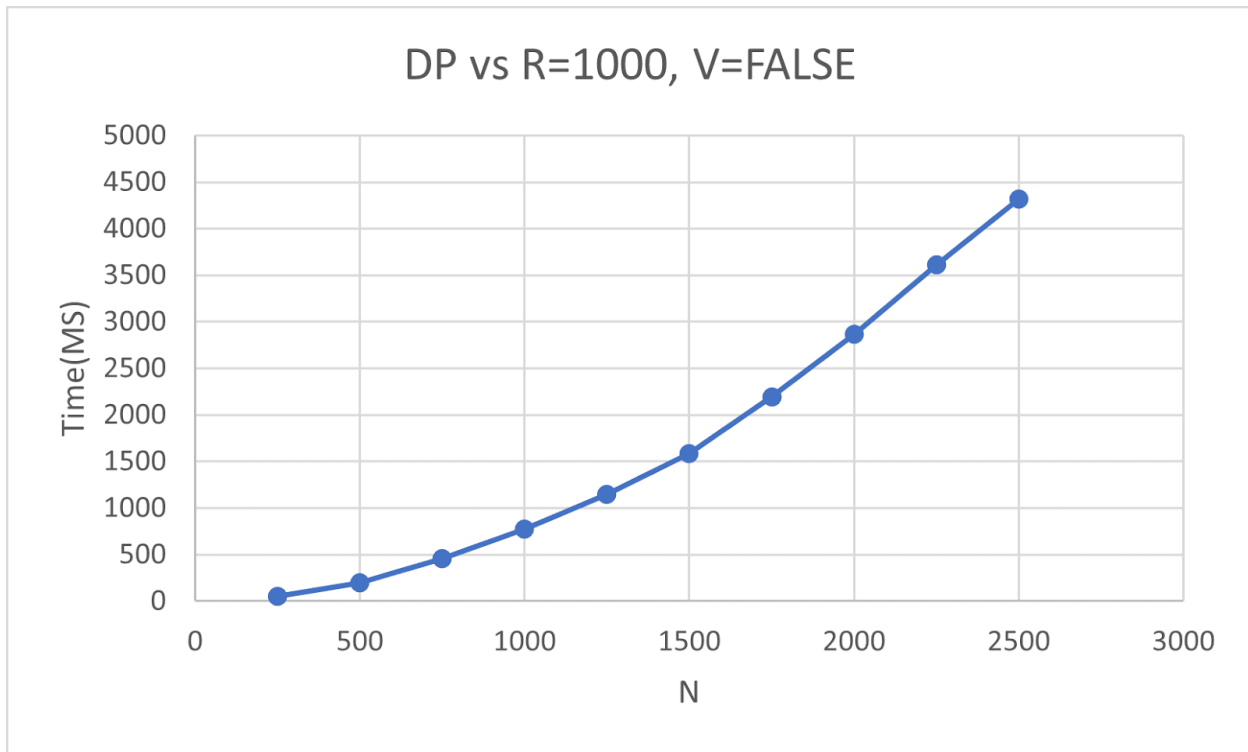
The first step listed in the algorithm from the homework pdf takes at worst $O(n)$ time as we are just splitting a list of size n . We can only note that Step 6 is even easier than this step, Step 6 is just $O(1)$. In Step 2, scanning is performed over all the subsets such that $I \subseteq L$. L contains $n/2$ elements, so there are $2^{n/2}$ subsets. For each one the algorithm computes $\sum_{i \in I} S[i]$ consisting of up to $n/2$ elements. This step does perform $\Theta(n2^{n/2})$ additions. Similar to step 2, Step 3 does $\Theta(n2^{n/2})$ additions. The table size is $N = 2^{n/2}$, step 4 can be done with a $\Theta(N \lg(n))$ sorting method. The cost is $\Theta(n2^{n/2})$ comparisons. Step 5 consists of finding the subset $J \subseteq H$ with maximum weight not passing $t - \text{weight}(I)$ can be done with a binary search on a table W at the cost $\Theta(\lg(n))$. Since the table size is $N = 2^{n/2}$, the cost per a search is $\Theta(n)$ comparisons. Since a search will be performed for each of the $2^{n/2}$ elements of $I \subseteq L$, the final cost of this step is $\Theta(n2^{n/2})$ comparisons. The total time complexity factoring all of this is $\Theta(n2^{n/2})$ arithmetic operations. This clever algorithm does require two tables of subsets of lists of length $n/2$ elements. There exists $2^{n/2}$ such subsets for each. This totals to $2 * 2^{n/2}$ list entries. Every subset can be shown by a sequence of $n/2$ bits. The overall space used is $(n/2) * 2 * 2^{n/2}$, or $O(n2^{n/2})$ or $O(n2^{n/2})$ elements.



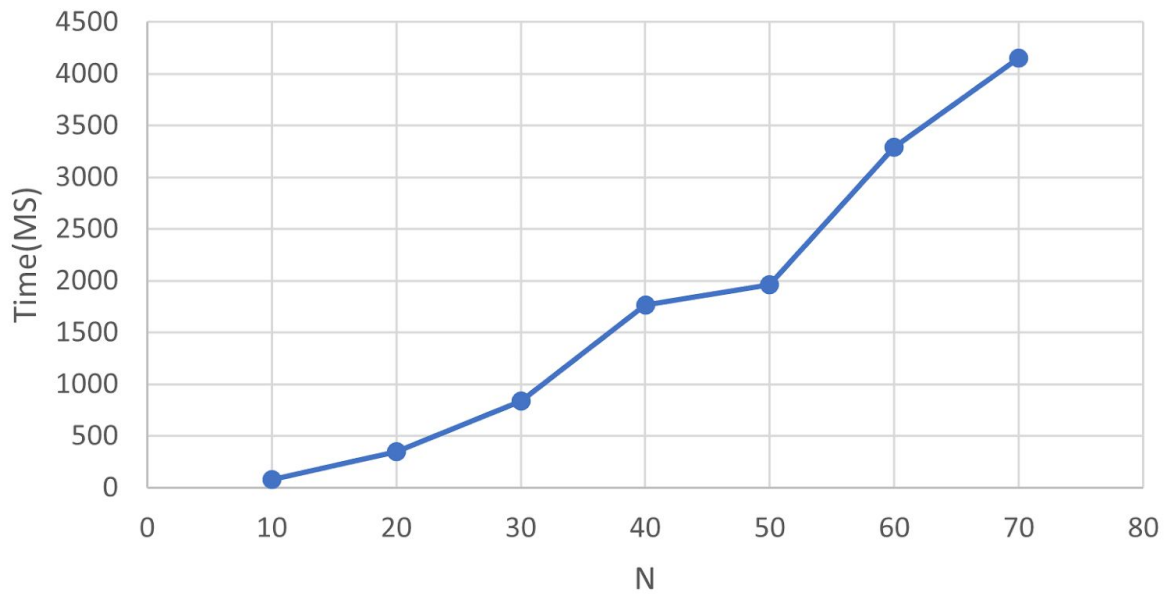


***DISCLAIMER 2*: We stopped at certain values not due to time but due to space.**

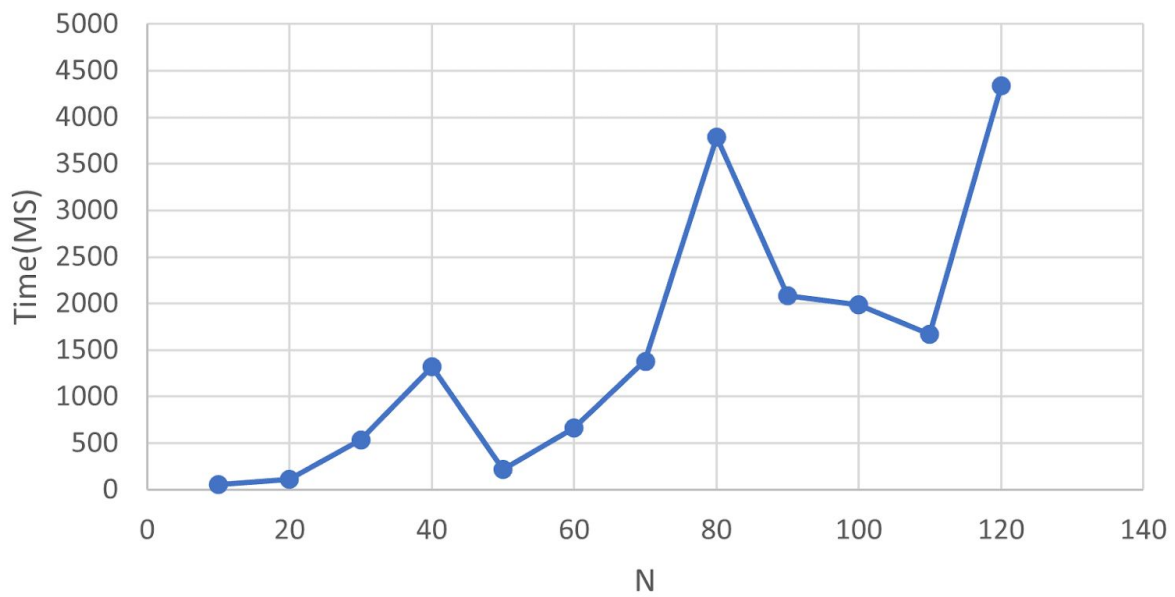
We ran out of memory (Space constraint).

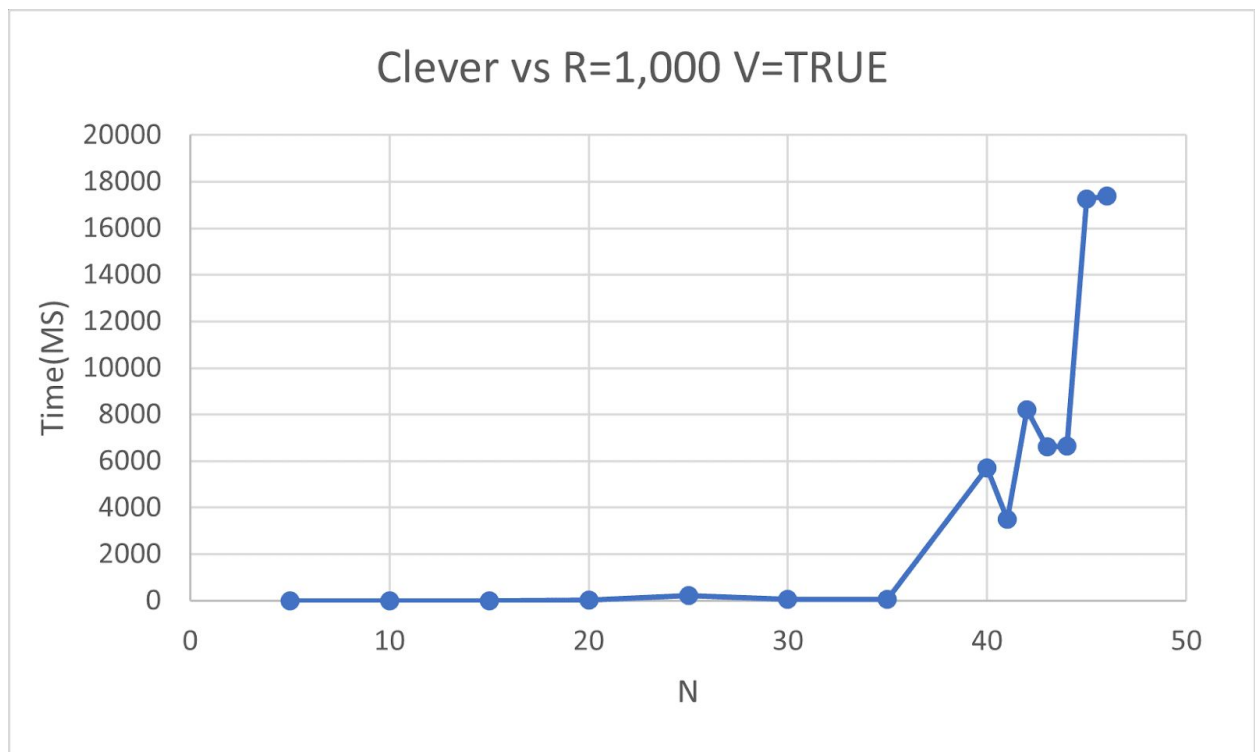
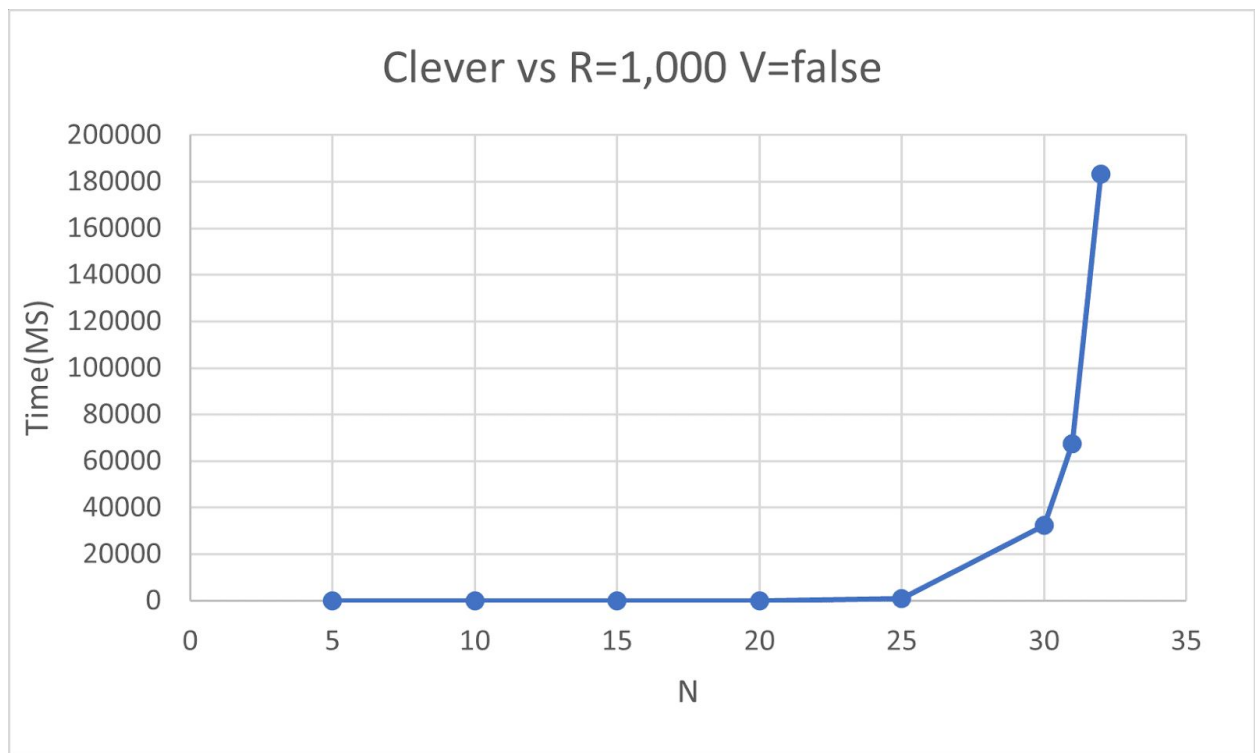


DP vs R=1,000,000, V=FALSE

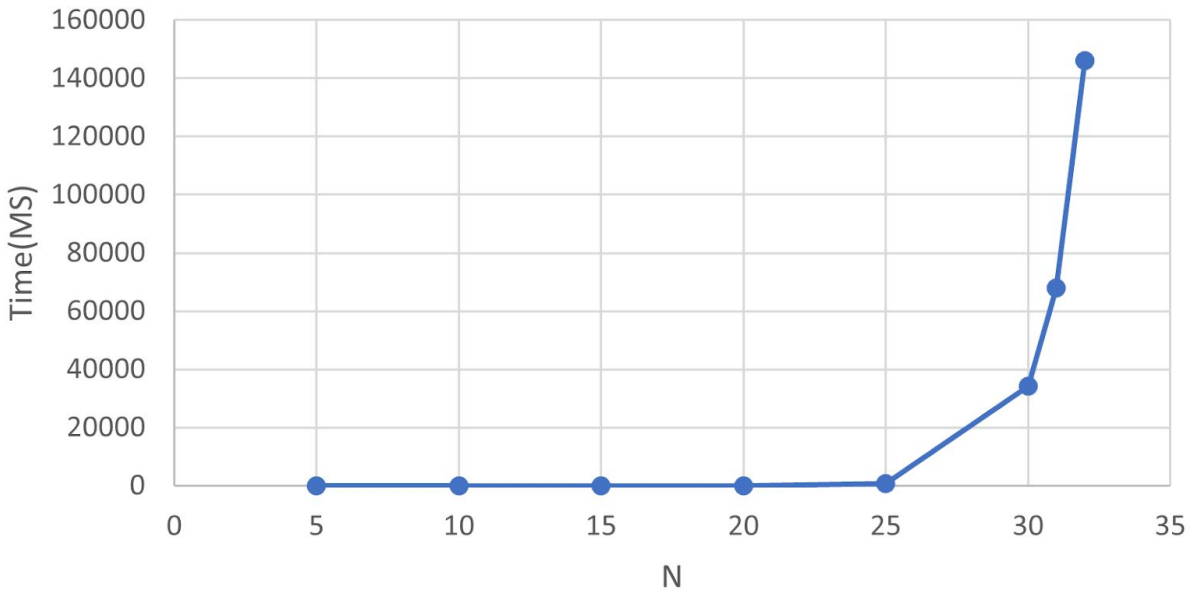


DP vs R=1,000,000, V=TRUE





Clever vs R=1,000,000 V=false



Clever vs R=1,000,000 V=TRUE

