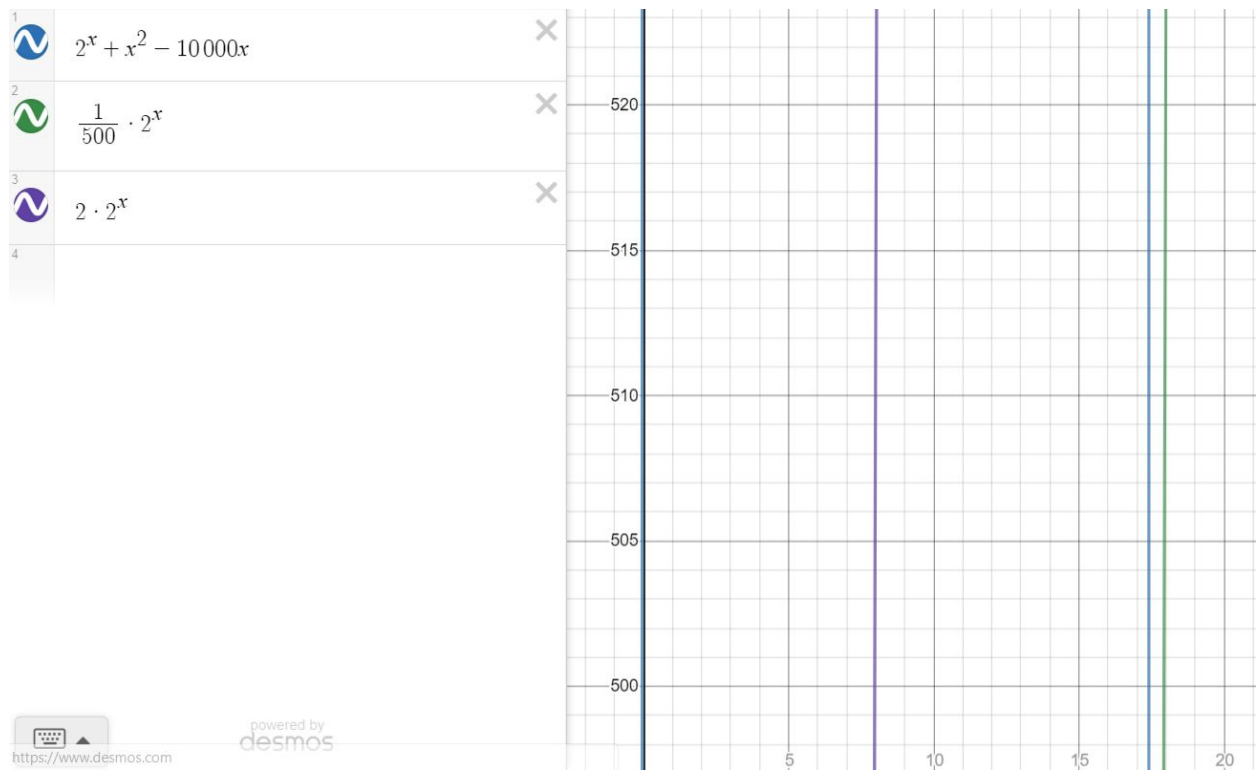


1. See each below.
 - a. The Big-O time for breadth-first search using an adjacency matrix is $O(V^2)$, where V is the number of vertices. This is because since an adjacency matrix is essentially an $V \times V$ matrix of ones and zeros depending on which vertices have edges between them, in order to find the neighbors of each vertex it must go through the entire $V \times V$, which is V^2 , matrix. In addition, processing each node has a Big-O time of $O(V)$, so the total Big-O time is $O(V^2 + V)$, which is simply $O(V^2)$.
 - b. The Big-O time for breadth-first search using an adjacency list is $O(E + V)$, where E is the number of edges and V is the number of vertices. This is because since an adjacency list uses links for each edge, finding the neighbors of a vertex has a time complexity of $O(E)$. In addition, we assume that each vertex (or at least the majority of the vertices) will be processed, which has a Big-O time of $O(V)$.
 - c. The Big-O time for depth-first search using an adjacency matrix is $O(V^2)$, where V is the number of vertices, for the same reason as explained in part a.
 - d. The Big-O time for depth-first search using an adjacency list is $O(E + V)$, where E is the number of edges and V is the number of vertices, for the same reason as explained in part b.
2. See each below.
 - a. The Big-O time for Dijkstra's shortest path algorithm is $O(V^2)$, where V is the number of vertices. This is because Dijkstra's algorithm uses nested for loops, which each repeat $V-1$ as it searches for the nearest neighbor and V times for each vertex, yielding a total complexity of $O(V^2)$.
 - b. Yes, the answer does depend on whether we are using an adjacency matrix or list. The answer given in part a is for an adjacency matrix. Using an adjacency list, however, would have a different Big-O time because the inner loop of traversing the graph would be more similar to breadth-first search, which has a Big-O time of $O(E + V)$ as explained in 1b. You could use a minimum heap to have the neighbors already sorted by the nearest, but heap sort due to the heapifying process has a time of $O(\log V)$. Thus, the complexity of the inner loop of $O(E + V)$, while removing from the heap has complexity $O(\log V)$, so the total Big-O time is $O((E + V) * \log V)$, which simplifies to $O(E \log V)$.
3. See each below.
 - a. By trial and error, I found that the values can be $c_1 = 1/500$, $c_2 = 2$, with $n_0 = 17.5$. Thus, there are constants c_1 and c_2 which, when multiplied by $g(n)$, are upper and lower bounds of $f(n)$. Therefore, the Θ time is $f(n) = \Theta(g(n))$.

b.



4. Quicksort is $O(N^2)$ for the worst case and $\Omega(N \log N)$ for the best case. Thus, there is not a Θ time for quicksort because the worst case and best case scenarios, and the upper and lower bounds, are not the same.