ASSIGNMENT 1: ALGORITHM EFFICIENCY AND SORTING

Question 1.a:

Show that $f(n) = 8n^4 + 5n^3 + 7$ is $O(n^5)$ by specifying appropriate c and no values in Big-O definition.

- f(n) is $O(n^5)$ if $f(n) \le c * n^5$ for some value $n_0 \le n$.
- If $8n^4 + 5n^3 + 7 \le c * n^5$ satisfies; then $\frac{8}{n} + \frac{5}{n^2} + \frac{7}{n^5} \le c$.
- Therefore, choosing $n_0 = 4$ where $n_0 \le n$ and choosing c = 3 holds for our big-Oh notation $8n^4 + 5n^3 + 7$ is $O(n^5)$.

Question 1.b:

Selection Sort:

There will be an index called lastSorted which will start at index -1. We will traverse the array and find the minimum element and swap it with the $(lastSorted + 1)^{th}$ element while updating lastSorted value. Sequence of traversals should stop when the lastSorted element equals to (arraySize - 1).

```
Initial state of array:
{ 22, 8, 49, 25, 18, 30, 20, 15, 35, 27 }
lastSorted = -1;

1st traversal:
    min = 8;
    swap (22, 8);

{ 8, 22, 49, 25, 18, 30, 20, 15, 35, 27 }
lastSorted = 0;

2nd traversal:
    min = 15;
    swap (22, 15);
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{ 8, 15, 49, 25, 18, 30, 20, 22, 35, 27 }
lastSorted = 1;
3<sup>rd</sup> traversal:
        min = 18;
        swap (49, 18);
{ 8, 15, 18, 25, 49, 30, 20, 22, 35, 27 }
lastSorted = 2;
4<sup>th</sup> traversal:
        min = 20;
        swap (25, 20);
{ 8, 15, 18, 20, 25, 30, 49, 22, 35, 27 }
lastSorted = 3;
5<sup>th</sup> traversal:
        min = 22;
        swap (25, 22);
{ 8, 15, 18, 20, 22, 30, 49, 25, 35, 27 }
lastSorted = 4;
6<sup>th</sup> traversal:
        min = 25;
        swap (30, 25);
{ 8, 15, 18, 20, 22, 25, 49, 30, 35, 27 }
lastSorted = 5;
7<sup>th</sup> traversal:
        min = 27;
        swap (49, 27);
{ 8, 15, 18, 20, 22, 25, 27, 30, 35, 49 }
lastSorted = 6;
8<sup>th</sup> traversal:
        min = 30;
        swap (30, 30); // no swap
{ 8, 15, 18, 20, 22, 25, 27, 30, 35, 49 }
lastSorted = 7;
9th traversal:
        min = 35;
        swap (35, 35); // no swap
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```
{ 8, 15, 18, 20, 22, 25, 27, 30, 35, 49 }
lastSorted = 8;
10<sup>th</sup> traversal:
    min = 35;
    swap (35, 35); // no swap

{ 8, 15, 18, 20, 22, 25, 27, 30, 35, 49 }
lastSorted = 9;
(lastSorted == size - 1) is true; no more traversals.
```

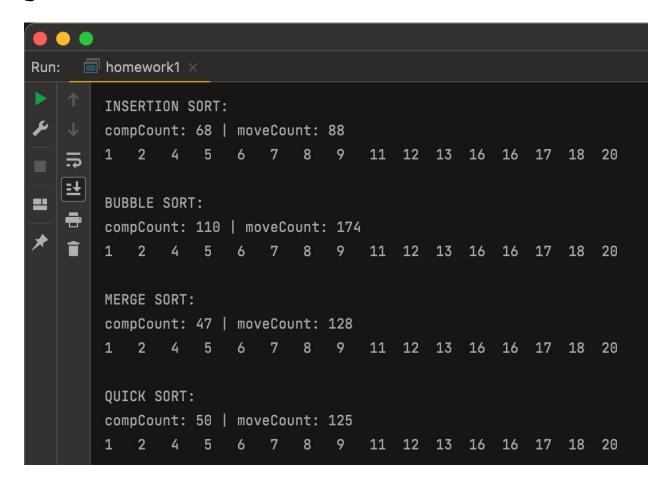
Bubble Sort:

There will be a counter called pass, which counts the number of traversals in order to prevent redundant traversals, and two indexes called cur and next, which points to two successive indexes and traverses the array from 0 to (arraySize – pass). If the element sitting at the cur index is greater than the element at the next index, two elements are swapped. Whenever a traversal without any swap operation is performed, bubble sort algorithm stops.

```
Initial state of array:
{ 22, 8, 49, 25, 18, 30, 20, 15, 35, 27 }
pass = 1;
1<sup>st</sup> traversal:
        swap (22, 8);
        swap (49, 25);
        swap (49,18);
        swap (49, 30);
        swap (49, 20);
        swap (49, 15);
        swap (49, 35);
        swap (49, 27);
{ 8, 22, 25, 18, 30, 20, 15, 35, 27, 49 }
pass = 2;
2<sup>nd</sup> traversal:
        swap (25, 18);
        swap (30, 20);
        swap (30, 15);
        swap (35, 27);
```

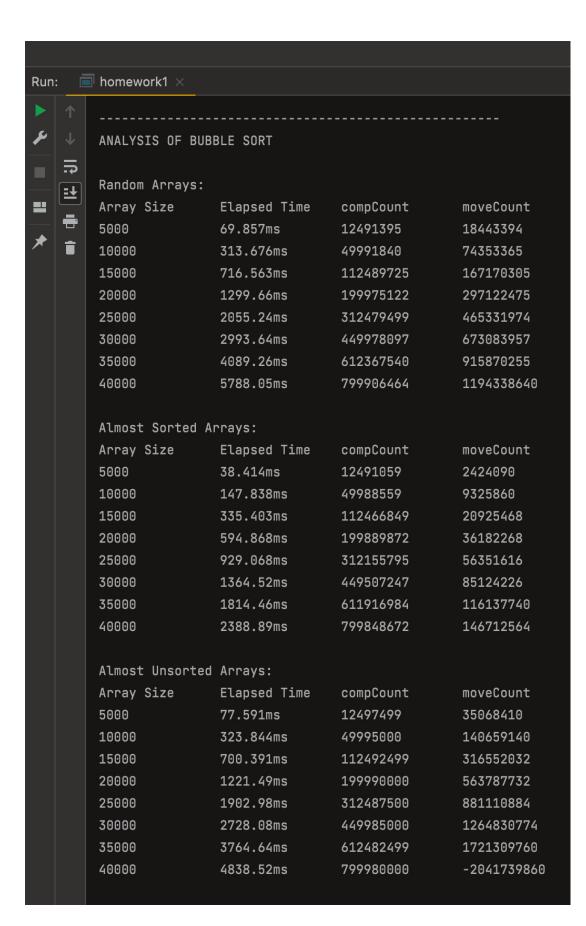
```
{ 8, 22, 18, 25, 20, 15, 30, 27, 35, 49 }
pass = 3;
3<sup>rd</sup> traversal:
        swap (22, 18);
        swap (25, 20);
        swap (25, 15);
        swap (30, 27);
{ 8, 18, 22, 20, 15, 25, 27, 30, 35, 49 }
pass = 4;
4<sup>th</sup> traversal:
        swap (22, 20);
        swap (22, 15);
{ 8, 18, 20, 15, 22, 25, 27, 30, 35, 49 }
pass = 5;
5<sup>th</sup> traversal:
        swap (20, 15);
{ 8, 18, 15, 20, 22, 25, 27, 30, 35, 49 }
pass = 6;
6<sup>th</sup> traversal:
        swap (18, 15);
{ 8, 15, 18, 20, 22, 25, 27, 30, 35, 49 }
pass = 7;
7<sup>th</sup> traversal:
        // no swaps
        sorted = true;
```

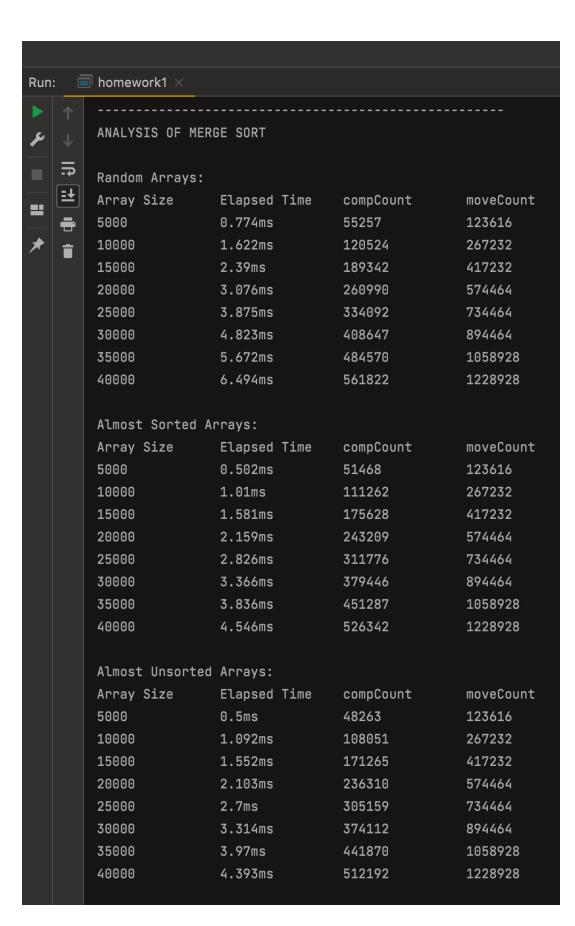
Question 2.c:

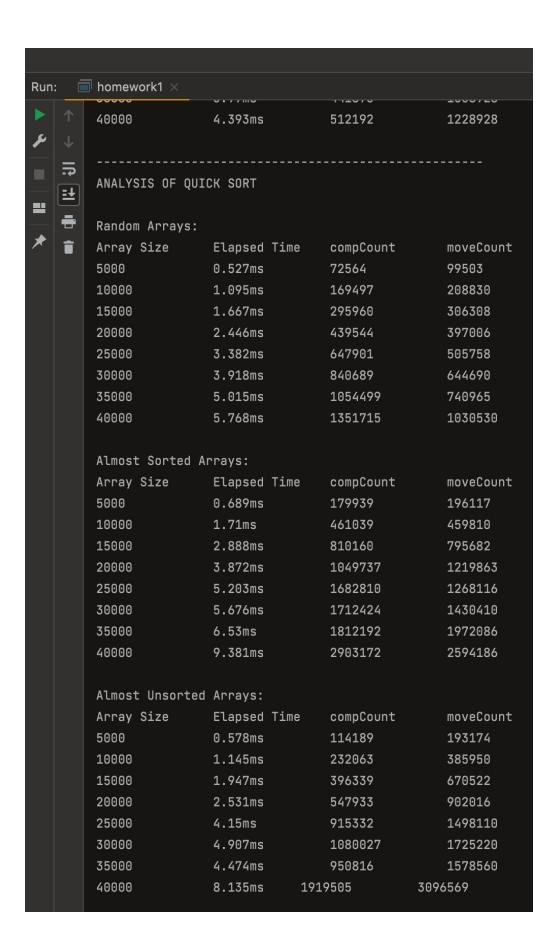


Question 2.d:

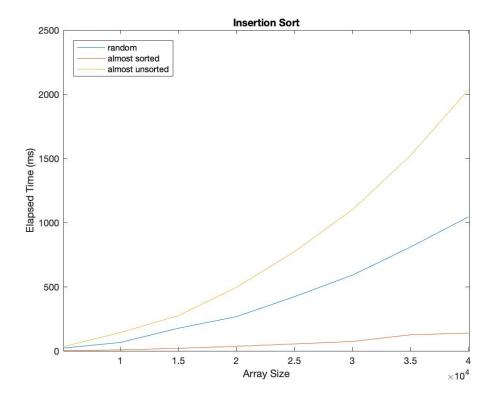
Run: $\overline{\blacksquare}$ homework1 $ imes$					
	↑	ANALYSIS OF INSERTION SORT			
۶					
	I P	Random Arrays:			
==		Array Size	Elapsed Time	compCount	moveCount
_	=	5000	19.472ms	6152786	6157796
*		10000	69.018ms	24794443	24804453
	-	15000	154.631ms	55738423	55753433
		20000	270.208ms	99060813	99080823
		25000	425.089ms	155135646	155160656
		30000	609.248ms	224391307	224421317
		35000	823.086ms	305325073	305360083
		40000	1071.5ms	398152868	398192878
	Almost Sorted Arrays:				
		Array Size	Elapsed Time	compCount	moveCount
		5000	2.225ms	813026	818028
		10000	8.4ms		
		15000		990154	
		20000	32.634ms	12080752	12100754
		25000	51ms 1	8808870	18833870
		30000	75.972ms	28404740	28434740
		35000	112.371ms	38747576	38782578
		40000	131.403ms	48944186	48984186
		Almost Unsorted Arrays:			
		Array Size	Elapsed Time	compCount	moveCount
		5000	31.612ms	11694440	11699468
		10000	133.641ms	46896333	46906378
		15000	283.771ms	105532309	105547342
		20000	502.02ms	187949210	187969242
		25000	793.065ms	293728574	293753626
		30000	1132.41ms	421640223	421670256
		35000	1541.51ms	573804890	573839918
		40000	1945.73ms	751115761	751155810

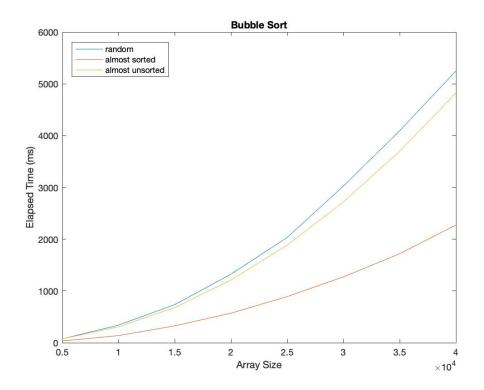


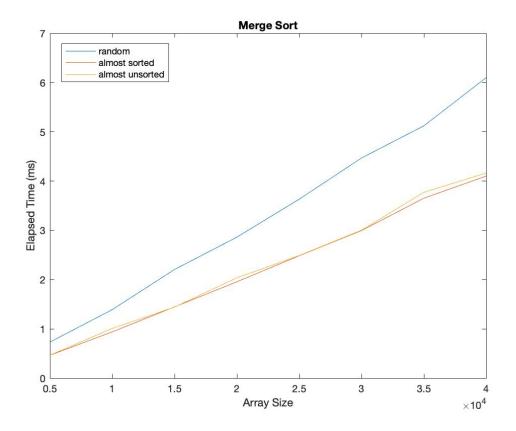


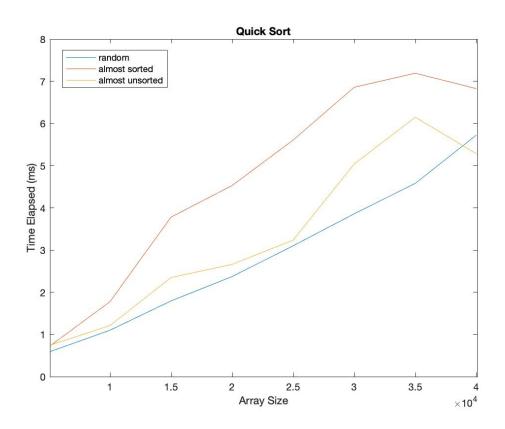


Question 3:









Question 3. Conclusion:

Ideally, **insertion sort** should be $\Omega(n)$ for best case. Our experimental result is matching the theoretical value as shown in the first graph with a red line which is nearly linear. The theoretical complexity in average case is $\theta(n^2)$, which roughly satisfies the experimental result as shown with the blue line. The reason behind the line being a bit more linear compared to a n^2 parabola is the lack of sample size, or size of the arrays in other words. If this experiment was conducted with a sample size of 20 for instance – maximum array size being 100,000 – the n^2 parabola would have been more visible. When it comes to worst case which is represented by the almost unsorted array line which is yellow, it is much more accurate to the theoretical complexity of $O(n^2)$ compared to best and average case, since insertion sorting an almost unsorted array requires significantly more work, hence more time. The ideal complexity is n^2 for average and worst case because there is a nested for loop in the insertion sort algorithm. On the other hand, if we have a completely sorted array, only one iteration is sufficient.

For **bubble sort**, both worst and average case should be $O(n^2)$ and $\theta(n^2)$ respectively. As shown in the graph above, blue representing the average case and yellow representing the worst case are in fact increasing approximately quadratically, meeting our expectations. When it comes to the best case, we expect the complexity to be $\Omega(n)$. Although the red line on the graph is not absolutely linear, one should keep in mind that the sample array is not completely sorted but the 10% of it was altered, meaning we could not expect the line to be perfectly linear, but a bit more parabola-like. The reason behind worst- and best-case complexities generating different lines is the bubble sort algorithm, which has a nested for loops. In the best case only one traversal is sufficient, however in the case of the worst case, we should make sure that the biggest value of the unsorted part is transferred to the end of the array, hence the sorted part. This operation requires n traversals at most.

Merge sort algorithm should have the same time complexities O(n * log(n)), $\theta(n * log(n))$, and $\Omega(n * log(n))$ for worst, average, and best cases. y = n(log(n)) is nearly a linear line, which justifies the graphs above, all of which are practically linear. The time required for the merge sort algorithm, which outperforms bubble and insertion sort algorithms is significantly less due to its efficiency. Its complexity is O(* log(n)), because *mergeSort* function recursively calls itself for both halves, dividing itself to two sections every time it is called. This corresponds to log(n) times for number of *mergeSort* function calls. There is a helper function, called *merge*, which iterates through every sub array it was given, corresponding to n number of iterations.

The case for **quick sort** is more complicated. We should expect time complexities of $\Omega(n * \log(n))$ and $\theta(n * \log(n))$ for best and average cases, whereas $O(n^2)$ for worst case. This is because *quickSort* function again recursively calls itself, just as *mergeSort*. However, this time the hard work is done before the recursive call, meaning the *partition* function is called before the recursive call. For the worst case, the pivot input could be the greatest or the least element of the array every time it was recursively called. This is possible with a completely sorted or completely unsorted array. Although, the graph is right with showing almost sorted and unsorted arrays taking more time than a random array; the red and yellow lines does not match our time complexity expectations completely. The justification for it could be the lack of sample size, therefore the arrays size. The sorting operation only take a few milliseconds which is not adequate for an accurate graph. For the best case, the value of the pivot index should be as close as it can be to the median of the array for each recursive call which would result as $(\log(n))$ calls. Again, partition is iterating over the array which corresponds to n. Blue line which represents average case is more accurate since it is close to a $(n * \log(n))$ line.