

Problem 3.1 Asymptotic Analysis (8 points) Considering the following pairs of functions f and g , show for each pair whether or not it belongs to each of the relations $f \in \Theta(g)$, $f \in O(g)$, $f \in o(g)$, $f \in \Omega(g)$, $f \in \omega(g)$, $g \in \Theta(f)$, $g \in O(f)$, $g \in o(f)$, $g \in \Omega(f)$, or $g \in \omega(f)$.

(a) (2 points) $f(n) = 9n$ and $g(n) = 5n^3$

using computation with limits:

$$\lim_{n \rightarrow +\infty} \frac{9n}{5n^3} = \lim_{n \rightarrow +\infty} \frac{9}{5n^2} = 0$$

$f \in \Theta(g)$	No
$f \in O(g)$,	Yes
$f \in o(g)$,	Yes
$f \in \Omega(g)$	No
$f \in \omega(g)$	No

now, the other way around:

$$\lim_{n \rightarrow +\infty} \frac{5n^3}{9n} = \lim_{n \rightarrow +\infty} \frac{5n^2}{9} = \infty$$

$g \in \Theta(f)$	No
$g \in O(f)$,	No
$g \in o(f)$,	No
$g \in \Omega(f)$	Yes
$g \in \omega(f)$	Yes

(b) (2 points) $f(n) = 9n^{0.8} + 2n^{0.3} + 14 \log n$ and $g(n) = \sqrt{n}$,

using computation with limits:

$$\lim_{n \rightarrow +\infty} \frac{9n^{0.8} + 2n^{0.3} + 14 \log n}{n^{0.5}} = \lim_{n \rightarrow +\infty} \frac{9n^{0.8}}{n^{0.5}} + \lim_{n \rightarrow +\infty} \frac{2n^{0.3}}{n^{0.5}} + \lim_{n \rightarrow +\infty} \frac{14 \log n}{n^{0.5}}$$

Now, evaluating each term separately:

$$\lim_{n \rightarrow +\infty} \frac{14 \log n}{n^{0.5}} = 0,$$

Since : \sqrt{n} grows faster than $\log n$. To “formally” prove this, L’Hopital’s rule can be used: Notice that $14 \log_2(n) = \frac{14 \ln(n)}{\ln(2)}$, so that $f'(n) = \frac{1}{n \ln(2)}$. Also, $g'(n) = \frac{1}{2\sqrt{n}}$, such that:

$\lim_{n \rightarrow +\infty} \frac{14 \log n}{n^{0.5}} = \lim_{n \rightarrow +\infty} \frac{2\sqrt{n}}{n \ln(2)}$. linear functions grow faster than square roots, so this evaluates to zero. L’Hopital’s rule can be applied again to show this:

$$\text{Derivative of top expression} = \frac{1}{\sqrt{n}}$$

Derivative of bottom expression: $\ln(2)$

thus,

$$\lim_{n \rightarrow +\infty} \frac{2\sqrt{n}}{n \ln(2)} = \lim_{n \rightarrow +\infty} \frac{\frac{1}{\sqrt{n}}}{\ln(2)} = \lim_{n \rightarrow +\infty} \frac{1}{\sqrt{n} \ln(2)} = 0$$

Next term:

$$\lim_{n \rightarrow +\infty} \frac{2n^{0.3}}{n^{0.5}} = \lim_{n \rightarrow +\infty} \frac{2}{n^{0.2}} = 0$$

$$\lim_{n \rightarrow +\infty} \frac{9n^{0.8}}{n^{0.5}} = \lim_{n \rightarrow +\infty} 9n^{0.3} = \infty$$

$$\lim_{n \rightarrow +\infty} \frac{\text{thus, } 9n^{0.8} + 2n^{0.3} + 14 \log n}{n^{0.5}} = \infty$$

$f \in \Theta(g)$	No
$f \in O(g),$	No
$f \in o(g),$	No
$f \in \Omega(g)$	Yes
$f \in \omega(g)$	Yes

now, the other way around

$\lim_{n \rightarrow +\infty} \frac{n^{0.5}}{9n^{0.8} + 2n^{0.3} + 14 \log n} = 0$ as shown above, the term with the greatest order of growth is $9n^{0.8}$, which evaluates the limit to zero as this term is in the denominator.

$g \in \Theta(f)$	No
$g \in O(f),$	Yes
$g \in o(f),$	Yes
$g \in \Omega(f)$	No
$g \in \omega(f)$	No

(c) (2 points) $f(n) = n^2 / \log n$ and $g(n) = n \log n$,

$$\lim_{n \rightarrow +\infty} \frac{\frac{n^2}{\log n}}{n \log n} = \lim_{n \rightarrow +\infty} \frac{n}{(\log n)^2}$$

again, we apply L'Hopital's rule to compute this limit

$$\text{derivative of bottom expression} = \frac{2 \ln n}{n \ln 2}$$

$$\lim_{n \rightarrow +\infty} \frac{n}{(\log n)^2} = \lim_{n \rightarrow +\infty} \frac{n \ln 2}{2 \ln n}$$

now we use L'Hopital's rule to prove that linear growth is greater than logarithmic as n approaches infinity:

$$\text{derivative of bottom expression} = \frac{2}{n}$$

$$\text{derivative of top expression} = \ln 2$$

$$\lim_{n \rightarrow +\infty} \frac{n \ln 2}{2 \ln n} = \lim_{n \rightarrow +\infty} \frac{n \ln 2}{2} = \infty$$

$f \in \Theta(g)$	No
$f \in O(g),$	No
$f \in o(g),$	No
$f \in \Omega(g)$	Yes
$f \in \omega(g)$	Yes

Now, the other way around:

$$\lim_{n \rightarrow +\infty} \frac{n(\log n)^2}{n^2} = 0, \text{ above, it is already shown that } n^2 \text{ grows at a greater rate than } n(\log n)^2 \text{ as } n \rightarrow \infty.$$

$g \in \Theta(f)$	No
$g \in O(f),$	Yes
$g \in o(f),$	Yes
$g \in \Omega(f)$	No
$g \in \omega(f)$	No

(d) (2 points) $f(n) = (\log(3n))^3$ and $g(n) = 9 \log n$.

$$\lim_{n \rightarrow +\infty} \frac{\left(\frac{\ln(3n)}{\ln 2}\right)^3}{9 \frac{\ln(n)}{\ln(2)}}$$

again, we apply L'Hopital's rule:

Derivative of top expression:

$$f'(x) = \frac{3 \left(\frac{\ln(3n)}{\ln 2}\right)^2}{n}$$

Derivative of bottom expression:

$$g'(x) = \frac{9}{n \ln(2)}$$

Dividing these terms, we get:

$$\lim_{n \rightarrow +\infty} \frac{\left(\frac{\ln(3n)}{\ln 2}\right)^3}{9 \frac{\ln(n)}{\ln(2)}} = \lim_{n \rightarrow +\infty} \frac{3n \ln(2) \left(\frac{\ln(3n)}{\ln(2)}\right)^2}{9n} = \lim_{n \rightarrow +\infty} \frac{3 \ln(2) \left(\frac{\ln(3n)}{\ln(2)}\right)^2}{9} = \infty$$

From this, we can also derive that $(\log(3N))^3$ grows at a greater rate than $9 \log n$ as $n \rightarrow \infty$

$f \in \Theta(g)$	No
$f \in O(g),$	No
$f \in o(g),$	No
$f \in \Omega(g)$	Yes
$f \in \omega(g)$	Yes

Now, the other way around:

$$\lim_{n \rightarrow +\infty} \frac{9 \frac{\ln(n)}{\ln(2)}}{\left(\frac{\ln(3n)}{\ln 2}\right)^3} = 0, \text{ the derivations above showed that } (\log(3n))^3 \text{ grows at a greater rate than}$$

$9 \log n$ as $n \rightarrow \infty$, so the expression evaluates to zero.

$g \in \Theta(f)$	No
$g \in O(f),$	Yes
$g \in o(f),$	Yes
$g \in \Omega(f)$	No
$g \in \omega(f)$	No

Problem 3.2

b) Loop invariant: at any time, the sub-array to the left of the current element is sorted. Consider the following trace of an execution of Selection Sort on the array {8,6,7,0,4,9,2,1,5,3}

```
*0 6 7 8 4 9 2 1 5 3
0 *1 7 8 4 9 2 6 5 3
0 1 *2 8 4 9 7 6 5 3
0 1 2 *3 4 9 7 6 5 8
0 1 2 3 *4 9 7 6 5 8
0 1 2 3 4 *5 7 6 9 8
0 1 2 3 4 5 *6 7 9 8
0 1 2 3 4 5 6 *7 9 8
0 1 2 3 4 5 6 7 *8 9
```

Where the * asterisk denotes the beginning of the leftmost sub-array. As It can be seen, the leftmost sub-array is always sorted in each iteration of the outermost for loop of Selection Sort.

Some special considerations are that:

- At iteration zero, the leftmost sub-array is of size 1, so it is trivially already sorted
- During any iteration, the leftmost sub-array is sorted by swapping positions with the smallest element in the right sub-array. This means that for the final iteration, since all of the left sub-array is sorted, then the final element of the complete array is also sorted, so the array is sorted.

c) For full code, see: testSelectionSort.cpp

The case with the least swaps is an array that is already sorted in increasing order: This is because no swaps need to be done: the following c++ code shows how to make an n-sized array with the least swaps for Selection Sort. (Note that the functions are void since they are snippets of the source code)

```
void best_Case(int n){
    int best_case[n];
    for (int i =0; i < n; i++){
        best_case[i] = i;
    }
}
```

The case with the most swaps is an array with the greatest element in the first position, followed by the rest of the elements in increasing order. This gives a total of n-1 swaps for any input size n. So for an input of size 10, the worst case is {10, 1, 2, 3, 4, 5, 6, 7, 8, 9}. This is the worst case because the greatest number will always be the comparison index for swapping, and thus will always swap until the algorithm terminates.

```
void worst_Case(int n){
    int worst_case[n];
    worst_case[0] = n;
    for (int i = 1; i < n; i++){
        worst_case[i] = i;
    }
}
```

The “average case” is made using a random number generator to make the numbers.

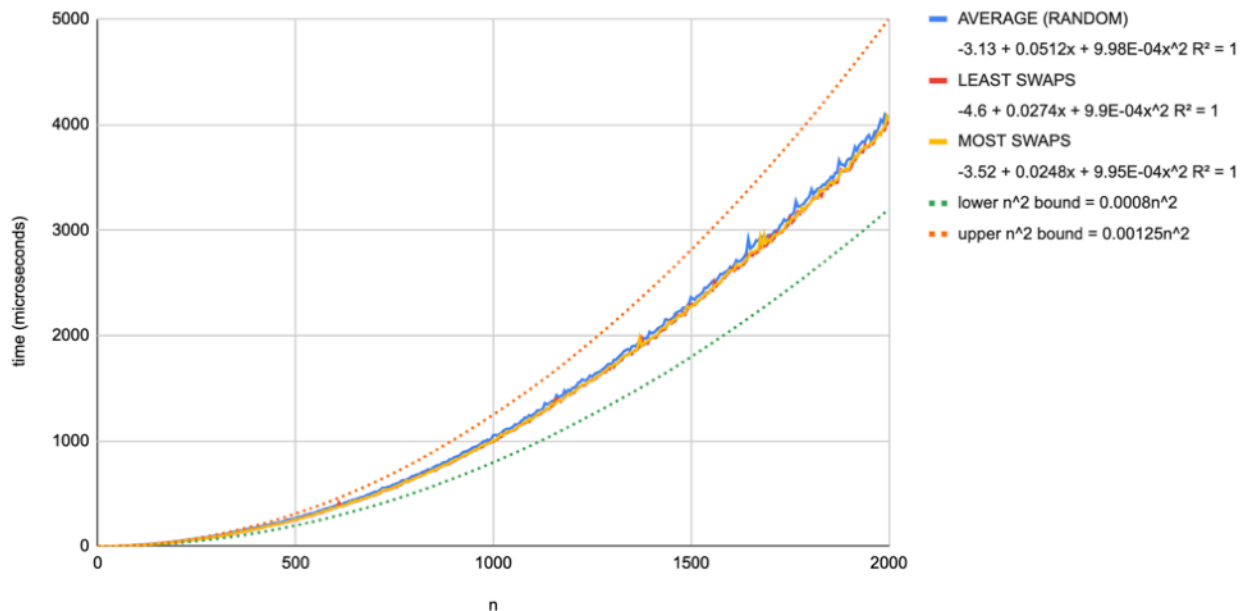
```
# include <ctime>
void average_Case(int n){
    srand(time(NULL));
    int average_case[n];
    for(int i =0; i < n; i++){
        average_case[i] = rand();
    }
}
```

d) The following results were obtained using the code from `testSelectionSort.cpp`, running the following parameters:

- Maximum number of elements (n) = 2000
- Increasing intervals of n from 0 to maximum = 5
- Number of repetitions of the experiment to calculate the mean: 100

The mean results (called `final_x.txt`) were then passed to google sheets and graphed.

Average time to sort n -sized array using SelectionSort



e) from the visual perspective it is easy to see that cases A, B, and the average case are all elements of $O(n^2)$. Furthermore, all three of the cases seem to have a very similar time complexity. To be more concise, let cases A, B, and the average case be defined as a function $f(n)$. In more mathematical terms:

$$0 \leq 0.0008n^2 \leq f(n) \leq 0.00125n^2, \forall n > n_0$$

Therefore, we can also conclude:

$$f(n) = \Theta(n^2)$$