```
1. R-4.8, p. 182, provide explanation
Order the following functions by asymptotic growth rate.
                 2^{10} 2^{\log(n)} 3 n+100 log(n) 4n 2^n n^2 +10n n^3
4 n log(n)+2n
                                                                             n log(n)
(BEST)
2<sup>10</sup>
       //constant
3 n + 100 log(n)
       //Big O - n Same as 4n. Both are linear. log(n) gets dropped since its a lower rate
4n
       //Linear graph. It's a constant growth graph.
n log(n)
       //linearithmic graph. Combines linear and log(n) graphs.
4n \log(n) + 2n
       //Big O = n \log(n). It is the same time complexity as the one above. In comparison, just
       because it has more constants, it can be worse.
n<sup>2</sup> +10n
       //Big O = n^2. Its quadratic graph. The growth rate is bigger than ones above.
n^3
       //Higher growth rate than quadratic.
2<sup>log(n)</sup>
       //Exponential growth.
2<sup>n</sup>
(WORST)
2. R-4.13, p. 182, provide explanation
Give a big-Oh characterization, in terms of n, of the running time of the example5 method
shown in Code Fragment 4.12.
There are 3 for loops nested within each other. Lets label the loops, A, B, and C. Loop C is
within loop B and look B is within loop A. There is a comparison that runs inside loop C. This
comparisons runs n * j(The count of times Loop B has already ran). This is linear. Loop B runs
n<sup>2</sup> times. Loop A runs n times.
(line 39) A = n
(line 41) B = n^2
(line 42) C = n^*(1+2+3...(j))
Conditional if statement = n^2 * (n*(1+2+3...(j)))
```

```
3. R-4.19, p. 184
Show that O(\max\{f(n),g(n)\}) = O(f(n) + g(n)).
```

Big $O = n^3$

$$C * max{ f(n),g(n)} = f(n) +g(n)$$

$$f(n) <= \max \{ \ f(n), g(n) \} \quad \text{and} \quad g(n) <= \max \{ \ f(n), g(n) \} \quad \text{when } n >= n_0$$

$$f(n) + g(n) <= 2 * \max \{ \ f(n), g(n) \} \quad \text{when } n >= n_0$$

$$So \ C = 2 \ n >= n_0$$

$$So \ f(n) + g(n) <= \max \{ \ f(n), g(n) \} \quad \text{(remove constants)}$$

4. Consider $f(n) = 4n^2 + 3n - 1$, mathematically show that f(n) is $O(n^2)$, $\Omega(n^2)$, and $O(n^2)$.

1)
$$O(n^2) = 4n^2 + 3n-1 >= cn^2 C > 0$$
 and $n >= n_0 n_0 = 1 C = 1$

2)
$$3n-1 >= cn^2 - 4n^2$$

3)
$$3n-1 >= n^2(c-4)$$
 4) $3n-1 >= n^2(c-4)$ 5) $\frac{f(n)}{g(n)} <= \frac{C g(n)}{g(n)}$

6)
$$\frac{3n-1}{n^2} >= \frac{n^2(c-4)}{n^2}$$
 7) $\frac{3n-1}{n^2} >= c-4$

Input n
$$3-1+4 \ge c$$
 $3-1+4 \ge 6$ so C = 6

So for all N when greater then n_0 , $f(n) \le Og(n)$

$$O(n^2)$$
 $C = 6 n_0 = 1$

$$\Omega$$
 (n²) C = 1 n₀ = 1

$$\theta$$
 (n²) C = 6 n₀ = 1

5. For finding an item in a sorted array, consider "ternary search," which is similar to binary search. It compares array elements at two locations and eliminates 2/3 of the array. To analyze the number of comparisons, the recurrence equations are T(n) = 2 + T(n/3), T(2) = 2, and T(1) = 1, where n is the size of the array. Explain why the equations characterize "tertiary search" and solve for T(n).

$$T(n) = 2 + T(n/3), T(2) = 2, \text{ and } T(1) = 1,$$

$$T(n) = 2 + (1 + T(n/9))$$

$$T(n) = 3 + (1 + T(n/27))$$

$$T(n) = x + T(n/3^x)$$
 base case $n/2^x = 1$ so $2^x = n$ so $x = \log_3 n$
$$T(n) = \log_3 n + T(n/n)$$

$$T(n) = \log_3 n + 1$$

$$T(n) = \log_3 n$$

6. To analyze the time complexity of the "brute-force" algorithm in the programming part of this assignment, we would like to count the number of all possible schedules.

- (a) Explain the number of all possible schedules in terms of n (number of candidate courses) and m (number of time slots per candidate course).
- (b) Consider a computer that can process 1 billion schedules per second, n is 10, m is 20, explain the number of hours needed to process all possible schedules.
- (c) If we do not want the computer to spend more than 1 minute, explain the largest n the computer can process when m is 20.

We would need to count the total amount of permutations possible, with considering order. We need to consider the order since it would we have a order of preference.

Total number of permutations = m! / (m - n)!

If we had 10 as total number of courses(n) and 20 as total amount of time available per course(m), there would be a total of 670,442,572,800 permutations possible. This would take the computer capable of doing 1 billion schedules per seconds a total of 670 seconds, which is 11 minutes which is .19 hours.

The largest n the computer can process can be calculated by first looking at the total amount of schedules it can process: 1 billion * 60 = 60 billion. Then we can input it into the equation: 60,000,000,000 = 20! / (20 - n)!. Then we can use algebra to solve this equation and find out the total amount of courses is between 6 and 7. Since we cannot look into half a course, the total amount of courses we could search with 20 total amounts of time available per course is 6 courses.