Data Structures and Algorithms - Analysis of Algorithms

1. Arrange the following expressions by growth rate from slowest to fastest.

```
4n^2, log_3n, n!, 3^n, 20n, 2, log_2n, n^{2/3}

Use Stirling's approximation in for help in classifying n!

Stirling's approximation states that n! \approx \sqrt{(2\pi n)} (n/e)^n

2, log_3n, log_3n, n^{2/3}, 20n, 4n^2, 3^n, n!
```

- 2. Estimate the number of inputs that could be processed in the following cases:
 - a. Suppose that a particular algorithm has time complexity T(n) = 3 x 2ⁿ, and that executing an implementation of it on a particular machine takes t seconds for n inputs. Now suppose that we are presented with a machine that is 64 times as fast. How many inputs could we process on the new machine in t seconds?
 The machine can do n + 6 inputs in t seconds (2ⁿ⁺⁶ = 64t).
 - b. Suppose that another algorithm has time complexity $T(n) = n^2$, and that executing an implementation of it on a particular machine takes t seconds for n inputs. Now suppose that we are presented with a machine that is 64 times as fast. How many inputs could we process on the new machine in t seconds? The new machine can do 8n inputs in t seconds $(8n^2 = 64t)$.
 - c. A third algorithm has time complexity T(n) = 8n. Executing an implementation of the algorithm on a particular machine takes t seconds for n inputs. Given a new machine that is 64 times as fast, how many inputs could we process in t seconds?
 The new machine can do 64n inputs in t seconds (512n = 64t).

3. A hardware vendor claims that their latest computer will run 100 times faster than that of their competitor. If the competitor's computer can execute a program on input of size n in one hour, what size input can vendor's computer execute in one hour for each algorithm with the following growth rate equations?

```
    a. n
    100n (n x 100)
    b. n<sup>2</sup>
    10n (n x 100<sup>1/2</sup>)
    c. n<sup>3</sup>
    4.64n (n x 100<sup>1/3</sup>)
    d. 2<sup>n</sup>
    6.64n (n x log<sub>2</sub>100)
```

4. Using the definition of big-Oh, show that 1 is in O(1) and that 1 is in O(n).

The definition of big-Oh states that if a function f(n) is less than or equal to a constant c, times another function g(n) that is greater than or equal to f(n) from an input size n_0 to infinity, f(n) belongs to g(n).

In the first problem, f(n) = 1 and g(n) = 1, and for all n, $f(n) \le g(n)$, so 1 is in O(1). In the second, f(n) = 1 and g(n) = n, so for all n greater than or equal to 1, $f(n) \le g(n)$. Once again, f(n) must be a part of g(n), therefore.

5. For each of the following pairs of functions, either f(n) is in O(g(n)), f(n) is in O(g(n)), or $f(n) = \Theta(g(n))$. For each pair, determine which relationship is correct. Justify your answer.

a.
$$f(n) = log n^2$$
; $g(n) = log n + 5$.

```
\lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
b. f(n) = \sqrt{n}; g(n) = \log n^2
     \lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
c. f(n) = \log^2 n; g(n) = \log n.
     \lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
d. f(n) = n; g(n) = log n.
     \lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
e. f(n) = n \log n + n; g(n) = \log n.
     \lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
f. f(n) = log n^2; g(n) = (log n)^2
     \lim_{n\to\infty} f(n)/g(n) is zero; f(n) = O(g(n))
g. f(n) = 10; g(n) = log 10.
     \lim_{n\to\infty} f(n)/g(n) is 10; f(n) = \Theta(g(n))
h. f(n) = 2^n; g(n) = 10n^2
     \lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
i. f(n) = 2^n; g(n) = n \log n.
     \lim_{n\to\infty} f(n)/g(n) is infinity; f(n) = \Omega(g(n))
j. f(n) = 2^n; g(n) = 3^n
     \lim_{n\to\infty} f(n)/g(n) \text{ is } 2/3; \ f(n) = \Theta(g(n))
k. f(n) = 2^n; g(n) = n^n
     \lim_{n\to\infty} f(n)/g(n) is zero; f(n) = O(g(n))
```

6. Determine Θ for the following code fragments in the average case. Assume that all variables are of type int.

```
a.
    a = b + c;
    d = a + e;
    1
b.
    sum = 0;
    for (i=0; i<3; i++)
           for (j=0; j<n; j++)
                   sum++;
    3n
c.
    sum=0;
           for (i=0; i<n*n; i++)
                   sum++;
    n^2
d.
    for (i=0; i < n-1; i++)
           for (j=i+1; j < n; j++) {
                   tmp = AA[i][j];
                   AA[i][j] = AA[j][i];
                   AA[j][i] = tmp;
            }
    \frac{1}{2} n^2
```

```
e.
    sum = 0;
            for (i=1; i<=n; i++)
      i.
                1. for (j=1; j \le n; j \ge 2)
                        a. sum++;
    \frac{1}{2} n^2
f.
    sum = 0;
            for (i=1; i<=n; i*=2)
                    for (j=1; j<=n; j++)
                            sum++;
    \frac{1}{2} n^2
    Assume that array A contains n values, random takes constant time, and sort
    takes n log n steps.
            for (i=0; i<n; i++) {
                    for (j=0; j<n; j++)
                            A[i] = util.random(n);
                            sort(A);
                    }
    n^3 \log(n)
h. Assume array A contains a random permutation of the values from 0 to n - 1.
```

sum = 0;

for (i=0; i<n; i++)

```
for (j=0; A[j]!=i; j++)

sum++;

n^2

i. sum = 0;

if (EVEN(n))

for (i=0; i < n; i++)

sum++;

else

sum = sum + n;
```

n/2