Homework 1

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Problem 1

```
# a
x < -c(9:16)
## [1] 9 10 11 12 13 14 15 16
# b
last <- length(x)</pre>
x[(last - 2):last]
## [1] 14 15 16
# c
for (i in 1:length(x)) {
    if (x[i]\%2 == 0)
        cat(x[i], " ")
}
## 10 12 14 16
# d
for (i in length(x):1) {
    if (x[i]\%2 == 0) {
        x \leftarrow x[-i]
    }
}
х
## [1] 9 11 13 15
Problem 2
forfunc <- function(n) {</pre>
    sum <- 0
    for (i in 1:n) {
        value <- (1/2)^i
        sum <- sum + value
    }
    return(sum)
}
whilefunc <- function(n) {</pre>
    sum <- 0
    i <- 1
    while (i \le n) {
        value \leftarrow (1/2)^i
        sum <- sum + value
        i = i + 1
    return(sum)
```

```
analyticfunc <- function(n) {
    sum <- (1 - (1/2)^n)
    return(sum)
}

v <- seq(1:30)
compare <- function(v) {
    forresults <- unlist(lapply(v, forfunc))
    whileresults <- unlist(lapply(v, whilefunc))
    analyticresults <- unlist(lapply(v, analyticfunc))

cat("Results of for-loop:\n", forresults, "\n")
    cat("Results of while loop:\n", whileresults, "\n")
    cat("Results of analytic formula:\n", analyticresults, "\n")
}
compare(v)</pre>
```

Results of for-loop:

0.5 0.75 0.875 0.9375 0.96875 0.984375 0.9921875 0.9960938 0.9980469 0.9990234 0.9995117 0.9997559 ## Results of while loop:

0.5 0.75 0.875 0.9375 0.96875 0.984375 0.9921875 0.9960938 0.9980469 0.9990234 0.9995117 0.9997559 ## Results of analytic formula:

0.5 0.75 0.875 0.9375 0.96875 0.984375 0.9921875 0.9960938 0.9980469 0.9990234 0.9995117 0.9997559

As n gets very large, all three approaches get closer and closer to 1 until n = 25, when R begins to round off to 1.

Problem 3

• If $x^3 \in \Theta(x^3)$ then there exists c_1, c_2, x_0 such that $c_1 g(x) \le x^3 \le c_2 g(x)$ for all $x \ge x_0$. $g(x) = x^3$ so the equation becomes

$$c_1 x^3 < x^3 < c_2 x^3$$

which can be satisfied if $c_1 \leq 1, c_2 \geq 1$ and $x_0 = 0$.

If $x^3 \in O(x^3)$ then there exists c, x_0 such that $0 \le x^3 \le cg(x)$ for all $x \ge x_0$. Inserting x^3 for g(x) the equation becomes

$$0 < x^3 < cx^3$$

which can be satisfied if $c \ge 1$ and $x_0 = 0$.

If $x^3 \epsilon \Theta(x^4)$ then there exists c_1, c_2, x_0 such that $c_1 g(x) \le x^4 \le c_2 g(x)$ for all $x \ge x_0$. $g(x) = x^4$ so the equation becomes

$$c_1 x^4 \le x^3 \le c_2 x^4$$

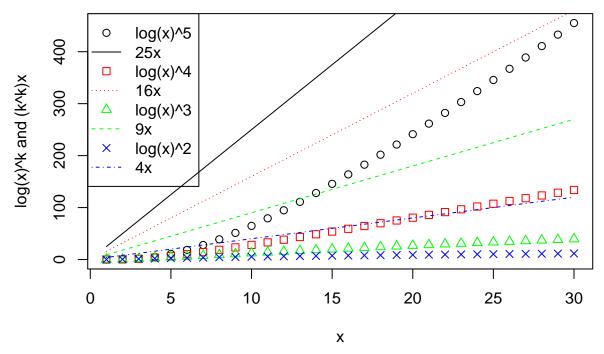
This equation cannot be satisfied for all $x \ge x_0$, since as x goes to infinity, $c_1 x^4$ will be larger than x^3 no matter how small c_1 is. We can also see that

$$\frac{f(x)}{g(x)} = \frac{x^3}{x^4}$$

goes to zero as x increases to infinity, which also indicates that x^3 does not belong to $\Theta(x^4)$, since if $x^3 \epsilon \Theta(x^4)$, the limit of $\frac{x^3}{x^4}$ as x approaches infinity would be a constant greater than 0.

- For any real constants a and b > 0, $(n+a)^b = n^b + bn^{b-1}a^{b-1} + ... + a^b$. As n goes to infinity, the growth rate is determined by n^b , since it is the largest power of n in the polynomial. Thus $g(n) = n^b$. Since the limit of $\frac{f(n)}{g(n)} = \frac{n^b + bn^{b-1}a^{b-1} + ... + a^b}{n^b}$ as n approaches infinity is 1, a constant greater than 0, $(n+a)^b \epsilon \Theta(n^b)$.
- If $(log(n))^k \epsilon O(n)$, then there exists c, n_0 such that $0 \le (log(n))^k \le cn$ for all $n \ge n_0$. One solution is $n_0 = 0, c = k^k$, as demonstrated by the plot below. Thus, $(log(n))^k \epsilon O(n)$.

Comparison between log(x)^k and (k^k)x



• Since $\frac{n}{n+1} = 1 + \frac{1}{n}$, we can write

$$T(\frac{n}{n+1}) = 1 + T(\frac{1}{n})$$

We know that T(1/n) = O(1/n) since the equation $0 \le \frac{1}{n} \le c \frac{1}{n}$, $n \ge n_0$ is satisfied by $c \ge 1$ and $n_0 = 0$. Thus, $\frac{n}{n+1} = 1 + O(\frac{1}{n})$.

• We know that $\sum_{i=0}^{\lceil log_2(n) \rceil} 2^i = 1 + 2 + 4 + 16 + ... + n$. Since

$$\frac{f(n)}{g(n)} = \frac{1+2+\ldots+n}{n} > 0$$

as n approaches infinity, we know that $\sum_{i=0}^{\lfloor \log_2(n) \rfloor} 2^i \epsilon \Theta(n)$.

Problem 4

```
bubblesort <- function(x) {</pre>
    n <- length(x)
    swapped <- TRUE
    while (swapped == TRUE) {
        swapped <- FALSE
        for (i in 1:(n - 1)) {
             if (x[i] > x[i + 1]) {
                 temp \leftarrow x[i]
                 x[i] \leftarrow x[i + 1]
                 x[i + 1] \leftarrow temp
                 swapped <- TRUE
             }
        }
    }
    return(x)
}
mergearrays <- function(x, y) {</pre>
    m = length(x)
    n = length(y)
    if (m == 0) {
        return(z = y)
    }
    if (n == 0) {
        return(z = x)
    if (x[1] \le y[1]) \{
        return(z = c(x[1], mergearrays(x[-1], y)))
        return(z = c(y[1], mergearrays(x, y[-1])))
    }
}
mergesort <- function(x) {</pre>
    n = length(x)
    mid = floor(n/2)
    if (n > 1) {
        return(mergearrays(mergesort(x[1:mid]), mergesort(x[(mid +
```

```
1):n])))
    } else {
        return(x)
    }
}
set.seed(12345)
reps <- 100
time1 \leftarrow time2 \leftarrow rep(0, reps)
for (i in 1:reps) {
    x \leftarrow rnorm(1000, 0, 1)
    ptm <- proc.time()</pre>
    mergesort(x)
    t1 <- proc.time() - ptm
    time1[i] = t1[["elapsed"]]
    ptm <- proc.time()</pre>
    bubblesort(x)
    t2 <- proc.time() - ptm
    time2[i] = t2[["elapsed"]]
}
cat("merge sort average time:", mean(time1))
## merge sort average time: 0.07563
cat("merge sort variance:", var(time1))
## merge sort variance: 0.001187003
cat("bubble sort average time:", mean(time2))
## bubble sort average time: 0.19298
cat("bubble sort variance:", var(time2))
```

bubble sort variance: 0.0002543228

The mean time for merge sort was near 0.07592 seconds while the mean time for bubble sort was near 0.20083 seconds. However, the variance for merge sort (near 0.001056377) was about 4 to 5 times as much as the variance for bubble sort (near 0.0003222637). (The reported values differ from the output of my R code, since the values changed slightly with each run when I modified this document.) These results agree with my expectations, since merge sort is $O(nlog_2n)$ while bubble sort is $O(n^2)$. Thus, as n increases we would expect the worst case for bubble sort to increase faster than the worst case for merge sort.

Bonus Problem

```
}
}
x <- x[-index]
}
if (k == 1) {
    print(smallest)
} else find_kth_smallest(x, k - 1)
}

y <- c(4, 2, 8, 1, -4)
find_kth_smallest(y, 3)</pre>
```

[1] 2

In order to find the k-th smallest element of the array, the algorithm loops through an n-length array k times, removing the largest element each time. The k-th element removed from the array is returned as the k-th smallest element.

Since the algorithm loops k times through an array that starts at length n and is reduced by 1 each time, the number of computational steps in the algorithm can be written as:

$$f(n) = n + (n-1) + (n-2) + \dots + (n-k)$$

This can be rewritten as:

$$f(n) = kn - \sum_{i=1}^{k} i$$

We can find c and n_0 such that $0 \le kn - \sum_{i=1}^k i \le cn$ where $n > n_0$. $n_0 = 0$ and c can be any number greater than k. Thus, $f(n) \in O(n)$.