# **Section Problems**

#### Comparing growth rates 1.

- (a) Simplify each of the following functions to a tight big- $\mathcal{O}$  bound in terms of n. Then order them from fastest to slowest in terms of asymptotic growth. (By "fastest", we mean which function increases the most rapidly as *n* increases.)
  - $\log_4(n) + \log_2(n)$

  - $\frac{n}{2} + 4$   $2^n + 3$
  - 750,000,000
  - $8n + 4n^2$
- (b) Order each of these more esoteric functions from fastest to slowest in terms of asymptotic growth. (By "fastest", we mean which function increases the most rapidly as n increases.) Also state a simplified tight  $\mathcal{O}$  bound for each.
  - 2<sup>n/2</sup>
  - 3<sup>n</sup>
  - 2<sup>n</sup>

#### True or false? 2.

- (a) In the worst case, finding an element in a sorted array using binary search is  $\mathcal{O}(n)$ .
- (b) In the worst case, finding an element in a sorted array using binary search is  $\Omega(n)$ .
- (c) If a function is in  $\Omega(n)$ , then it could also be in  $\mathcal{O}(n^2)$ .
- (d) If a function is in  $\Theta(n)$ , then it could also be in  $\mathcal{O}(n^2)$ .
- (e) If a function is in  $\Omega(n)$ , then it is always in  $\mathcal{O}(n)$ .

#### 3. Code to summation

For each of the following code blocks, give a summation that represents the worst-case runtime in terms of *n*.

int x = 0;for (int i = 0; i < n; i++) {</pre> for (int  $j = 0; j < i; j^{++}$ ) { (a) x++; } }

```
(b) int x = 0;
for (int i = n; i >= 1; i /= 2) {
    x += i;
}
```

# 4. Code modeling

For each of the following code blocks, construct a mathematical function modeling the worst-case runtime of the code in terms of n. Then, give a tight big-O bound of your model.

```
(a)
                    int x = 0;
                    for (int i = 0; i < n; i++) {
                        for (int j = 0; j < n * n / 3; j++) {</pre>
                            x += j;
                        }
                    }
(b)
                    int x = 0;
                    for (int i = n; i >= 0; i -= 1) {
                        if (i % 3 == 0) {
                            break;
                        } else {
                            x += n;
                        }
                    }
(c)
                    int x = 0;
                    for (int i = 0; i < n; i++) {</pre>
                        if (i % 5 == 0) {
                            for (int j = 0; j < n; j^{++}) {
                                 if (i == j) {
                                     x += i * j;
                                 }
                            }
                        }
                    }
(d)
                    int x = 0;
                    for (int i = 0; i < n; i++) {
                        if (n < 100000) {
                            for (int j = 0; j < n; j++) {</pre>
                                x += 1;
                            }
                        } else {
                            x += 1;
                        }
                    }
```

```
(e) int x = 0;
if (n % 2 == 0) {
    for (int i = 0; i < n * n * n * n; i++) {
        x++;
    }
} else {
    for (int i = 0; i < n * n * n; i++) {
        x++;
    }
}
```

# 5. Applying definitions

For each of the following, choose a c and  $n_0$  which show  $f(n) \in \mathcal{O}(g(n))$ . Explain why your values of c and  $n_0$  work.

(a)  $f(n) = 3n + 4, g(n) = 5n^2$ 

(b) 
$$f(n) = 33n^3 + \sqrt{n} - 6, g(n) = 17n^4$$

(c) 
$$f(n) = 17\log(n), g(n) = 32n + 2n\log(n)$$

### 6. Using our definitions

Most of the time in the real world, we don't write formal big-O proofs. The point of having these definitions is not to use them every single time we think about big-O. Instead, we use the formal definitions when a question is particularly tricky, or we want to make a very general statement.

Here are some particularly tricky or general statements that are easier to justify with the formal definitions than with just your intuition.

- (a) We almost never say a function is  $\mathcal{O}(5n)$ , we always say it is  $\mathcal{O}(n)$  instead. Show that this transformation is ok, i.e. that if f(n) is  $\mathcal{O}(5n)$  then it is  $\mathcal{O}(n)$  as well.
- (b) When we decide on the big- $\mathcal{O}$  running time of a function, we like to say that whatever happens on small n doesn't matter. Let's see why with an actual proof. You write two functions to solve the same problem: method1 and method2. method1 takes  $\mathcal{O}(n^2)$  time and method2 takes  $\mathcal{O}(n)$  time. What is the big- $\mathcal{O}$  running time of the following function:

```
public void combined(n){
    if(n < 10000)
    method1(n);
    else
    method2(n);
}</pre>
```

### 7. Memory analysis

For each of the following functions, construct a mathematical function modeling the amount of memory used by the algorithm in terms of n. Then, give a **tight** big-O bound of your model.

```
(a)
                   List<Integer> list = new LinkedList<Integer>();
                   for (int i = 0; i < n * n; i++) {</pre>
                        list.insert(i);
                   }
                   Iterator<Integer> it = list.iterator();
                   while (it.hasNext()) {
                        System.out.println(it.next());
                   }
(b)
                   int[] arr = {0, 0, 0};
                   for (int i = 0; i < n; i++) {</pre>
                        arr[0]++;
                   }
(c)
                   ArrayDictionary<Integer, String> dict = new ArrayDictionary<>();
                   for (int i = 0; i < n; i++) {
                        String curr = "";
                        for (int j = 0; j < i; j++) {</pre>
                            for (int k = 0; k < j; k++) {
                                curr += "?";
                            }
                        }
                        dict.put(i, curr);
                   }
```

Note 1: For simplicity, assume the dictionary has an internal capacity of exactly *n*.

Note 2: The amount of memory used by a *single* character (c) and the amount of memory used by a single int (x) are both constant.

**Note 3**: An ArrayDictionary stores its key-value pairs contiguously, and performs scans through (potentially) the entire data structure when performing an insert() or a find().

### 8. Case and Asymptotic Analysis

(a) What is the BEST case runtime of the following code? What is the case in which it happens?

```
void foo(int n) {
    int result = 0;
    if (n % 2 == 0) {
        for (int i = 0; i < n; i++) {
            for (int j = 0; j < i; j++) {
                result++;
            }
        }
    }
    else {
        for (int i = 1; i < n; i*=2) {
            result++;
        }
    }
    return result;
}</pre>
```

(b) What is the Big-Theta runtime of the code above? (Hint: What are the Big-O and Big-Omega runtimes?)

(c) What is the WORST case runtime for the following function? In what case?

```
public class ArrayIntList {
    private int size;
    private int arr;
    public void foo(int val) {
        size++;
        if (size >= arr.length) {
            temp = new int[arr.length * 2];
            for (int i = 0; i < size; i++) {</pre>
                temp[i] = arr[i];
            }
            temp[size] = val;
            this.arr = temp;
        } else {
            arr[size] = val;
        }
    }
}
```

(d) What is the WORST case Big-O runtime? Give only the TIGHTEST Big-O bound possible.

```
// assume 0 <= n < arr.length</pre>
int foo(int n, int[] arr) {
    if (n < 1000) {
        for (int i = 0; i < n * n; i += 4) {</pre>
             arr[i] = n;
        }
    } else if (n > 50000) {
        arr[n] = 65;
    } else {
        int x = 1000000
        while (x > 50) {
             x /= 2;
        }
    }
    return arr[n];
}
```

(e) What is the BEST case runtime for this problem? What is the WORST case?

```
// n == arr.length
boolean find(int[]arr, int n, int k) {
    for(int i=0; i < n; ++i)
        if(arr[i] == k) return true;
    return false;
}</pre>
```