

Prolegomena unit outline:

- ▶ Algorithms and correctness (last week Wednesday and Friday)
- ▶ Algorithms and efficiency (**this week Wednesday and Friday**)
- ▶ Abstract data types (next week Monday)
- ▶ Data Structures (next week Wednesday and Friday)

Today and Friday:

- ▶ Go over Ex 1.(6 & 7)
- ▶ The general meaning of efficiency
- ▶ The analyses of bounded linear search, binary search, and selection sort
- ▶ The precise meaning of big-oh, big-theta, and big-omega
- ▶ The costs of elemental algorithms
- ▶ The analysis of merge sort and quick sort

How to succeed in CSCI 345:

- ▶ Know your DMFP and Programming II stuff.
- ▶ Read the textbook.
- ▶ Do the practice problems.
- ▶ Figure out the quiz questions.
- ▶ Do the projects on time.
- ▶ Use the project to understand the data structures and algorithms—don't just fiddle with the code until the tests pass.
- ▶ Keep electronic devices away during class.

Quiz question

Loop invariant. A proposition about the state of execution preserved through all iterations.

Correctness claim. A proposition about what an algorithm returns.

Recursion invariant. A proposition about the preconditions to every call to a recursive method or function.

Class invariant. A proposition about the aspects of the state of an instance of a class that do not change while other aspects of the object's state change.

Unused answers

- ▶ A propositions about the interface of a class.
- ▶ A proposition about the special cases of a class.
- ▶ A conjecture about an algorithm's efficiency.
- ▶ A proposition about the number of iterations a loop performs.

1.6 Write a loop invariant to capture the relationships among `sequence`, `smallest_so_far`, `smallest_pos`, and `i` in the following algorithm to find the smallest element in a sequence.

```
def find_smallest(sequence):
    smallest_so_far = sequence[0]
    smallest_pos = 0
    i = 1
    while i < len(sequence) :
        if sequence[i] < smallest_so_far :
            smallest_pos = i
            smallest_so_far = sequence[i]
        i += 1
    return smallest_pos
```

1.7 State and prove a loop invariant to show that the following loop clears the list `sequence`, that is, it sets all of its positions to `None`. Your loop invariant should explain and relate the variables `sequence` and `i`.

```
i = 0
while i < len(sequence):
    sequence[i] = None
    i += 1
```

From the correctness proof of `bounded_linear_search`:

By Invariant 1.c [i is the number of iterations], after at most n iterations, $i = n$ and the guard will fail.

From the correctness proof of `binary_search` (rewritten):

Let i be the number of iterations completed. Suppose $i \geq \lg n$. Then $2^i \geq n$ and $\frac{n}{2^i} \leq 1$.

By Invariant 3.b, [$\text{high} - \text{low} \leq \frac{n}{2^i}$], we have $\text{high} - \text{low} \leq 1$ and the guard fails.

```

def bounded_linear_search(sequence, P):
    a0 found = False
    i = 0
    while not found and i < len(sequence): a1(n + 1)
        a2n found = P(sequence[i])
        i += 1
    if found: a3
        a4 return i - 1
    else :
        a5 return -1

```

$$\begin{aligned}
 T_{bls}(n) &= a_0 + a_1(n + 1) + a_2n + a_3 + \max(a_4, a_5) \\
 &= b_0 + b_1n
 \end{aligned}$$

```

def binary_search(sequence, T0, item):
    c0 low = 0
    high = len(sequence)
    while high - low > 1: c1(lg n + 1)
        c2 lg n mid = (low + high) / 2
        compar = T0(item, sequence[mid])
        if compar < 0: # item comes before mid
            high = mid
        elif compar > 0: # item comes after mid
            low = mid + 1
        else: # item is at mid
            assert compar == 0
            low = mid
            high = mid + 1
    if low < high and T0(item, sequence[low]) == 0: c3
        c4 return low
    else:
        c5 return -1

```

$$\begin{aligned}
 T_{bs}(n) &= c_0 + c_1(\lg n + 1) + c_2 \lg n + c_3 + \max(c_4, c_5) \\
 &= d_0 + d_1 \lg n
 \end{aligned}$$


```

def selection_sort(sequence, T0):
    for i in range(len(sequence)):  $e_0 + e_1 n$ 
        min_pos = i
        min = sequence[i]
        for j in range(i + 1, len(sequence)):  $e_3 n + e_4 \sum_{i=0}^{n-1} (n - i - 1)$ 
            if T0(sequence[j], min) < 0 :  $e_5 \sum_{i=0}^{n-1} (n - i - 1)$ 
                min = sequence[j]
                min_pos = j
            sequence[min_pos] = sequence[i]
            sequence[i] = min

```

$e_2 n$

$$T_{sel}(n) = f_1 + f_2 n + f_3 n^2$$

- ▶ $\exists T : D \rightarrow \mathbb{N}$ relating input to running time on some platform. Interpret the codomain \mathbb{N} as natural numbers in some unit time.
- ▶ $\nexists T_{\text{absolute}} : \mathbb{N} \rightarrow \mathbb{N}$ relating input size to running time on some platform. Interpret the domain \mathbb{N} as the number of items in the list (or other structure, for other algorithms).
- ▶ $\exists T_{\text{worst}} : \mathbb{N} \rightarrow \mathbb{N}$ relating input size to the maximum running time on some platform for all inputs of the given size.
- ▶ $\exists T_{\text{best}} : \mathbb{N} \rightarrow \mathbb{N}$ relating input size to the minimum running time on some platform for all inputs of the given size.
- ▶ $\exists T_{\text{expected}} : \mathbb{N} \rightarrow \mathbb{N}$ relating input size to the expected value of the running time on some platform over all inputs of the given size.

What is big-oh notation?

Big-oh is a way to categorize *functions*:

$O(g)$ is the set of functions that can be bounded above by a scaled version of g .

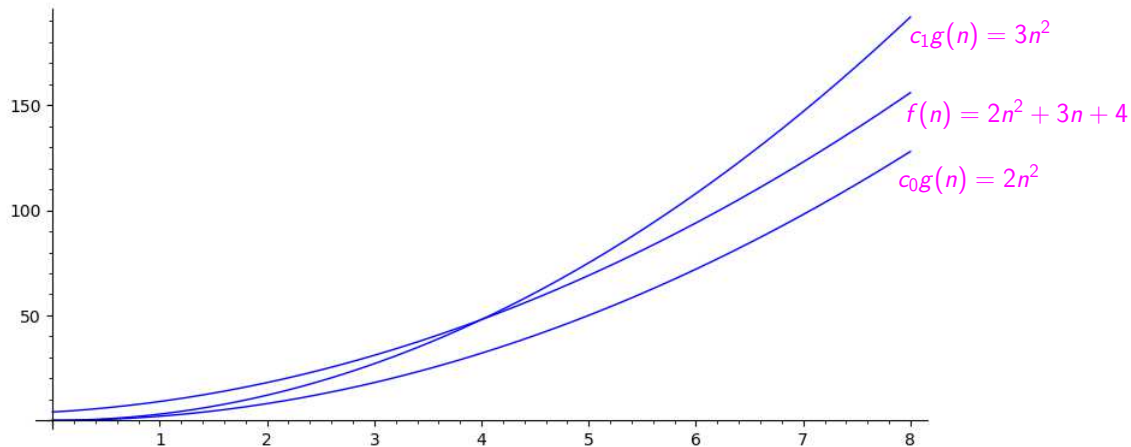
$f(n) = O(g(n))$ (or, more properly $f \in O(g)$) means

$$\exists c, n_0 \in \mathbb{N} \text{ such that } \forall n \in [n_0, \infty), f(n) \leq cg(n)$$

Objections to and misconceptions of big-oh notation take forms such as

- ▶ Big-oh notation specifies only an upper bound of running time, which might be widely imprecise.
- ▶ Big-oh notation measures only the worst case, when the best case or the typical case might be much better.
- ▶ Big-oh ignores constants, which can greatly affect running time in practice.
- ▶ Algorithms that have the same big-oh category can have widely different running times in practice.
- ▶ Big-oh considers only the *size* of the input, when in fact other attributes of the input can greatly affect running time.

$$\Theta(g) = \{f : \mathbb{N} \rightarrow \mathbb{N} \mid \exists c_0, c_1, n_0 \in \mathbb{N} \text{ such that } \forall n \in [n_0, \infty), c_0g(n) \leq f(n) \leq c_1g(n)\}$$



Algorithmic element 1

Can you jump directly to the thing you're looking for?

Algorithmic element 2

Are you descending a binary tree of the data?

Algorithmic element 3

Do you need to touch every element in the data?

Algorithmic element 4

For every element, do you need to descend a tree, or for every element in the tree, do you touch every element?

Algorithmic element 5

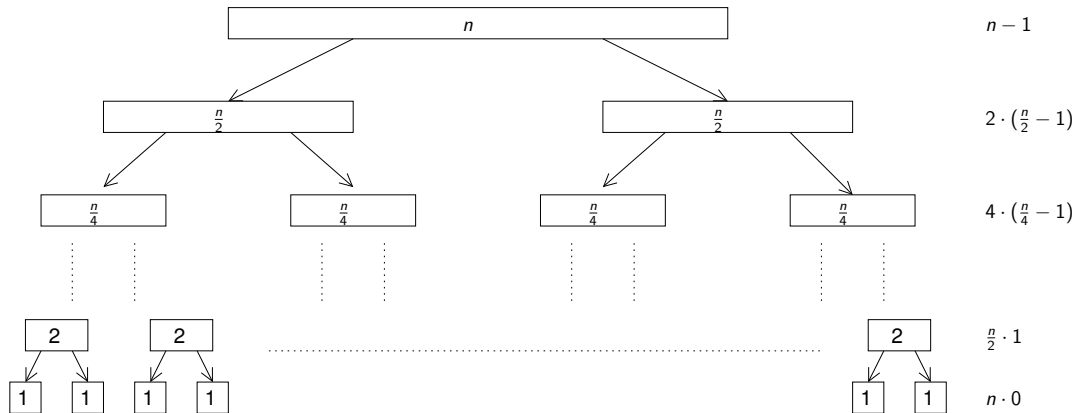
For every element in the data, do you need to a suboperation on the rest of the data?

Algorithmic element 6

Do you need to consider all combinations of input elements?

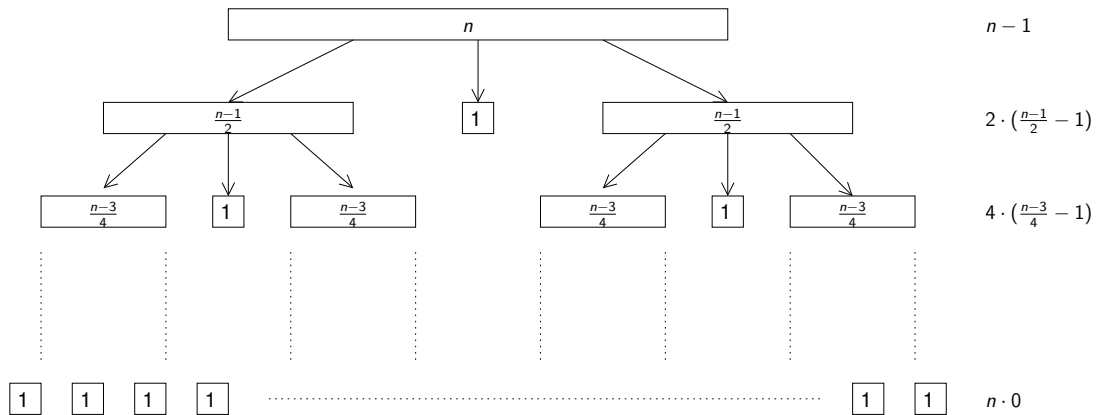
```
int merge_sort_r(int sequence[], int aux[], int low, int high)
{
    if (low + 1 >= high)
        return 0;
    else {
        int compar = 0; // the number of comparisons
        int midpoint = (low + high) / 2; // index to the middle of the range
        int k, n;
        n = high - low;
        compar += merge_sort_r(sequence, aux, low, midpoint);
        compar += merge_sort_r(sequence, aux, midpoint, high);
        compar = merge(sequence, aux, low, high);
        return compar;
    }
}
```

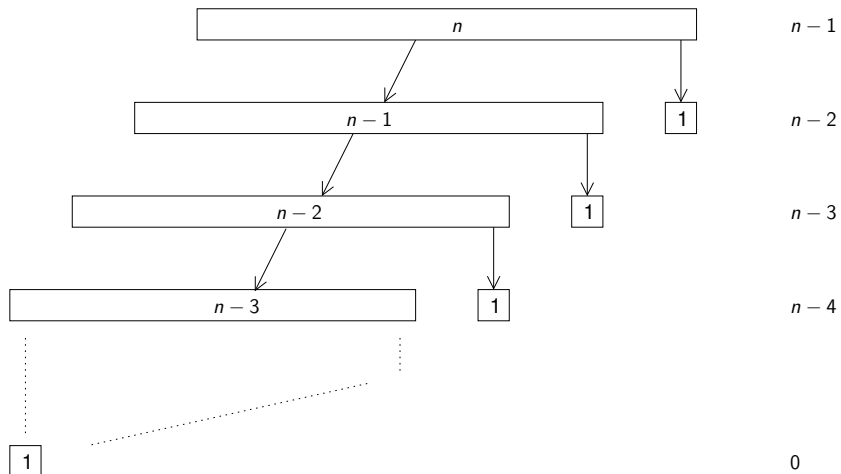
$$C_{ms}(n) = \begin{cases} 0 & \text{if } n \leq 1 \\ n - 1 + 2C_{ms}(\frac{n}{2}) & \text{otherwise} \end{cases}$$



$$\begin{aligned} \sum_{i=0}^{\lg n - 1} 2^i \cdot (\frac{n}{2^i} - 1) &= \sum_{i=0}^{\lg n - 1} n - \sum_{i=0}^{\lg n - 1} 2^i \\ &= n \lg n - n + 1 \end{aligned}$$


```
int quick_sort_r(int sequence[], int low, int high)
{
    if (low + 1 >= high) return 0;
    int i, j, temp;
    int compar = 0;
    for (i = j = low; j < high-1; j++) {
        compar++;
        if (sequence[j] < sequence[high-1])
        {
            temp = sequence[j];
            sequence[j] = sequence[i];
            sequence[i] = temp;
            i++;
        }
    }
    temp = sequence[i];
    sequence[i] = sequence[j];
    sequence[j] = temp;
    return compar + quick_sort_r(sequence, low, i)
        + quick_sort_r(sequence, i+1, high);
}
```





$$(n-1) + (n-2) + (n-3) + \cdots + 1 + 0 = \sum_{i=1}^{n-1} i = \frac{n \cdot (n-1)}{2} = \frac{n^2 - n}{2}$$

Coming up:

*Due **Friday, Jan 20** (end of day):*

Read Sections 1.(3 & 4)

Do Exercises 1.(17–19)

Take quiz

*Due **Tues, Jan 24** (end of day):*

Read Section 2.1

Do Exercise 1.11

Take quiz