Prolegomena unit outline:

- Algorithms and correctness (Friday, Aug 30 and Wed, Sept 4)
- ► Algorithms and efficiency (Fri, Sept 6 and Mon, Sept 9)
- Abstract data types (Wed, Sept 11)
- ▶ Data Structures (Fri, Sept 13 and Mon, Sept 16)

Today and Friday:

- ► Go over quiz and Ex 1.6
- ► 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

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 asnwers

- 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.

Quiz question

What is (not) true about a class invariant?

- ▶ It can be assumed as a precondition to any method call. ✓
- ▶ It caputers what doesn't change about an instance of a class when other parts of that object's state do change. ✓
- ▶ It must be satisfied as a postcondition to any method call. ✓
- It applies specifically to static variables X

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</pre>
```

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 $i \ge \lg n$. Then $2^i \ge n$ and $\frac{n}{2^i} \le 1$.

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

```
def bounded_linear_search(sequence, P):
 an (found = False
    i = 0
    while (not found and i < len(sequence) : a_1(n+1)
    \frac{a_2n}{a_2n} found = P(sequence[i])
        i += 1
    if (found): 3
       return i - 1
    else :
     as return -1
                 T_{bls}(n) = a_0 + a_1(n+1) + a_2n + a_3 + \max(a_4, a_5)
                          = b_0 + b_1 n
```

```
def binary_search(sequence, T0, item):
   \log = 0
    high = len(sequence)
    while high - low > 1): c_1(\lg n + 1)
  \log n \pmod{= (\text{low} + \text{high}) / 2}
        compar = TO(item, sequence[mid])
        if compar < 0 : # item comes before mid</pre>
             high = mid
        elif compar > 0 : # item comes after mid
             low = mid + 1
                             # item is at mid
        else :
             assert compar == 0
             low = mid
             high = mid + 1
    if (low < high and TO(item, sequence[low]) == 0 : </pre>
     c4 (return low)
    else :
     c<sub>5</sub> (return -1)
                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
```

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

- ▶ \exists $T: D \to \mathbb{N}$ relating input to running time on some platform. Interpret the codomain \mathbb{N} as natural numbers in some unit time.
- ▶ \not $T_{absolute}: \mathbb{N} \to \mathbb{N}$ 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 $T_{worst} : \mathbb{N} \to \mathbb{N}$ relating input size to the maximum running time on some platform for all inputs of the given size.
- $ightharpoonup \exists \mathcal{T}_{\mathsf{best}} : \mathbb{N} \to \mathbb{N}$ relating input size to the minimum running time on some platform for all inputs of the given size.
- ▶ \exists $T_{\text{expected}} : \mathbb{N} \to \mathbb{N}$ 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.

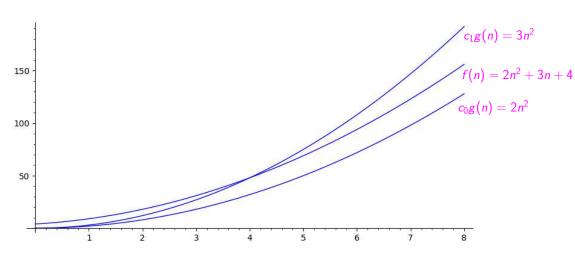
$$f(n) = O(g(n))$$
 (or, more properly $f \in O(g)$) means

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

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.

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



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?

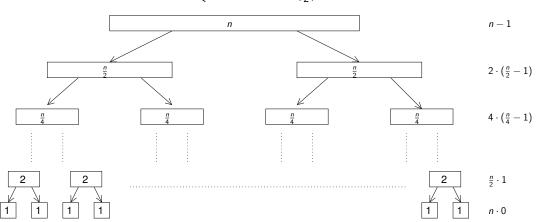
Algorithmic element 6

Do you need to consider all combinations of input elements?



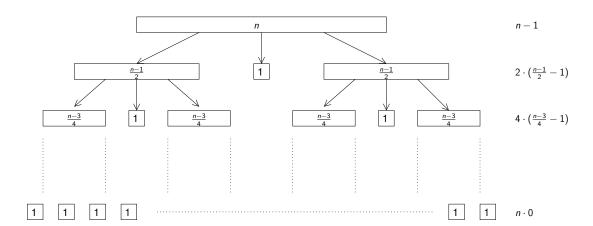
```
int merge_sort_r(int sequence[], int aux[], int low, int high)
 if (low + 1 >= high)
    return 0;
  else {
    int compars = 0; // the number of comparisons
    int midpoint = (low + high) / 2; // index to the middle of the range
    int k, n;
    n = high - low;
    compars += merge_sort_r(sequence,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

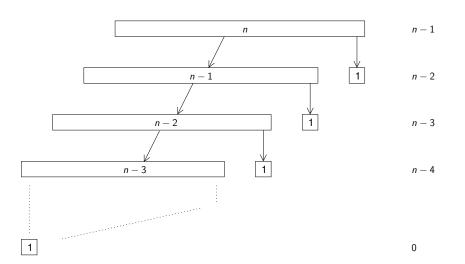
$$C_{ms}(n) = \left\{ egin{array}{ll} 0 & ext{if } n \leq 1 \ n-1+2C_{ms}(rac{n}{2}) & ext{otherwise} \end{array}
ight.$$



$$\sum_{i=0}^{\lg n-1} 2^{i} \cdot \left(\frac{n}{2^{i}} - 1\right) = \sum_{i=0}^{\lg n-1} n - \sum_{i=0}^{\lg n-1} 2^{i}$$
$$= n \lg n - n + 1$$

```
int quick_sort_r(int sequence[], int low, int high)
 if (low + 1 >= high) return 0;
  int i, j, temp;
  int compars = 0;
 for (i = j = low; j < high-1; j++) {
    compars++;
    if (sequence[j] < sequence[high-1])</pre>
        temp = sequence[j];
        sequence[j] = sequence[i];
        sequence[i] = temp;
        i++:
 temp = sequence[i];
  sequence[i] = sequence[j];
  sequence[j] = temp;
 return compars + 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 Mon, Sept 9 (end of day):
Read Sections 1.(3 & 4)
Do Exercises 1.(17 & 18)
Take quiz
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Due **Wed, Sept 11** (end of day): Read Section 2.1 Do Exercise 1.11 Take quiz