Data Structures and Algorithms

Live Class 5

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Announcements

Homework 1 deadline tomorrow:

- Your functions should almost always return rather than print.
- Make sure not to print anything unless explicitly requested.

I will post a mock exam next week

Today

- 1. Recap
- 2. Sorting algorithms:
 - ▶ Selection sort
 - ▶ Merge sort

Go to menti.com

What is the output?

```
1  a = 'hiphop'
2  b = a[2]
3  c = a[:len(a) - 1]
4  print(b, c)
```

- A. i hipho
- B. p hipho
- C. i hiph
- D. p hiph
- E. I don't know

What is the output?

```
1  a = 'hip'
2  for i in range(1, len(a)):
3     print(a[i] + a[i - 1])
```

- A. An error
- B. hp, ih, pi
- C. ih, pi
- D. hi, ip
- E. I don't know

This algorithm is...?

```
def fun_function(s):
    # s is a list of length n
    new_list = []

for v1 in s:
    for v2 in s:
        new_list.append(v1 + v2)
return new_list
```

- A. Constant time O(1)
- B. Linear time O(n)
- C. Quadratic time $O(n^2)$
- D. None of the above
- E. I don't know

Which of these reverses the string?

```
A:
                                         B٠
s = 'hiphop'
                                          s = 'hiphop'
new_s = s[0:len(s):1]
                                          new_s = s[::-1]
C:
                                         D:
 s = 'hiphop'
                                          s = 'hiphop'
new_s = ''
                                          new_s = ''
for char in s:
                                          for i in range(len(s)):
                                              new_s += s[len(s) - i - 1]
     new_s = new_s + char
```

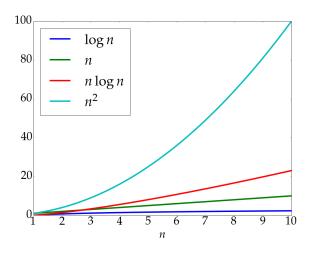
E: More than one of the above

Asymptotic analysis

- 1: measure number of basic operations as function of input size
- 2: focus on worst-case analysis
- 3: ignore constant factors and lower-order terms
- 4: only care about large inputs

Formal way to describe this approach.

▶ Big-O notation: upper bound on worst-case running time



Big O: for **large enough inputs**, an O(n) algorithm will be slower than $O(\log(n))$

Basic operations

Operations that a computer can perform "quickly" (constant time O(1) for any input)

- Arithmetic operations
- Comparisons
- ► Variable assignment
- Memory access

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What if the data structure is more complicated?

- ► For example, if L is a list: L.append(), L[5]?
- Are these basic constant-time operations?
- ▶ Wait for it... assume for now that list operations are constant time

Complexity classes

Fast algorithm: worst-case running time grows slowly with input size

- \triangleright O(1): constant running time basic operations
- \triangleright $O(\log n)$: logarithmic running time binary search
- ► *O*(*n*): linear running time linear search
- \triangleright $O(n \log n)$: log-linear running time ??
- $ightharpoonup O(n^c)$: polynomial running time ??
- $ightharpoonup O(c^n)$: exponential running time ??

Sorting algorithms

So if we have an unsorted list, should we sort it first?

- ► Suppose complexity *O*(*sort*(*n*))
- Is it less work to sort and then do binary search than to do linear search?
- ▶ Is $sort(n) + \log(n) < n$?
- ► No...

But what if we need to search repeatedly, say *k* times?

- ▶ Is $sort(n) + k \log(n) < kn$?
- Depends on k...

56 24 99 32 9 61 57 79	56	24	99	32	9	61	57	79
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56	24	99	32	9	61	57	79
9	24	99	32	56	61	57	79

56	3 24	99	32	9	61	57	79
9	24	99	32	56	61	57	79
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9	24		32	56		99	61	1	57	79
9	24		32	56		57	61	1	99	79
9	24		32	56		57	61		99	79
9	24	1	32	56		57	61		79	99



In words: Find smallest item and move to front (swap with first unsorted item). Repeat with remaining unsorted items.

$$n-1+n-2+...+2+1 = n(n-1)/2$$
 comparisons -> $O(n^2)$

Selection sort algorithm

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Python:

```
def selection_sort(L):
    M = L[:] # make a copy to preserve original list
    n = len(M)

for index in range(n):
    min_index = find_min_index(M, index) # index with smallest element
    M[index], M[min_index] = M[min_index], M[index] # swap positions
return M
```

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Can we do better?

- ▶ Yes! Merge sort is $O(n \log n)$
- But you can't do any better than that...

The factorial of n is the product of integers 1, ..., n.

- ► As a function: fact(n) = $n \times (n-1) \times (n-2) \times \cdots \times 2 \times 1$
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Factorial can be expressed as a smaller version of itself:

```
1  def fact_rec(n):
2     if n == 0:
3         return 1
4     else:
5         return n*fact_rec(n-1)
6     print(fact_rec(4))
```

This is called recursion

- Function calls itself
- ► Can make some problems easier to define → merge sort!

Merge sort idea

Divide and conquer:

- Identify smallest possible "base case" subproblems that are easy to solve
- Divide large problem and solve smaller subproblems
- Find a way to combine subproblem solutions to solve larger problems

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- ▶ Base case: if list length n < 2, the list is sorted
- ▶ Divide: if list length $n \ge 2$, split into two lists and merge sort each
- ► Combine (merge) the results of the two smaller merge sorts

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Compare items at indices i1 = i2 = 0, update with every copy operation

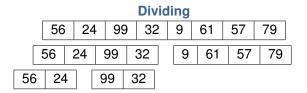
What is the complexity of this operation?

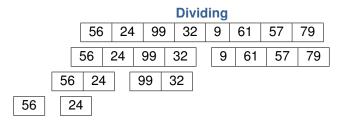
- ▶ Lengths of lists are n_1 and n_2
- ► Two lists of lengths n_1 and n_2 : $O(n_1 + n_2)$ copy operations (need to copy each item)
- No more comparisons than copy operations

Dividing

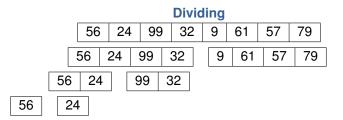
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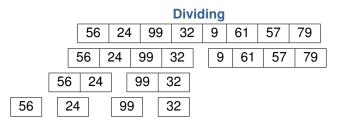


Merging



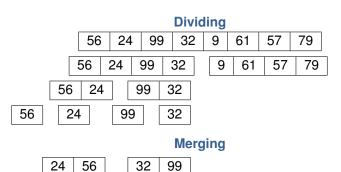
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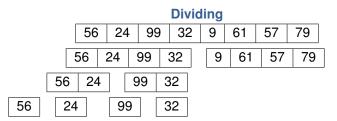
24 56

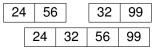


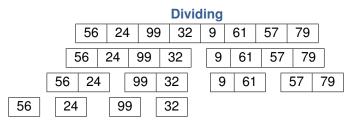
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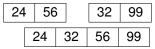
24 56

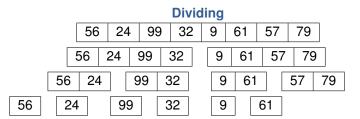


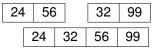


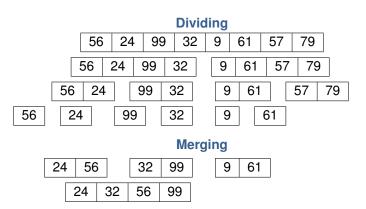


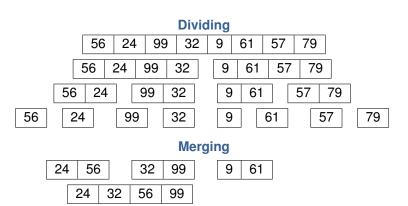


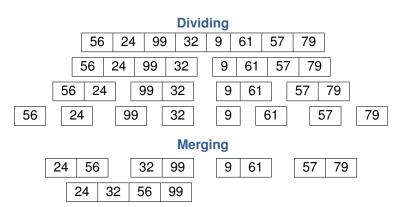


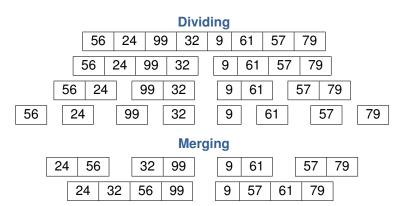


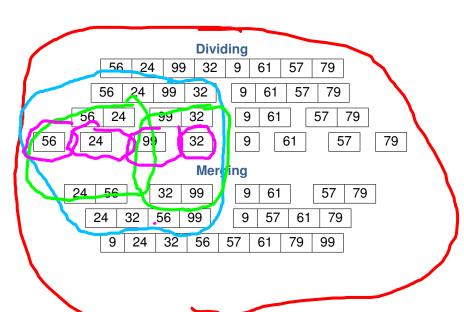












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Merge sort complexity = merging \times # number of divisions

► Number of division levels $O(\log n)$ (like binary search)

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- Does need some more space due to copying lists

Complexity classes

Fast algorithm: worst-case running time grows slowly with input size

- ► O(1): constant running time primitive operations
- \triangleright $O(\log n)$: logarithmic running time binary search
- ► *O*(*n*): linear running time linear search
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But other algorithms are better on average

- Python uses timsort (In 2002, a Dutch guy called Tim got frustrated with existing algorithms)
- Exploit the fact that lists tend to be partly sorted already

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Review

Sorting is a canonical computer science problem

- ► We've looked at two (of many) algorithms
- Selection sort involves repeatedly finding minimum element – intuitive but slow
- Merge sort is blazingly fast and has a neat recursive structure

Review exercises

- Implement sorting
- ► More looping and function practice