

SEARCHING AND SORTING ALGORITHMS

6.0001 LECTURE 12

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- search algorithm – method for finding an item or group of items with specific properties within a collection of items
- collection could be implicit
 - example – find square root as a search problem
 - exhaustive enumeration
 - bisection search
 - Newton-Raphson
- collection could be explicit
 - example – is a student record in a stored collection of data?

- linear search
 - **brute force** search (aka British Museum algorithm)
 - list does not have to be sorted
- bisection search
 - list **MUST be sorted** to give correct answer
 - saw two different implementations of the algorithm

LINEAR SEARCH ON UNSORTED LIST: RECAP

```
1 def linear_serch(L,e):
2     found = False
3     for i in range(len(L)):
4         if e==L[i]:
5             found = True
6     return found
```

- must look through all elements to decide it's not there
- $O(\text{len}(L))$ for the loop * $O(1)$ to test if $e == L[i]$
- overall complexity is $O(n)$ -where n is $\text{len}(L)$. Assumes we can retrieve element of list in constant time
- speed up a little by returning True here, but speed up doesn't impact worst case

```
1 if e==L[i]:
2     found = True
```

LINEAR SEARCH ON UNSORTED LIST: RECAP

Example code:

```
1 def linear_serch(L,e):
2     found = False
3     for i in range(len(L)):
4         if e==L[i]:
5             found = True
6     return found
7 lista = [4,2,6,7]
8 print(linear_serch(lista,8))
9 print(linear_serch(lista,6))
```

Output:

False
True

LINEAR SEARCH ON SORTED LIST:RECAP

```
1 def search(L,e):
2     for i in range(len(L)):
3         return True
4     if L[i] > e:
5         return False
6     return False
```

- must only look until reach a number greater than e
- $O(\text{len}(L))$ for the loop * $O(1)$ to test if $e == L[i]$
- overall complexity is $O(n)$ -where n is $\text{len}(L)$

LINEAR SEARCH ON SORTED LIST: RECAP

Example code:

```
1 def search(L,e):
2     for i in range(len(L)):
3         return True
4     if L[i] > e:
5         return False
6     return False
7 lista = [3,4,6,8,9]
8 print(linear_serch(lista,6))
9 print(linear_serch(lista,5))
```

Output:

True
False

USE BISECTION SEARCH: RECAP

- 1 Pick an index, i , that divides list in half
- 2 Ask if $L[i] == e$
- 3 if not, ask if $L[i]$ is larger or smaller than e
- 4 Depending on answer, search left or right half of L for e

A new version of a divide-and conquer algorithm

- Break into smaller version of problem (smaller list), plus some simple operations
- Answer to smaller version is answer to original problem

BISECTION SEARCH IMPLEMENTATION: RECAP

```
1 def bisect_search2(L,e):
2     def bisect_search_helper(L,e,low,high):
3         if high == low:
4             return L[low]==e
5         mid = (low+high)//2
6         if L[mid] == e:
7             return True
8         elif L[mid]>e:
9             if low == mid: #nothing left to search
10                return False
11            else:
12                return bisect_search_helper(L,e,low,mid-1)
13        else:
14            return bisect_search_helper(L,e,mid+1,high)
15 if len(L)==0:
16     return False
17 else:
18     return bisect_search_helper(L,e,0,len(L) -1)
```

COMPLEXITY OF BISECTION SEARCH: RECAP

- **bisect_serch2** and its helper
 - $O(\log n)$ bisection serch calls
 - reduce size of problem by factor of 2 on each step
 - pass list and indices as parameters
 - list never copied, just re-passed as pointer
 - constant work inside funcion
 - \rightarrow **$O(\log n)$**

BISECTION SEARCH IMPLEMENTATION: RECAP

Example code:

```
1 def bisect_search2(L,e):
2     def bisect_search_helper(L,e,low,high):
3         if high == low:
4             return L[low]==e
5         mid = (low+high)//2
6         if L[mid] == e:
7             return True
8         elif L[mid]>e:
9             if low == mid: #nothing left to search
10                return False
11            else:
12                return bisect_search_helper(L,e,low,mid-1)
13        else:
14            return bisect_search_helper(L,e,mid+1,high)
15    if len(L)==0:
16        return False
17    else:
18        return bisect_search_helper(L,e,0,len(L) -1)
19 lista = [4,2,6,7]
20 print(bisect_search2(lista, 6))
21 print(bisect_search2(lista, 5))
```

Output:

True
False

SEARCHING A SORTED LIST – n is $\text{len}(L)$

- using **linear search**, search for an element is **$O(n)$**
- using **binary search**, can search for an element in **$O(\log n)$**
 - assumes the **list is sorted!**
- when does it make sense to **sort first then search?**
 - $\text{SORT} + O(\log n) < O(n) \rightarrow \text{SORT} < O(n) - O(\log n)$
 - when sortAng is less than $O(n)$
- **NEVER TRUE!**
 - to sort a collection of n elements must look at each one at least once!

AMORTIZED COST – n is $\text{len}(L)$

- why bother sorting first?
- in some cases, may **sort a list once** then do **many searches**
- **AMORTIZE cost** of the sort over many searches
- $\text{SORT} + kO(\log n) < kO(n)$
→ for large K , **SORT time becomes irrelevant**, if cost of sorting is small enough

- Want to efficiently sort a list of entries (typically numbers)
- Will see a range of methods, including one that is quite efficient

MONKEY SORT

- aka bogosort, stupid sort, slowsort, permutaAon sort, shotgun sort
- to sort a deck of cards
 - throw them in the air
 - pick them up
 - are they sorted?
 - repeat if not sorted



COMPLEXITY OF BOGO SORT

```
1 def bogo_sort(L):  
2     while not is_sorted(L):  
3         random.shuffle(L)
```

- best case: **$O(n)$** where **n** is **$\text{len}(L)$** to check if sorted
- worst case: $O(?)$ it is **unbounded** if really unlucky

COMPLEXITY OF BOGO SORT

Example code:

```
1 import random
2 def is_sorted(L):
3     sort = True
4     for i in range(0, len(L)-1):
5         if L[i] > L[i+1]:
6             sort = False
7     return sort
8 def bogo_sort(L):
9     while not is_sorted(L):
10         random.shuffle(L)
11     return L
12
13 lista=[2,1,7,9,10]
14 print(bogo_sort(lista))
```

Output:

[1, 2, 7, 9, 10]

BUBBLE SORT

- **compare consecutive pairs** of elements
- **swap elements** in pair such that smaller is first
- when reach end of list, **start over** again
- stop when **no more swaps** have been made
- largest unsorted element always at end a_er pass, so at most n passes



COMPLEXITY OF BUBBLE SORT

```
1 def bubble_sort(L):
2     swap = False
3     while not swap:
4         swap = True
5         for j in range(1, len(L)):
6             if L[j-1] > L[j]:
7                 swap = False
8                 temp = L[j]
9                 L[j] = L[j-1]
10                L[j-1] = temp
```

- inner for loop is for doing the **comparisons**
- outer while loop is for doing **multiple passes** until no more swaps
- **$O(n^2)$ where n is $\text{len}(L)$** to do $\text{len}(L)-1$ comparisons and $\text{len}(L)-1$ passes

COMPLEXITY OF BUBBLE SORT

Example code:

```
1 def bubble_sort(L):
2     swap = False
3     while not swap:
4         swap = True
5         for j in range(1, len(L)):
6             if L[j-1] > L[j]:
7                 swap = False
8                 temp = L[j]
9                 L[j] = L[j-1]
10                L[j-1] = temp
11     return L
12 lista=[7,1,4,5,3]
13 print(bubble_sort(lista))
```

Output:

[1, 3, 4, 5, 7]

SELECTION SORT

- first step
 - extract **minimum element**
 - **swap it** with element at **index 0**
- subsequent step
 - in remaining sublist, extract **minimum element**
 - **swap it** with the element at **index 1**
- keep the left portion of the list sorted
 - at i 'step, **first i elements in list are sorted**
 - all other elements are bigger than first i elements

- loop invariant
 - given prefix of list $L[0:i]$ and suffix $L[i+1:\text{len}(L)]$, then prefix is sorted and no element in prefix is larger than smallest element in suffix
 - 1 base case: prefix empty, suffix whole list - invariant true
 - 2 induction step: move minimum element from suffix to end of prefix. Since invariant true before move, prefix sorted after append
 - 3 when exit, prefix is entire list, suffix empty, so sorted

COMPLEXITY OF SELECTION SORT

```
1 def selection_sort(L):
2     suffixSt = 0
3     while suffixSt != len(L):
4         for i in range(suffixSt, len(L)):
5             if L[i] < L[suffixSt]:
6                 L[suffixSt], L[i] = L[i], L[suffixSt]
7         suffixSt += 1
```

- outer loop executes $\text{len}(L)$ times
- inner loop executes $\text{len}(L) - i$ times
- complexity of selection sort is **$O(n^2)$ where n is $\text{len}(L)$**

COMPLEXITY OF SELECTION SORT

Example code:

```
1 def selection_sort(L):
2     suffixSt = 0
3     while suffixSt != len(L):
4         for i in range(suffixSt, len(L)):
5             if L[i]<L[suffixSt]:
6                 L[suffixSt], L[i] = L[i], L[suffixSt]
7             suffixSt +=1
8     return L
9 lista=[4,2,5,9,3]
10 print(selection_sort(lista))
```

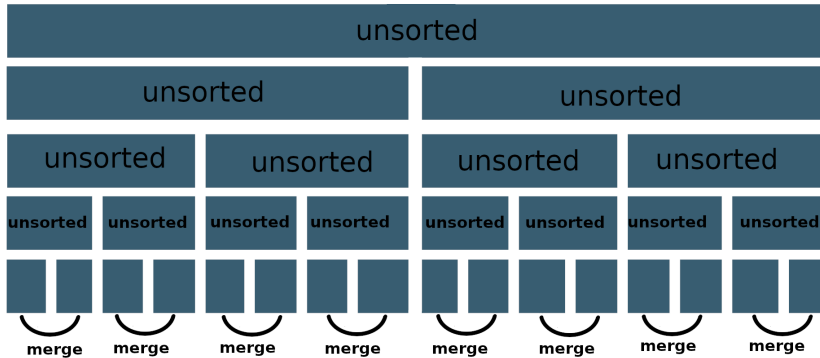
Output:

[2, 3, 4, 5, 9]

- use a divide-and-conquer approach:
 - ① if list is of length 0 or 1, already sorted
 - ② if list has more than one element, split into two lists, and sort each
 - ③ merge sorted sublists
 - ① look at first element of each, move smaller to end of the result
 - ② when one list empty, just copy rest of other list

MERGE SORT

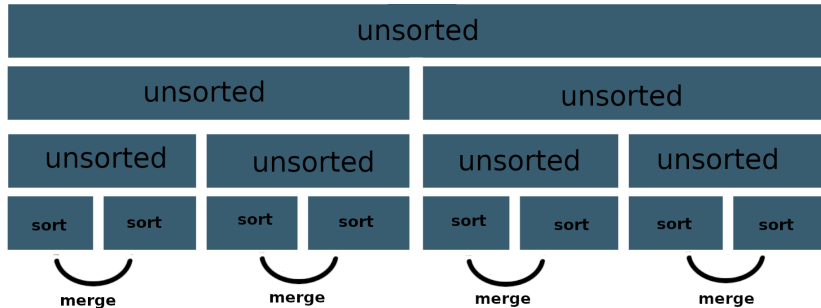
- divide and conquer



- **split list in half** until have sublists of only 1 element

MERGE SORT

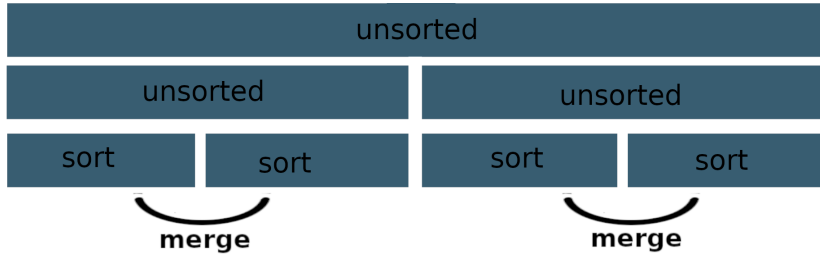
- divide and conquer



- merge such that **sublists will be sorted after merge**

MERGE SORT

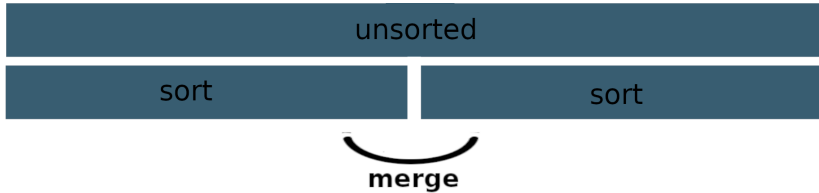
- divide and conquer



- merge sorted sublists
- sublists will be sorted after merge

MERGE SORT

- divide and conquer



- merge sorted sublists
- sublists will be sorted after merge

- divide and conquer - done!

sort

EXAMPLE OF MERGING

Left in list 1	Left in list 2	Compare	Result
[1,5,12,18,19,20]	[2,3,4,17]	1,2	[]
[5,12,18,19,20]	[2,3,4,17]	5,2	[1]
[5,12,18,19,20]	[3,4,17]	5,3	[1,2]
[5,12,18,19,20]	[4,17]	5,4	[1,2,3]
[5,12,18,19,20]	[17]	5,17	[1,2,3,4]
[12,18,19,20]	[17]	12,17	[1,2,3,4,5]
[18,19,20]	[17]	18,17	[1,2,3,4,5,12]
[18,19,20]	[]	18,-	[1,2,3,4,5,12,17]
[]	[]		[1,2,3,4,5,12,17,18,19,20]

MERGING SUBLISTS STEP

```
1 def merge(left, right):
2     result = []
3     i,j = 0,0
4     while i < len(left) and j < len(right):
5         if left[i] < right[j]:
6             result.append(left[i])
7             i += 1
8         else:
9             result.append(right[j])
10            j += 1
11     while (i < len(left)):
12         result.append(left[i])
13         i += 1
14     while (j < len(right)):
15         result.append(right[j])
16         j += 1
17     return result
```


MERGING SUBLISTS STEP - CODE EXPLAINED

```
1 if left[i] < right[j]:
2     result.append(left[i])
3     i += 1
4 else:
5     result.append(right[j])
6     j += 1
```

- left and right sublists are ordered
- move indices for sublists depending on which sublist holds next smallest element

```
1 while (i < len(left)):
2     result.append(left[i])
3     i += 1
```

- when right sublist is empty

```
1 while (j < len(right)):
2     result.append(right[j])
3     j += 1
```

- when left sublist is empty

MERGING SUBLISTS

Example code:

```
1 def merge(left, right):
2     result = []
3     i,j = 0,0
4     while i < len(left) and j < len(right):
5         if left[i] < right[j]:
6             result.append(left[i])
7             i += 1
8         else:
9             result.append(right[j])
10            j += 1
11    while (i < len(left)):
12        result.append(left[i])
13        i += 1
14    while (j < len(right)):
15        result.append(right[j])
16        j += 1
17    return result
18 lista1=[2,4,7,9]
19 lista2=[3,8,13]
20 merge(lista1,lista2)
```

Output:

[2, 3, 4, 7, 8, 9, 13]

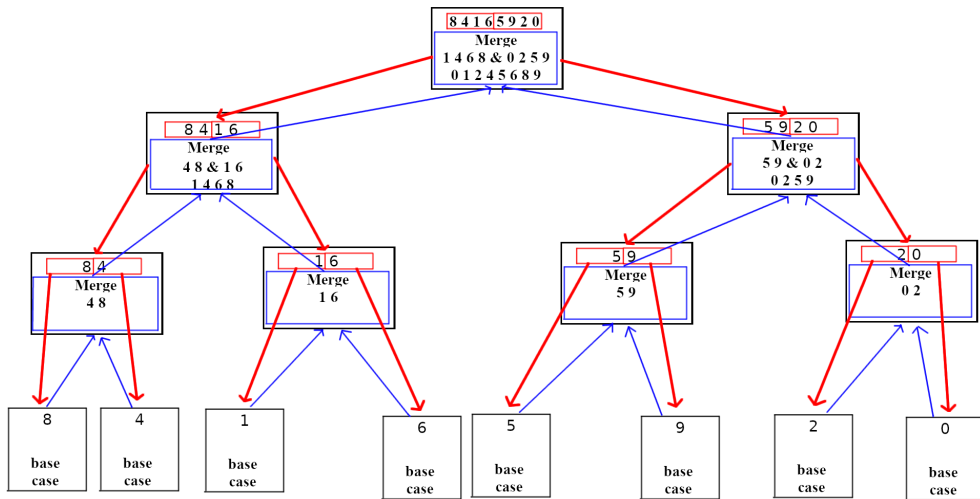
COMPLEXITY OF MERGING SUBLISTS STEP

- go through two lists, only one pass
- compare only **smallest elements in each sublist**
- $O(\text{len}(\text{left}) + \text{len}(\text{right}))$ copied elements
- $O(\text{len}(\text{longer list}))$ comparisons
- **linear in length of the lists**

MERGE SORT ALGORITHM – RECURSIVE

```
1 def merge_sort(L):
2     if len(L) < 2:
3         return L[:]
4     else:
5         middle = len(L)//2
6         left = merge_sort(L[:middle])
7         right = merge_sort(L[middle:])
8         return merge(left, right)
```

- **divide list** successively into halves
- depth-first such that **conquer smallest pieces down one branch** first before moving to larger pieces



COMPLEXITY OF MERGE SORT

- **at first recursion level**
 - $n/2$ elements in each list
 - $O(n) + O(n) = O(n)$ where n is $\text{len}(L)$
- **at second recursion level**
 - $n/4$ elements in each list
 - two merges $\rightarrow O(n)$ where n is $\text{len}(L)$
- each recursion level is $O(n)$ where n is $\text{len}(L)$
- **dividing list in half** with each recursive call
 - $O(\log(n))$ where n is $\text{len}(L)$
- overall complexity is **$O(n \log(n))$ where n is $\text{len}(L)$**

SORTING SUMMARY – n is $\text{len}(L)$

- bogo sort
 - randomness, unbonded $O()$
- bubble sort
 - $O(n^2)$
- selection sort
 - $O(n^2)$
 - guaranteed the first i elements were sorted
- merge sort
 - $O(n \log(n))$
- $O(n \log(n))$ is the fastest a sort can be

WHAT HAVE WE SEEN IN 6.0001?

- represent knowledge with **data structures**
- **iteration and recursion** as computational metaphors
- **abstraction** of procedures and data types
- **organize and modularize** systems using object classes and methods
- different classes of **algorithms**, searching and sorting
- **complexity** of algorithms

- learn computational modes of thinking
- begin to master the art of computational problem solving
- make computers do what you want them to do

Hope we have started you down the path to being able to think and act like a computer scientist

WHAT DO COMPUTER SCIENTISTS DO?

- they think computationally
 - abstractions, algorithms, automated execution
- just like the three r's: reading, 'riting, and 'rithmetic - computational thinking is becoming a fundamental skill that every well-educated person will need



Alan Turing
Image in the
Public
Domain,
courtesy of
[Wikipedia](#)
[Commons](#).



Ada Lovelace
Image in the
Public
Domain,
courtesy of
[Wikipedia](#)
[Commons](#).

THE THREE A'S OF COMPUTATIONAL THINKING

- abstraction
 - choosing the right abstractions
 - operating in multiple layers of abstraction simultaneously
 - defining the relationships between the abstraction layers
- automation
 - think in terms of mechanizing our abstractions
 - mechanization is possible – because we have precise and exacting notations and models; and because there is some “machine” that can interpret our notations
- algorithms
 - language for describing automated processes
 - also allows abstraction of details
 - language for communicating ideas & processes

- how difficult is this problem and how best can I solve it?
 - theoretical computer science gives precise meaning to these and related questions and their answers
- thinking recursively
 - reformulating a seemingly difficult problem into one which we know how to solve
 - reduction, embedding, transformation, simulation

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