

Quicksort

Madhavan Mukund

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Programming, Data Structures and Algorithms using Python
Week 3

Shortcomings of merge sort

- Merge needs to create a new list to hold the merged elements
 - No obvious way to efficiently merge two lists in place
 - Extra storage can be costly
- Inherently recursive
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- Merging happens because elements in the left half need to move to the right half and vice versa
 - Consider an input of the form `[0,2,4,6,1,3,5,9]`

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- Inherently recursive
 - Recursive calls and returns are expensive
- Merging happens because elements in the left half need to move to the right half and vice versa
 - Consider an input of the form `[0,2,4,6,1,3,5,9]`
- Can we divide the list so that everything on the left is smaller than everything on the right?
 - No need to merge!

Divide and conquer without merging

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 - So $T(n)$ is $O(n \log n)$
- How do we find the median?
 - Sort and pick up the middle element
 - But our aim is to sort the list!
 - Instead pick some value in L — **pivot**
 - Split L with respect to the pivot element

Quicksort [C.A.R. Hoare]

- Choose a pivot element
 - Typically the first element in the array

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High level view of quicksort

- Input list

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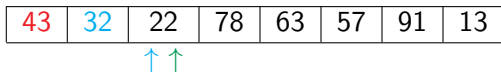
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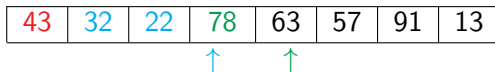
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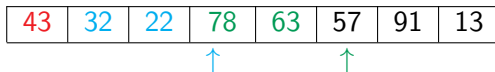
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- Maintain two indices to mark the end of the **Lower** and **Upper** segments
- After partitioning, exchange the pivot with the last element of the **Lower** segment

Quicksort code

- Scan the list from left to right
- Four segments: **Pivot**, **Lower**, **Upper**, Unclassified
- Classify the first unclassified element
 - If it is larger than the pivot, extend **Upper** to include this element
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def quicksort(L,l,r): # Sort L[l:r]
    if (r - l <= 1):
        return(L)
    (pivot,lower,upper) = (L[l],l+1,l+1)
    for i in range(l+1,r):
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 - Can build the lower and upper segments from opposite ends and meet in the middle

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- By partitioning the list carefully, we avoid a merge step
 - This allows an in place sort
- We can also provide an iterative implementation to avoid the cost of recursive calls
- The partitioning strategy we described is not the only one used in the literature
 - Can build the lower and upper segments from opposite ends and meet in the middle
- Need to analyse the complexity of quick sort

Analysis of Quicksort

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- Worst case? Pivot is maximum or minimum
 - Partitions are of size 0, $n - 1$
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 - $T(n) = n + (n - 1) + \dots + 1$
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- Already sorted array: worst case!

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- Sorting is a rare situation where we can compute this
 - Values don't matter, only relative order is important
 - Analyze behaviour over permutations of $\{1, 2, \dots, n\}$
 - Each input permutation equally likely

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- Sorting is a rare situation where we can compute this
 - Values don't matter, only relative order is important
 - Analyze behaviour over permutations of $\{1, 2, \dots, n\}$
 - Each input permutation equally likely
- Expected running time is $O(n \log n)$

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- Instead, choose pivot position **randomly** at each step

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```

Randomization

- Any fixed choice of pivot allows us to construct worst case input
- Instead, choose pivot position **randomly** at each step
- Expected running time is again $O(n \log n)$

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Iterative quicksort

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 - No recombination of results is required
- Can explicitly keep track of left and right endpoints of each segment to be sorted

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```

Quicksort in practice

- In practice, quicksort is very fast
- Very often the default algorithm used for in-built sort functions
 - Sorting a column in a spreadsheet
 - Library sort function in a programming language

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Summary

- The worst case complexity of quicksort is $O(n^2)$
- However, the average case is $O(n \log n)$
- Randomly choosing the pivot is a good strategy to beat worst case inputs
- Quicksort works in-place and can be implemented iteratively
- Very fast in practice, and often used for built-in sorting functions
 - Good example of a situation when the worst case upper bound is pessimistic

Sorting: Concluding Remarks

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Programming, Data Structures and Algorithms using Python
Week 3

Stable sorting

- Often list values are tuples
 - Rows from a table, with multiple columns / attributes
 - A list of students, each student entry has a roll number, name, marks, ...

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- Suppose students have already been sorted by roll number
- If we now sort by name, will all students with the same name remain in sorted order with respect to roll number?
- **Stability** of sorting is crucial in many applications
- Sorting on column *B* should not disturb sorting on column *A*

Stable sorting

- The quicksort implementation we described is not stable
 - Swapping values while partitioning can disturb existing sorted order

Stable sorting

- The quicksort implementation we described is not stable
 - Swapping values while partitioning can disturb existing sorted order
- Merge sort is stable if we merge carefully
 - Do not allow elements from the right to overtake elements on the left
 - While merging, prefer the left list while breaking ties

Other criteria

- Minimizing data movement
 - Imagine each element is a heavy carton
 - Reduce the effort of moving values around

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- Merge sort is typically used for “external” sorting
 - Database tables that are too large to store in memory all at once
 - Retrieve in parts from the disk and write back
- Other $O(n \log n)$ algorithms exist — heapsort
- Sometimes hybrid strategies are used
 - Use divide and conquer for large n
 - Switch to insertion sort when n becomes small (e.g., $n < 16$)

Lists and Arrays

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Programming, Data Structures and Algorithms using Python
Week 3

Sequences

- Two basic ways of storing a sequence of values
 - Lists
 - Arrays
- What's the difference?

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 - Flexible length
 - Easy to modify the structure
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Sequences

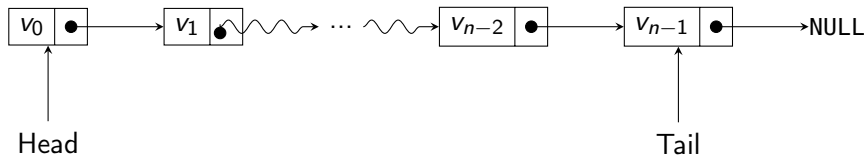
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- Arrays
 - Fixed size
 - Allocate a contiguous block of memory
 - Supports **random access**

Lists

- Typically a sequence of nodes
- Each node contains a value and points to the next node in the sequence
 - “Linked” list

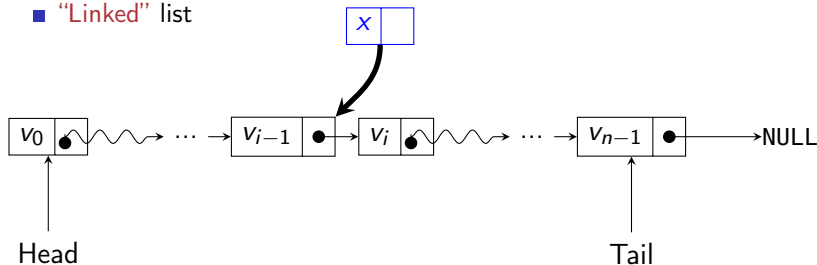
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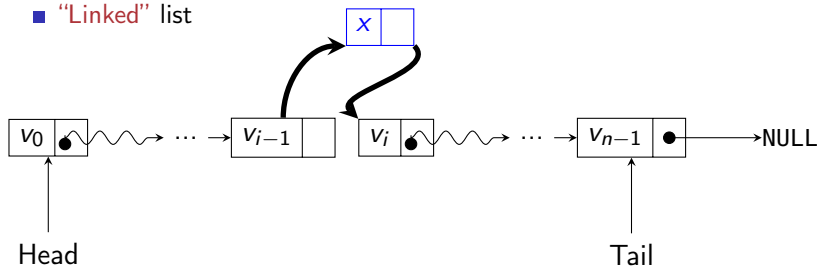
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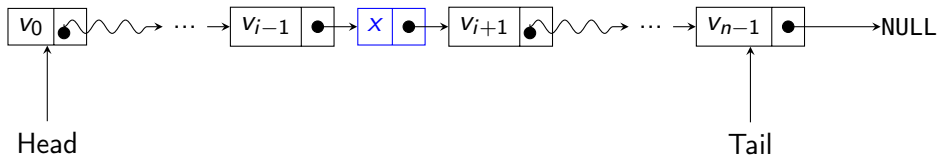
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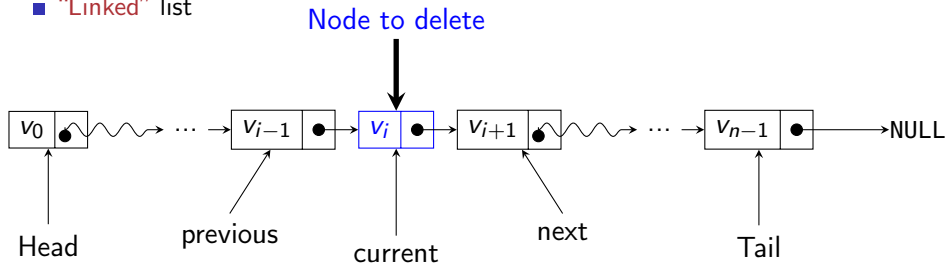
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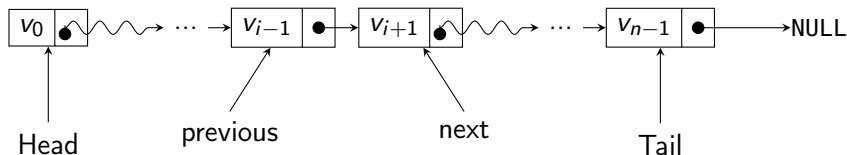
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Arrays

- Fixed size, declared in advance
- Allocate a contiguous block of memory
 - n times the storage for a single value

Index	Value
$A[0]$	v_0
$A[1]$	v_1
\vdots	\vdots
$A[i]$	v_i
\vdots	\vdots
$A[n-1]$	v_{n-1}

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 - Compute offset to $A[i]$ from $A[0]$
 - Accessing $A[i]$ takes constant time, independent of i
- Inserting and deleting elements is expensive
 - Expanding and contracting requires moving $O(n)$ elements in the worst case

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Operations

- Exchange $A[i]$ and $A[j]$
 - Constant time for arrays
 - $O(n)$ for lists
- Delete $A[i]$, insert v after $A[i]$
 - Constant time for lists if we are already at $A[i]$
 - $O(n)$ for arrays
- Need to keep implementation in mind when analyzing data structures
 - For instance, can we use binary search to insert in a sorted sequence?
 - Either search is slow, or insertion is slow, still $O(n)$

Summary

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 - Numpy library provides arrays — are these faster than lists?

Designing a flexible list

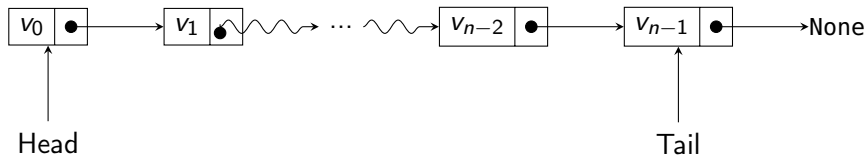
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Programming, Data Structures and Algorithms using Python
Week 3

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Implementing lists in Python

■ Python class `Node`

```
class Node:
    def __init__(self, v = None):
        self.value = v
        self.next = None
        return

    def isempty(self):
        if self.value == None:
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Appending to a list

- Add `v` to the end of list `l`
- If `l` is empty, update `l.value` from `None` to `v`
- If at last value, `l.next` is `None`
 - Point `next` at new node with value `v`
- Otherwise, recursively append to rest of list

```
def append(self,v):  
    # append, recursive  
    if self.isempty():  
        self.value = v  
    elif self.next == None:  
        self.next = Node(v)  
    else:  
        self.next.append(v)  
    return
```

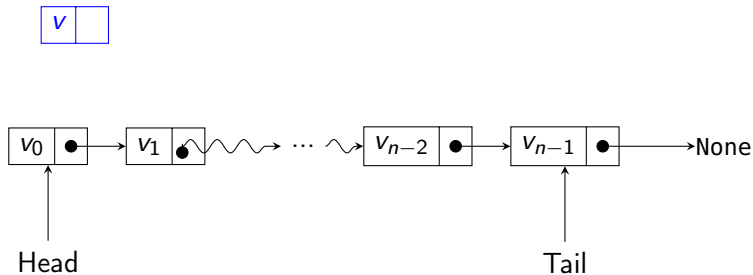
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- Iterative implementation
 - If empty, replace `l.value` by `v`
 - Loop through `l.next` to end of list
 - Add `v` at the end of the list

```
def appendi(self,v):  
    # append, iterative  
    if self.isempty():  
        self.value = v  
        return  
  
    temp = self  
    while temp.next != None:  
        temp = temp.next  
  
    temp.next = Node(v)  
    return
```

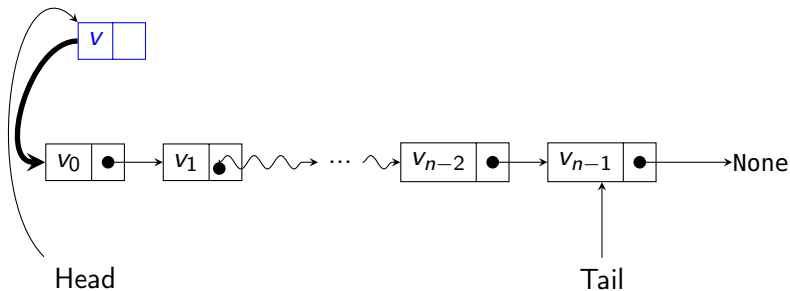
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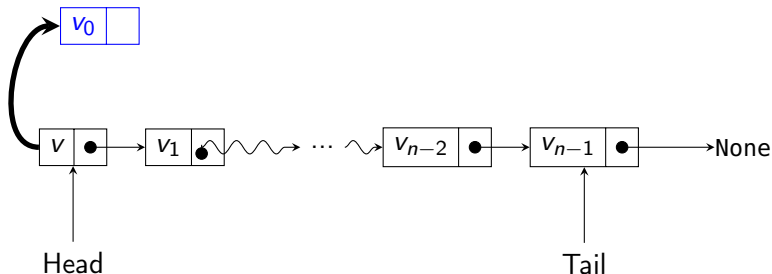
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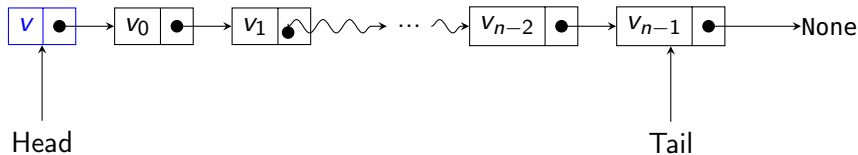
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- Exchange the values v_0, v
- Make new node point to `head.next`
- Make `head.next` point to new node



Appending to a list

- Create a new node with v
- Exchange the values v_0 , v
- Make new node point to `head.next`
- Make `head.next` point to new node

```
def insert(self,v):
    if self.isempty():
        self.value = v
        return

    newnode = Node(v)

    # Exchange values in self and newnode
    (self.value, newnode.value) =
        (newnode.value, self.value)

    # Switch links
    (self.next, newnode.next) =
        (newnode, self.next)

    return
```


Delete a value v

- Remove first occurrence of v
- Scan list for first v — look ahead at next node
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 - Instead, copy the second node value to head
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- Recursive implementation

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def delete(self,v):  
    # delete, recursive  
  
    if self.isempty():  
        return  
  
    if self.value == v:  
        self.value = None  
        if self.next != None:  
            self.value = self.next.value  
            self.next = self.next.next  
        return  
    else:  
        if self.next != None:  
            self.next.delete(v)  
            if self.next.value == None:  
                self.next = None  
  
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- Recursive implementation
- Exercise: write an iterative version

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            self.next = self.next.next  
        return  
    else:  
        if self.next != None:  
            self.next.delete(v)  
            if self.next.value == None:  
                self.next = None  
  
    return
```

Summary

- Use a linked list of nodes to implement a flexible list
- Append is easy
- Insert requires some care, cannot change where the head points to
- When deleting, look one step ahead to bypass the node to be deleted

Lists and Arrays in Python

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Programming, Data Structures and Algorithms using Python
Week 3

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- Keep track of the last position of the list in the array
 - `l.append()` and `l.pop()` are constant time, amortised — $O(1)$
 - Insertion/deletion require time $O(n)$

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 - Double the size if the list overflows the array
- Keep track of the last position of the list in the array
 - `l.append()` and `l.pop()` are constant time, amortised — $O(1)$
 - Insertion/deletion require time $O(n)$
- Effectively, Python lists behave more like arrays than lists

Arrays vs lists in Python

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- Mutability **aliases** different values
- Instead, use list comprehension

```
zeromatrix = [ [ 0 for i in range(3) ] for j in range(3) ]
```


Numpy arrays

- The Numpy library provides arrays as a basic type

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- Can operate on a matrix as a whole

- `C = 3*A + B`

- `C = np.matmul(A,B)`

- Very useful for data science

Summary

- Python lists are not implemented as flexible linked structures
- Instead, allocate an array, and double space as needed
- Append is cheap, insert is expensive
- Arrays can be represented as multidimensional lists, but need to be careful about mutability, aliasing
- Numpy arrays are easier to use

Implementing dictionaries

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Programming, Data Structures and Algorithms using Python
Week 3

- An array/list allows access through positional indices

Dictionary

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- How is a dictionary implemented?

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 - SHA-256 is an industry standard hashing function whose range is 256 bits
 - Use to hash large files — avoid uploading duplicates to cloud storage

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- Dictionary keys in Python must be immutable
 - If value changes, hash also changes!

Summary

- A dictionary is implemented as a hash table
 - An array plus a hash function
- Creating a good hash function is important (and hard!)
- Need a strategy to deal with collisions
 - Open addressing/closed hashing — probe for free space in the array
 - Open hashing — each slot in the hash table points to a list of key-value pairs
 - Many heuristics/optimizations possible for dea