

# 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
  - Recursive calls and returns are expensive

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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!

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- 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

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- Choose a pivot element
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## High level view of quicksort

- Input list

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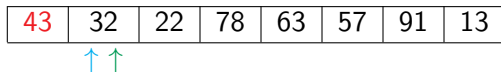
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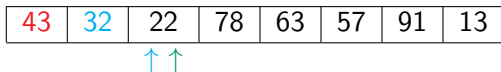
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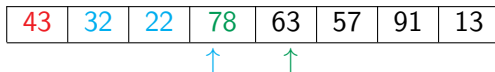
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- Maintain two indices to mark the end of the **Lower** and **Upper** segments
- After partitioning, exchange the pivot with the last element of the **Lower** segment

# 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
  - If it is less than or equal to the pivot, exchange with the first element in **Upper**. This extends **Lower** and shifts **Upper** by one position.

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def quicksort(L,l,r):  # Sort L[l:r]
    if (r - l <= 1):
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    (pivot,lower,upper) = (L[l],l+1,l+1)
    for i in range(l+1,r):
        if L[i] > pivot:  # Extend upper segment
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        else: # Exchange L[i] with start of upper segment
            (L[i], L[lower]) = (L[lower], L[i])
            # Shift both segments
            (lower,upper) = (lower+1,upper+1)
    # Move pivot between lower and upper
    (L[l],L[lower-1]) = (L[lower-1],L[l])
    lower = lower-1
    # Recursive calls
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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,  $n - 1$
  - $T(n) = T(n - 1) + n$
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- Already sorted array: worst case!

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  - Each input permutation equally likely
- Expected running time is  $O(n \log n)$

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- Expected running time is again  $O(n \log n)$

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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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- 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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- **Stability** of sorting is crucial in many applications
- Sorting on column *B* should not disturb sorting on column *A*

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# 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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- 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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  - Lists
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  - Easy to modify the structure
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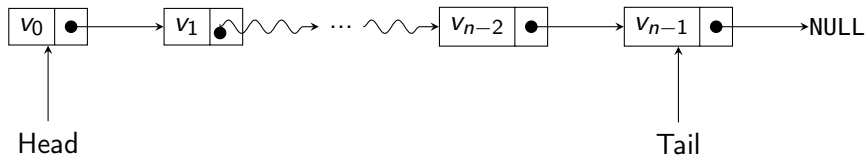
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  - Values are scattered in memory
- Arrays
  - Fixed size
  - Allocate a contiguous block of memory
  - Supports **random access**

# Lists

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

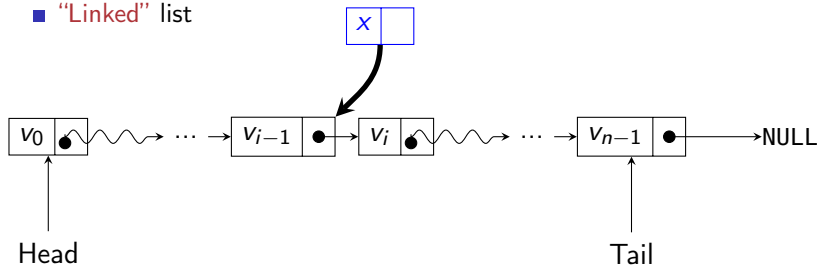
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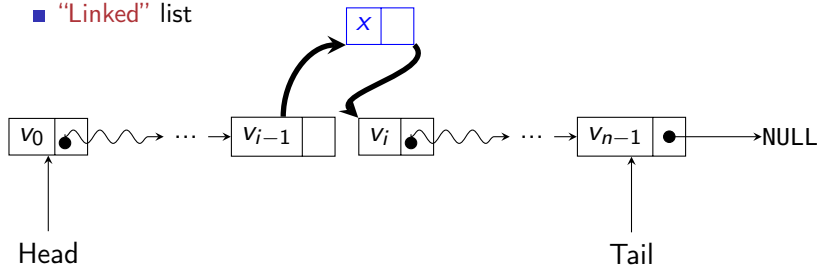
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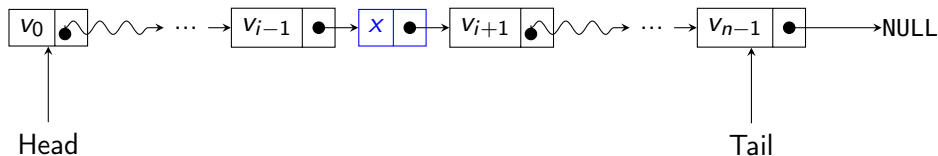
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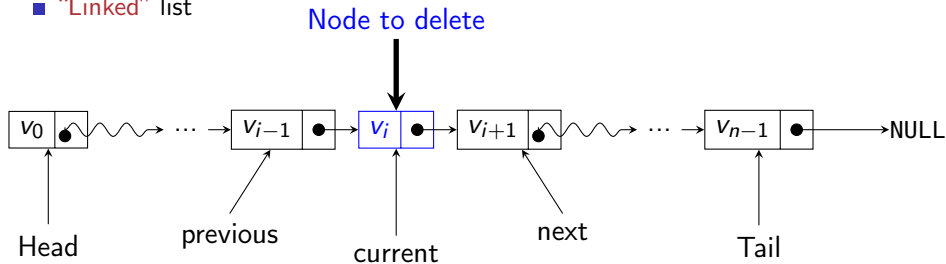
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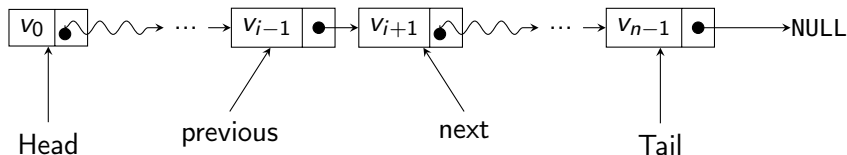
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- Need to follow links to access  $A[i]$ 
  - Takes time  $O(i)$

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- 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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  - 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  $O(n)$

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- Arrays support random access but are difficult to expand, contract
- Algorithm analysis needs to take into account the underlying implementation
- How does it work in Python?
  - Is the built-in list type in Python really a “linked” list?
  - Numpy library provides arrays — are these faster than lists?

# Designing a flexible list

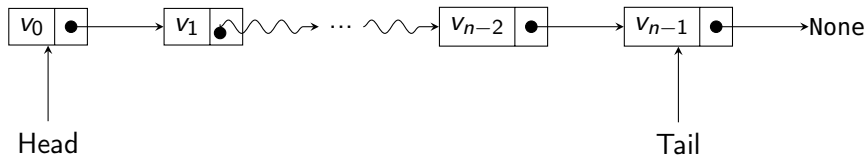
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Programming, Data Structures and Algorithms using Python  
Week 3

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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:
            return(True)
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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
```

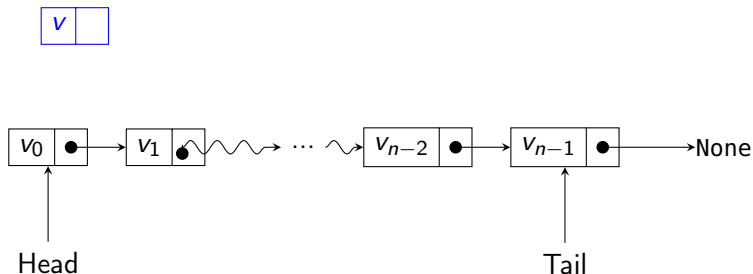
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- Iterative implementation
  - If empty, replace `l.value` by `v`
  - Loop through `l.next` to end of list
  - Add `v` at the end of the list

```
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
```

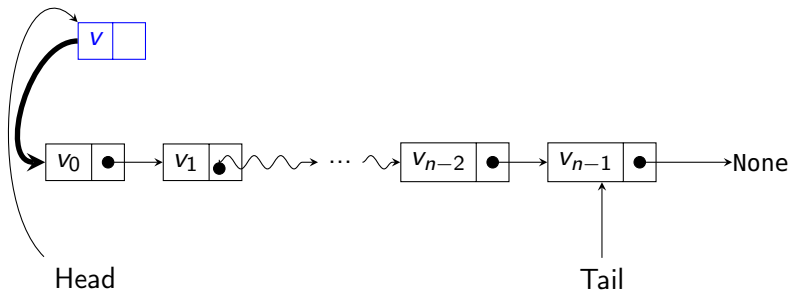
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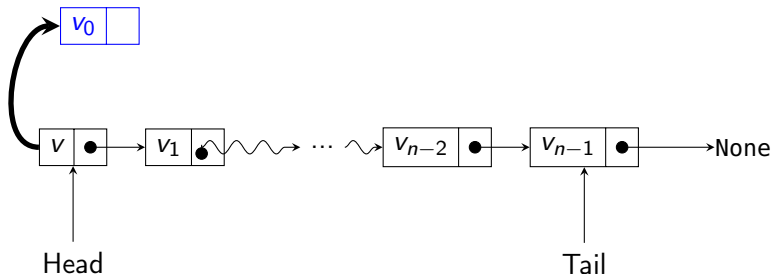
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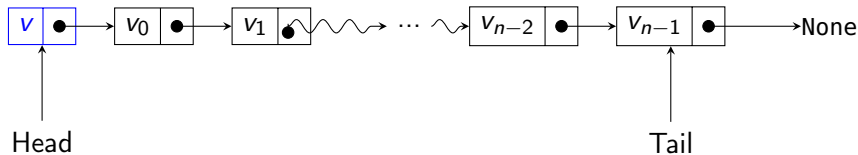
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def insert(self,v):  
    if self.isempty():  
        self.value = v  
        return  
  
    newnode = Node(v)  
  
    # Exchange values in self and newnode  
    (self.value, newnode.value) =  
        (newnode.value, self.value)  
  
    # Switch links  
    (self.next, newnode.next) =  
        (newnode, self.next)  
  
    return
```



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- Recursive implementation

```
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  
  
    return
```

# Delete a value $v$

- Remove first occurrence of  $v$
- Scan list for first  $v$  — look ahead at next node
- If next node value is  $v$ , bypass it
- Cannot bypass the first node in the list
  - Instead, copy the second node value to head
  - Bypass second node
- Recursive implementation
- Exercise: write an iterative version

```
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:  
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```

# 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

Madhavan Mukund

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Programming, Data Structures and Algorithms using Python  
Week 3

# Lists and arrays in Python

- Sequences can be stored as lists or arrays
- Lists are flexible but accessing an element is  $O(n)$
- Arrays support random access but are difficult to expand, contract
- Algorithm analysis needs to take into account the underlying implementation
- How does it work in Python?
  - Is the built-in list type in Python really a “linked” list?
  - Numpy library provides arrays — are these faster than lists?

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  - `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) ]
```



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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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# Dictionary

- An array/list allows access through positional indices
- A dictionary allows access through arbitrary **keys**
  - A collection of key-value pairs
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- An array/list allows access through positional indices
- A dictionary allows access through arbitrary **keys**
  - A collection of key-value pairs
  - Random access — access time is the same for all keys
- 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