

COMP2521 – Course Notes



Contents

Week 1	3
1.1: Introduction	3
1.2: COMP1511 Code Gap	4
COMP2521 Style	4
Switch Statements	4
Ternary Operator	4
1.3: Compilation and Makefiles	5
Compilers.....	5
Makefiles	6
1.4: Recursion (Linked List)	7
Recursion	7
Pattern for a Recursive Function	7
Week 2	10
2.1: Analysis of Algorithms	10
Empirical Analysis	10
Theoretical Analysis	10
Primitive Operations.....	10
Counting Primitive Operations	10
Big-Oh Notation.....	11
Big-Oh Rules.....	11
Asymptotic Analysis of Algorithms	11
Binary Search	12
Exercises	13
Complexity Classes	13
Generate and Test Algorithms.....	14
2.2: Abstract Data Types	15
DTs, ADTs, GADTs	15
Interface/Implementation	15
Collections	15
Week 3	17
3.1: Tree, Search Trees	17
Searching	17

Tree Data Structures.....	17
Binary Tree.....	18
Other Special Kinds of Tree.....	19
Binary Search Tree	19
Properties of Binary Search Trees.....	19
Insertion into BST	20
Representing BSTs	21
Tree Algorithms	22
Tree Traversal.....	22
Joining Two Trees	24
Deletion from BSTs	25
3.2: Function Pointers in C	27
Week 4	29
4.1: Balancing Search Trees.....	29

Week 1

1.1: Introduction

- This course will get you to:
 - Think like a computer scientist
 - Know fundamental techniques/structures
 - Able to reason about applicability/effectiveness
 - Able to analyse behaviour/correctness of programs

1.2: COMP1511 Code Gap

COMP2521 Style

- Allowed to use all C control structures:
 - if
 - switch
 - while
 - for
 - break
 - continue
- Other constructs:
 - Assignment statements in expressions
 - Use conditional statements, but use: $x = c ? e1 : e2$ (ternary operator) with care
 - Functions may have multiple return statements

Switch Statements

- This:

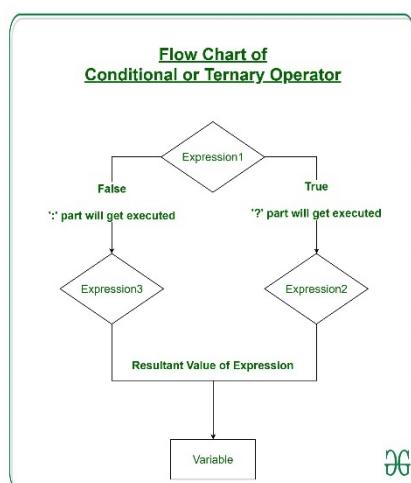
```
if (v == C1) {  
    S1;  
} else if (v == C2) {  
    S2;  
}  
...  
else if (v == Cn) {  
    Sn;  
}  
else {  
    Sn+1;  
}
```

can become this:

```
switch (v) {  
case C1:  
    S1; break;  
case C2:  
    S2; break;  
...  
case Cn:  
    Sn; break;  
default:  
    Sn+1;  
}
```

- Note: `break` is critical, otherwise it will fall through to the next case

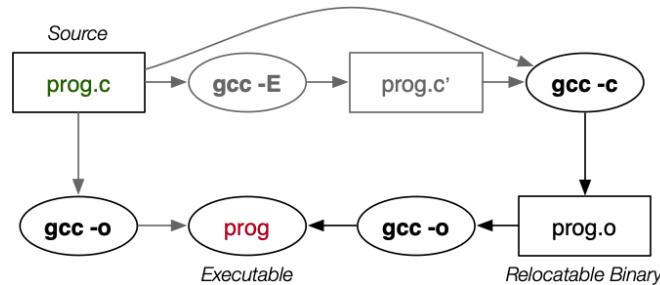
Ternary Operator



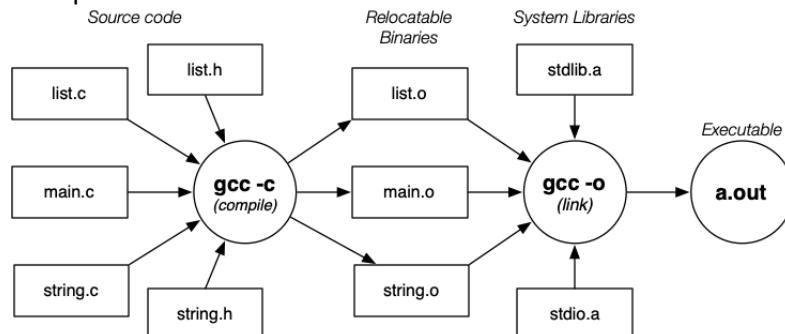
1.3: Compilation and Makefiles

Compilers

- Compilers are programs that:
 - convert program source code to executable form
 - ‘executable’ might be machine code or byte code
- The Gnu C Compiler (gcc)
 - Applies source-to-source transformation (pre-processor)
 - Compiles source code to produce object files
 - Links object files and libraries to produce executables
- **clang** is an alternative C compiler
- Stages in C compilation: pre-processing, compilation, linking



- When we have multiple C files involved:



- Compilation/linking with **gcc**

```
gcc -c Stack.c
```

produces Stack.o, from Stack.c and Stack.h

```
gcc -c bracket.c
```

produces bracket.o, from bracket.c and Stack.h

```
gcc -o rbt bracket.o Stack.o
```

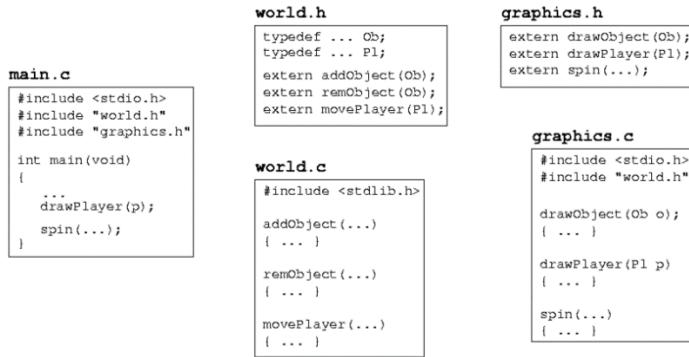
links bracket.o, Stack.o and libraries

producing executable program called rbt

- **gcc** is a multi-purpose tool, it compiles (-c), links, makes executables (-o)

Makefiles

- Compiling large systems is very complex
- The **make** command assists by allowing:
 - programmers to document dependencies in code
 - minimal re-compilation, based on dependencies
- For example:



- **make** is driven by dependencies given in the **Makefile**

- A dependency specifies:

```
target : source1 source2 ...
          commands to build target from sources
```

for example:

```
game : main.o graphics.o world.o
      gcc -o game main.o graphics.o world.o
```

- Target is rebuilt if it is older than any other source

- For example:

```
game : main.o graphics.o world.o
      gcc -o game main.o graphics.o world.o
```

```
main.o : main.c graphics.h world.h
      gcc -Wall -Werror -c main.c
```

```
graphics.o : graphics.c world.h
      gcc -Wall -Werror -c graphics.c
```

```
world.o : world.c
      gcc -Wall -Werror -c world.c
```

- If make arguments are targets, build just those targets:

```
make world.o
```

```
gcc -Wall -Werror -c world.c
```

- If no arguments, build first target in the Makefile

```
make
```

```
gcc -Wall -Werror -c main.c
gcc -Wall -Werror -c graphics.c
gcc -Wall -Werror -c world.c
gcc -o game main.o graphics.o world.o
```

- Makefiles can also contain variables

- e.g. **CC**, **CFLAGS**, **LDFLAGS**
- can easily change which C compiler used, etc

- make has rules, which allow it to interpret e.g.

```
Stack.o : Stack.c Stack.h
```

As

```
Stack.o : Stack.c Stack.h
      $(CC) $(CFLAGS) -c Stack.c
```

1.4: Recursion (Linked List)

Recursion

- Recursion is a programming pattern where a function calls itself

Pattern for a Recursive Function

- Base case(s)
 - Situations when we do not call the same function (no recursive call), because the problem can be solved easily without recursion
 - All recursive cases eventually lead to one of the base cases
- Recursive case
 - We call the same function for a problem with smaller size
 - Decrease in a problem size eventually leads to one of the base cases
- Sum of list data fields

```
// return sum of list data fields: using recursive call

int sum(struct node *head) {
    if (head == NULL) { ← Base case
        return 0;
    }
    return head->data + sum(head->next); ← Recursive case,
                                                Recursive call for a
                                                smaller problem
                                                (size-1)
}
```

- Length of linked list

```
// return count of nodes in list

int length(struct node *head) {
    if (head == NULL) { ← Base case
        return 0;
    }
    return 1 + length(head->next); ← Recursive call
}
```

The diagram illustrates the recursive call for the length of a linked list. It shows three levels of recursion:

- 1st Recursive call (returns 3):** The initial call with head pointing to a node with data 13. This call returns 1 + length(head->next).
- 2nd Recursive call (returns 2):** The call from the first level with head pointing to a node with data 17. This call returns 1 + length(head->next).
- 3rd Recursive call (returns 1):** The call from the second level with head pointing to a node with data 42. This call returns 1 + length(head->next).

The base case is Head == NULL returns 0.

- Last node of a linked list

```
struct node *last(struct node *head) {
    // list is empty
    if(head == NULL) {
        return NULL;
    }
    // found the last node! return it.
    else if (head->next == NULL) {
        return head;
    }
    // return last node from the rest of the list
    // using a recursion
    else {
        return last(head->next);
    }
}
```

The diagram illustrates the recursive call for finding the last node of a linked list. It shows three levels of recursion:

- 1st recursive call (returns head of 3rd):** The initial call with head pointing to a node with data 13. This call returns last(head->next).
- 2nd recursive call (returns head of 3rd call):** The call from the first level with head pointing to a node with data 17. This call returns last(head->next).
- 3rd Recursive call (returns head of this list):** The call from the second level with head pointing to a node with data 42. This call returns last(head->next).

The base case is head->next == NULL returns head of 3rd call.

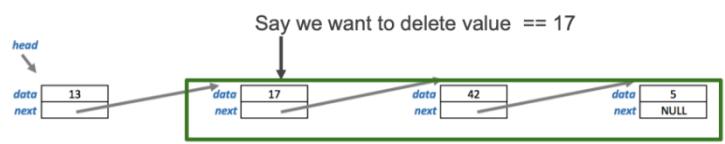
- Find node using recursion

```
// return pointer to first node with specified data value
// return NULL if no such node

struct node *find_node(struct node *head, int data) {
    // empty list, so return NULL
    if (head == NULL) {
        return NULL;
    }
    // Data at "head" is same as the "data" we are searching,
    // Found the node! so return head.
    else if (head->data == data) {
        return head;
    }
    // Find "data" in the rest of the list, using recursion,
    // return whatever answer we get from the recursion
    else {
        return find_node(head->next, data); ← Recursive call
    }
}
```

- Delete a node from a linked list

```
// Delete a Node from a List: Recursive
struct node *deleteR(struct node *list, int value) {
    if (list == NULL) {
        fprintf(stderr, "warning: value %d is not in list\n", value);
    } else if (list->data == value) {
        struct node *tmp = list;
        list = list->next;           // remove first item
        free(tmp);
    } else {
        list->next = deleteR(list->next, value); ← Recursive call
    }
    return list;
}
```



- Insert a node to a linked list

```
// Insert a Node into an Ordered List: recursive
struct node *insertR(struct node *list, int value) {
    if (list == NULL || list->data >= value) {

        struct node *newHead = create_node(value, NULL);
        newHead->next = list;
        return newHead;

        // Alternatively, in one line
        // return create_node(value, list); ← Recursive call
    }

    list->next = insertR(list->next, value); ← Recursive call
}

return list;
}
```



- Print Python List using Recursion

```
// print contents of list in Python syntax

void print_list(struct node *head) {
    printf("[");
    if (head != NULL) {
        print_list_items(head);
    }
    printf("]");
}

void print_list_items(struct node *head) {
    printf("%d", head->data);
    if (head->next != NULL) {
        printf(", ");
        print_list_items(head->next);
    }
}
```

Recursive function → **print_list_items**(head);

Recursive call → **print_list_items**(head->next);

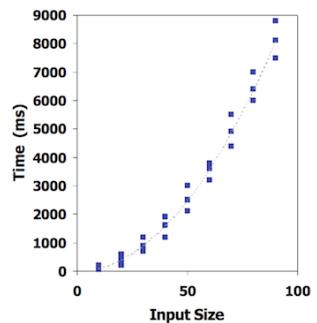
Week 2

2.1: Analysis of Algorithms

- An algorithm is a step-by-step procedure
 - For solving a problem
 - In a finite amount of time
- Most algorithms map input to output
 - Running time typically grows with input size
 - Average time is often difficult to determine
 - Focus on worst case running time

Empirical Analysis

- Write a program that implements an algorithm
- Run program with inputs of varying size and composition
- Measure actual running time
- Plot the results
- Limitations:
 - Requires implementing the algorithm, which may be difficult
 - Results may not be indicative of running time on other inputs
 - Same hardware and OS must be used to compare two algorithms



Theoretical Analysis

- Uses a high-level description of the code instead of implementation (pseudocode)
- Characterises running time as a function of the input size, n
- Takes into account all possible inputs
- Allows us to evaluate the speed of an algorithm independent of the hardware/software environment

Primitive Operations

- Basic computations performed by an algorithm
- Identifiable in pseudocode
- Largely independent of the programming language
- Examples:
 - Evaluating an expression
 - Indexing into an array
 - Calling/returning from a function

Counting Primitive Operations

```
arrayMax(A) :  
| Input array A of n integers  
| Output maximum element of A  
  
| currentMax=A[0]           1  
| for all i=1..n-1 do      n+(n-1)  
| | if A[i]>currentMax then 2(n-1)  
| | | currentMax=A[i]        n-1  
| | end if  
| end for  
| return currentMax         1  
                           -----  
                           Total   5n-2
```

- The time complexity is $O(n)$

```

matrixProduct(A,B) :
|   Input n×n matrices A, B
|   Output n×n matrix A·B
|
|   for all i=1..n do           2n+1
|       for all j=1..n do       n(2n+1)
|           C[i,j]=0            n2
|           for all k=1..n do    n2(2n+1)
|               C[i,j]=C[i,j]+A[i,k] · B[k,j]  n3 · 5
|           end for
|       end for
|   end for
|   return C                   1
|
|                                         -----
|                                         Total 7n3+4n2+3n+2

```

- The order is $O(n^3)$

Big-Oh Notation

- Big-Oh notation gives an upper bound on the growth rate of a function
- Use big-Oh to rank functions according to their rate of growth

Big-Oh Rules

- If $f(n)$ is a polynomial of degree $d \rightarrow f(n) = O(n^d)$
 - Lower-order terms are ignored
 - Constant factors are ignored

Asymptotic Analysis of Algorithms

- Asymptotic analysis of algorithms determines running time in big-Oh notation:
 - Find worst-case number of primitive operations as a function of input size
 - Express this function using big-Oh notation
- Example:
 - Algorithm arrayMax executes at most

```

prefixAverages1(X) :
|   Input array X of n integers
|   Output array A of prefix averages of X
|
|   for all i=0..n-1 do          O(n)
|       s=X[0]                     O(n)
|       for all j=1..i do         O(n2)
|           s=s+X[j]              O(n2)
|       end for
|       A[i]=s/(i+1)             O(n)
|   end for
|   return A                   O(1)

```

- The time complexity is $O(n^2)$

Binary Search

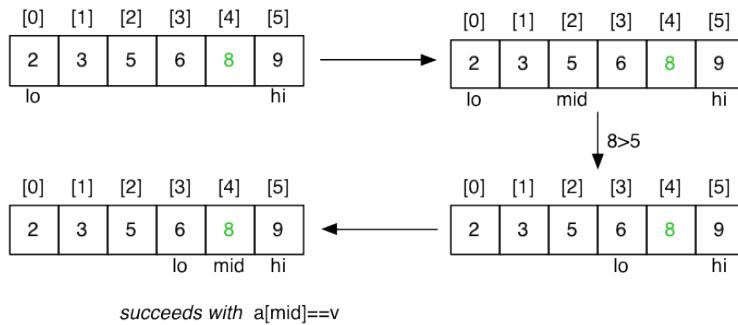
- Given a sorted array

```

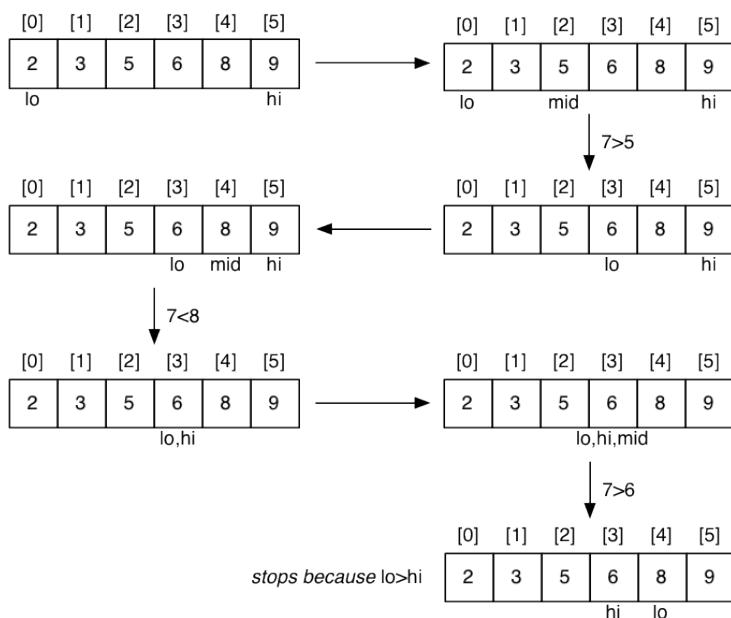
search(v,a,lo,hi):
| Input value v
|         array a[lo..hi] of values
| Output true if v in a[lo..hi]
|         false otherwise
|
|     mid=(lo+hi)/2
|     if lo>hi then return false
|     if a[mid]==v then
|         return true
|     else if a[mid]<v then
|         return search(v,a,lo,mid+1,hi)
|     else
|         return search(v,a,lo,mid-1)
|     end if

```

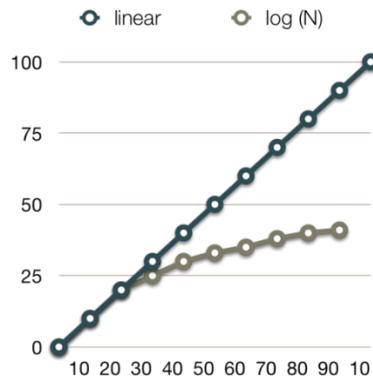
- Successful search for a value of 8:



- Unsuccessful search for a value of 7:



- Cost analysis:
 - C_i = #calls to **search()** for array of length i
 - for best case, $C_n = 1$
 - for $a[i..j]$, $j < i$ (length=0)
 - $C_0 = 0$
 - for $a[i..j]$, $i \leq j$ (length=n)
 - $C_n = 1 + C_{n/2} \Rightarrow C_n = \log_2 n$
- Logarithmic complexity is good:



Exercises

```
splitList(L):
| Input non-empty linked list L
| Output L split into two halves
|
| // use slow and fast pointer to traverse L
| slow=head(L), fast=head(L).next
| while fast=NULL  $\wedge$  fast.next=NULL do
|   slow=slow.next, fast=fast.next.next // advance pointers
| end while
| cut L between slow and slow.next
• Time complexity is  $O(|L|)$ 
```

```
binaryConversion(n):
| Input positive integer n
| Output binary representation of n on a stack
|
| create empty stack S
| while n>0 do
|   push (n mod 2) onto S
|   n=n/2
| end while
| return S
```

- Time complexity is $O(\log n)$

Complexity Classes

- Problems in Computer Science
 - Some have polynomial worst-case performance
 - Some have exponential worst case performance
- Classes of problems:
 - P = problems for which an algorithm can compute answer in polynomial time
 - NP = includes problems for which no P algorithm is known
- Beware: NP stands for ‘nondeterministic, polynomial time’

- Computer Science jargon for difficulty:
 - tractable: have a polynomial-time algorithm (useful in practice)
 - intractable: no tractable algorithm is known (feasible for only small n)
 - non-computable: no algorithm can exist

Generate and Test Algorithms

- In scenarios where
 - It is simple to test whether a given state is a solution
 - It is easy to generate new states (preferably likely solutions)
- Then a generate and test strategy can be used
- It is necessary that states are generated systematically
 - So that we are guaranteed to find a solution, or know that none exists
 - Some randomised algorithms do not require this

```
isPrime(n):
| Input natural number n
| Output true if n prime, false otherwise
|
| for all i=2..n-1 do          // generate
| | if n mod i = 0 then        // test
| | | return false            // i is a divisor => n is not prime
| | end if
| end for
| return true                 // no divisor => n is prime
```

- The complexity is $O(n)$

```
subsetsum1(A, k):
| Input set A of n integers, target sum k
| Output true if  $\sum_{b \in B} b = k$  for some  $B \subseteq A$ 
|           false otherwise
|
| for s=0.. $2^n - 1$  do
| | if k =  $\sum_{(i^{\text{th}} \text{ bit of } s \text{ is } 1)} A[i]$  then
| | | return true
| | end if
| end for
| return false
```

- The complexity is $O(2^n)$

```
subsetsum2(A, n, k):
| Input array A, index n, target sum k
| Output true if some subset of  $A[0..n-1]$  sums up to k
|           false otherwise
|
| if k=0 then
| | return true // empty set solves this
| else if n=0 then
| | return false // no elements => no sums
| else
| | return subsetsum2(A, n-1, k-A[n-1])  $\vee$  subsetsum2(A, n-1, k)
| end if
```

- The complexity is $O(2^n)$
- The above two examples are NP-complete problems (intractable)

2.2: Abstract Data Types

- A data type is:
 - a set of values (atomic or structured values)
 - a collection of operations on those values
- An abstract data type is:
 - an approach to implementing data types
 - separates interface from implementation
 - users of the ADT see only the interface
 - builders of the ADT provide an implementation

DTs, ADTs, GADTs

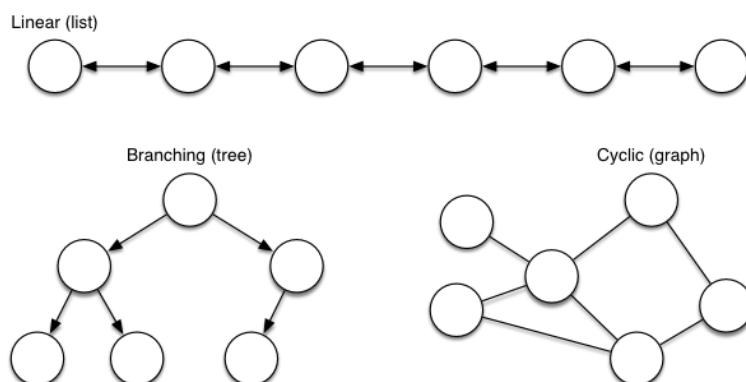
- We want to distinguish:
 - DT = (non-abstract) data type (e.g. C strings)
 - Internals of data structures are visible (e.g. char s[10];)
 - ADT = abstract data type (e.g. C files)
 - Can have multiple instances (e.g. Set a, b, c;)
 - GADT = generic (polymorphic) abstract data type
 - Can have multiple instances (e.g. Set <int> a, b, c;)
 - Can have multiple types (e.g. Set <int> a; Set <char> b;)
 - Not available natively in the C language

Interface/Implementation

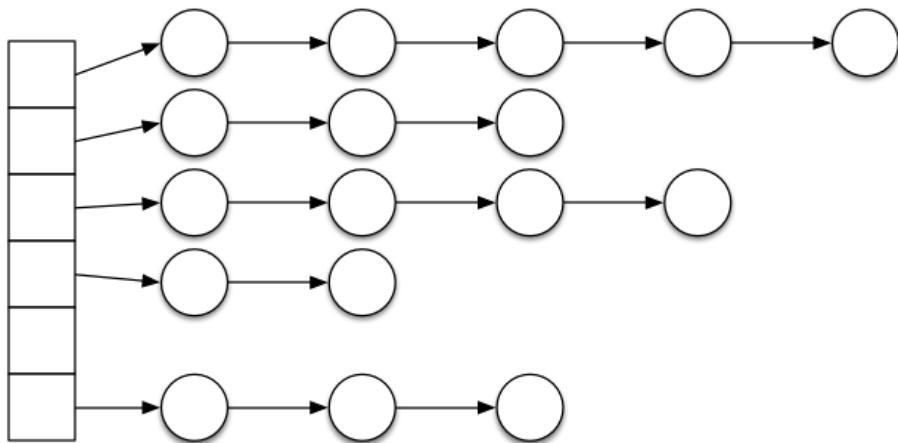
- ADT interface provides
 - A user-view of the data structure (e.g. FILE *)
 - Function signatures (prototypes) for all operations
 - Semantics of operations (via documentation)
 - A contract between ADT and its clients
- ADT implementation gives
 - Concrete definition of the data structures
 - Definition of functions for all operations

Collections

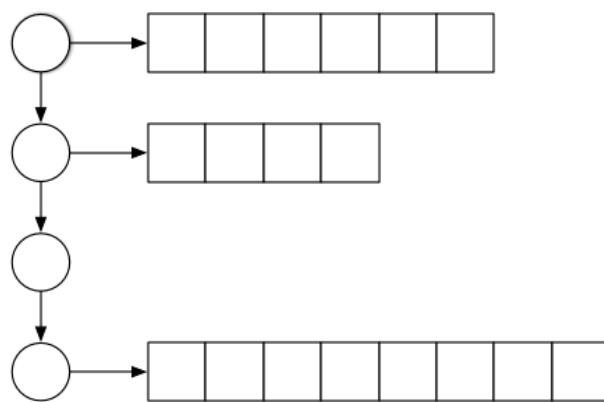
- Many of the ADTs we deal with
 - Consist of a collection of items
 - Where each item may be a simple type or an ADT
 - And items often have a key (to identify them)
- Collections may be categorised by:
 - Structure: Linear (list), branching (tree), cyclic (graph)
 - Usage: Set, matrix, stack, queue, search-tree, dictionary



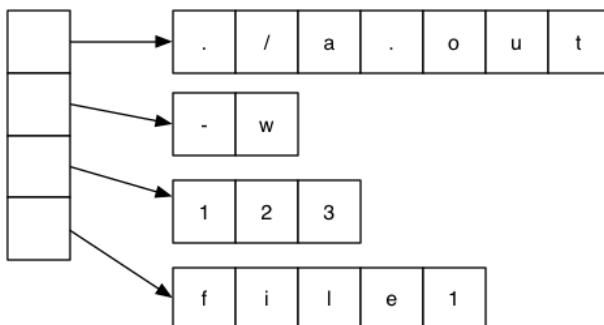
- Or even a hybrid structure:



Or this:



Or this:



- Typical operations on collections:

- Create an empty collection
- Insert one item into the collection
- Remove one item from the collection
- Find an item in the collection
- Check properties of the collection (size, empty?)
- Drop the entire collection
- Display the collection

Week 3

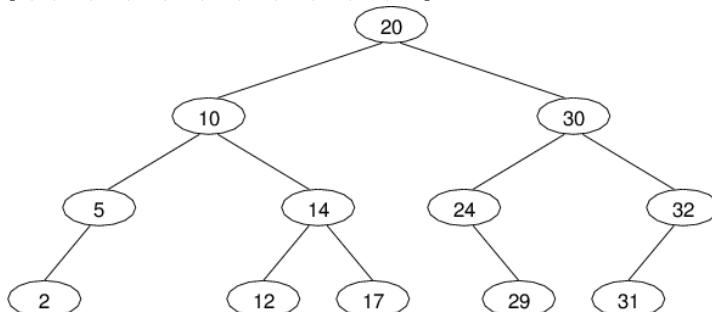
3.1: Tree, Search Trees

Searching

- Search is an extremely common application in computing
 - Given a (large) collection of items and a key value
 - Find the item(s) in the collection containing that key
 - item = (key, val₁, val₂, ...) (i.e. a structured data type)
 - key = value used to distinguish items (e.g. student ID)
- Applications: Google, databases, ...
- Many approaches have been developed for the ‘search’ problem
- Different approaches determined by properties of data structures:
 - Arrays: Linear, random-access, in-memory
 - Linked-lists: Linear, sequential access, in-memory
 - Files: Linear, sequential access, external
- Search costs:

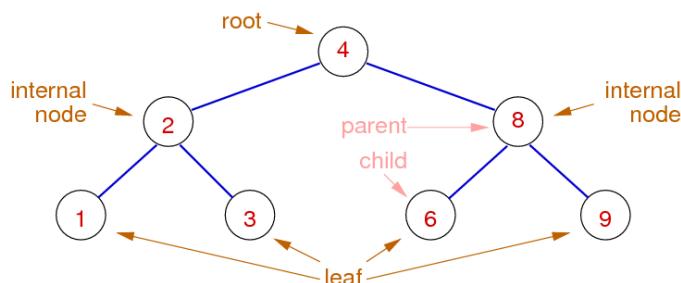
	Array	List	File
Unsorted	O(n) (linear scan)	O(n) (linear scan)	O(n) (linear scan)
Sorted	O(log n) (binary search)	O(n) (linear scan)	O(log n) (lseek,lseek,...)

- Maintaining arrays and files in sorted order is costly
- Search trees are efficient to search but also easy efficient to maintain
- For example, the following tree corresponds to the sorted array:
[2,5,10,12,14,17,20,24,29,30,31,32]



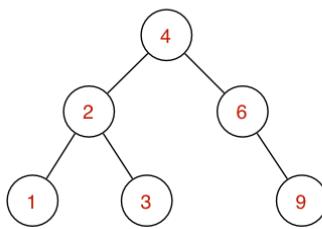
Tree Data Structures

- Trees are connected graphs
 - With nodes and edges (called links), but no cycles (no ‘up-links’)
 - Each node contains a data value (or key+data)
 - Each node has links to <= 2 other child notes

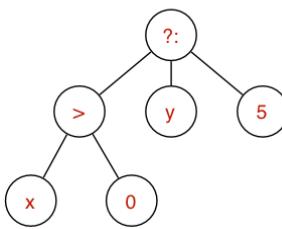


- Trees are used in many contexts, e.g.:
 - Representing hierachal data structures (e.g. expressions)
 - Efficient searching (e.g. sets, symbol tables, ...)

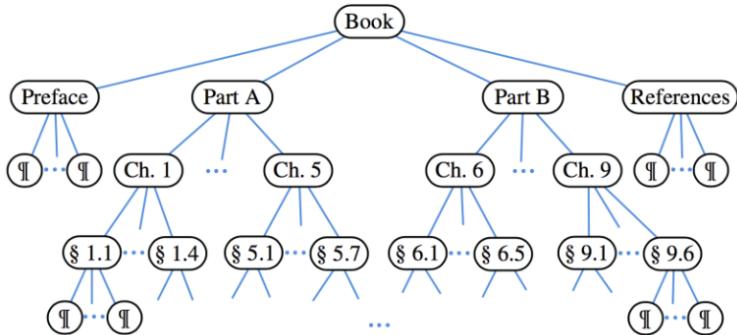
Search Tree



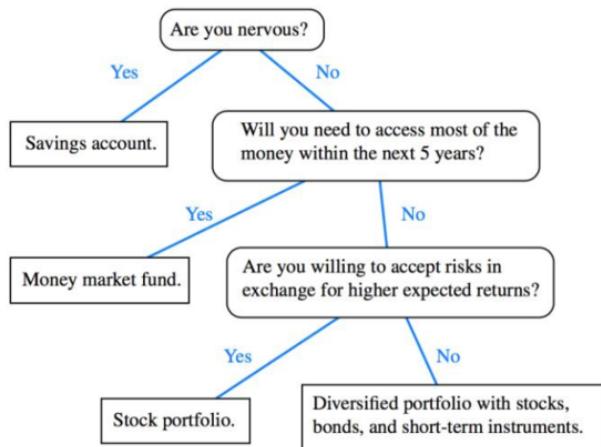
Expression Tree



- Real word example:

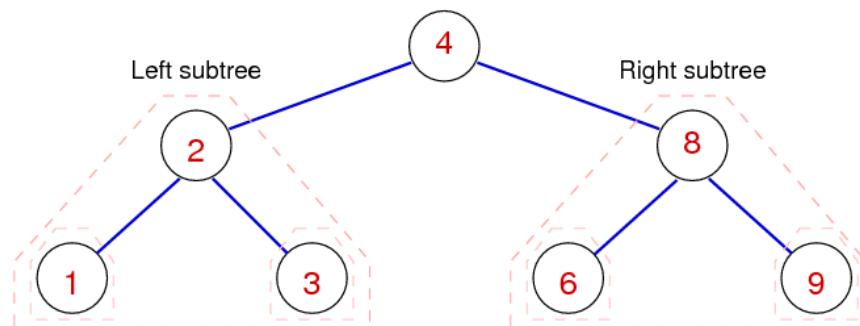


- Real world example: A decision tree



Binary Tree

- A binary tree is either:
 - Empty or
 - Consists of a node, with two subtrees
 - Node contains a value (typically key+data)
 - Left and right subtrees are binary trees (recursive)



Other Special Kinds of Tree

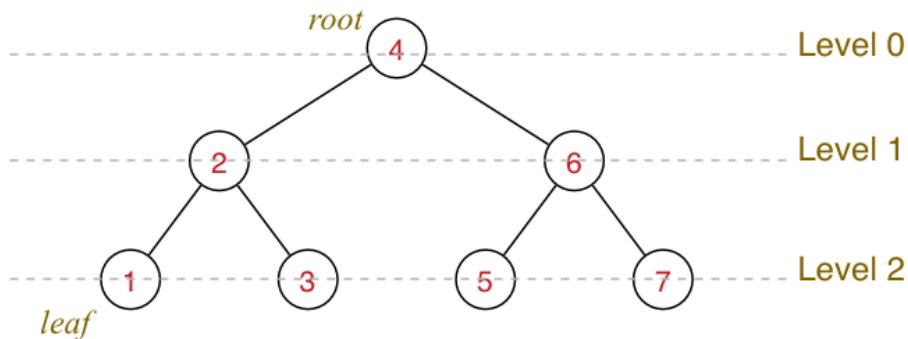
- m -ary tree: Each internal node has exactly m children
- B-tree: Each internal node has $n/2 \leq \#children \leq n$
- Ordered tree: all left values $< root$, all right values $> root$
- Balanced tree: has minimal height for a given number of nodes
- Degenerate tree: has maximal height for a given number of nodes

Binary Search Tree

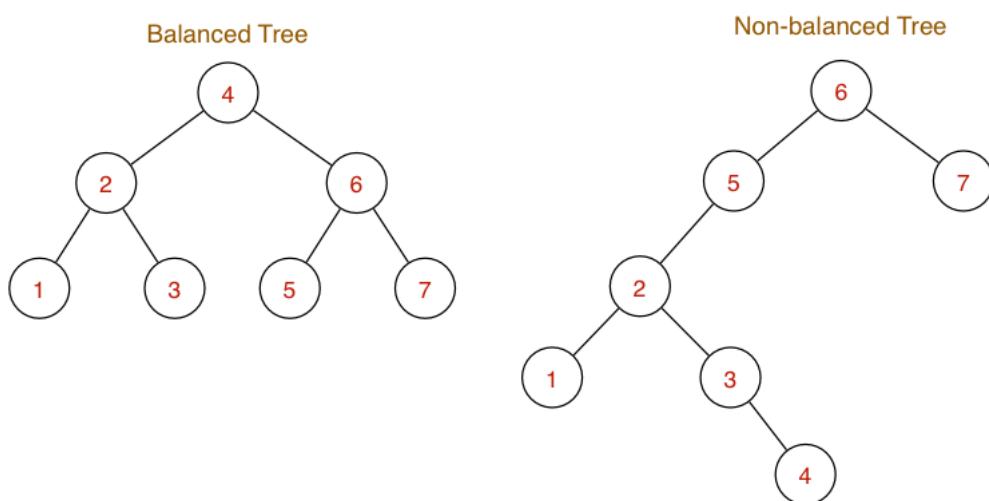
- Binary search trees (or BSTs) are ordered trees:
 - Each node is the root of 0, 1 or 2 subtrees
 - All values in any left subtree are less than root
 - All values in any right subtree are greater than root
 - These properties applies over all nodes in the trees

Properties of Binary Search Trees

- This tree has a height (or depth) of 2



- Ordered tree: For all nodes, $\max(\text{left subtree}) < \text{root} < \min(\text{right subtree})$
- Perfectly-balanced tree: For all nodes,
 $|\text{number_of_nodes(left subtree)}| - |\text{number_of_nodes(right subtree)}| \leq 1$
- Height balanced tree: For all nodes,
 $|\text{height(left subtree)}| - |\text{height(right subtree)}| \leq 1$
- Time complexity of a balanced tree algorithm is typically $O(\text{height})$



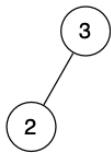
Insertion into BST

- Example 1:

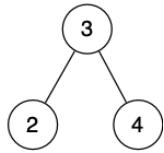
insert 3



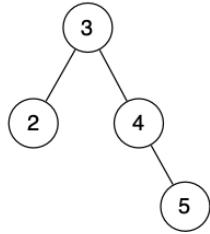
insert 2



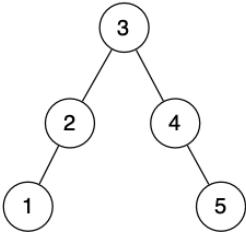
insert 4



insert 5

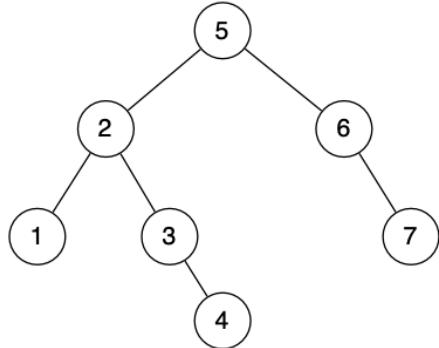


insert 1



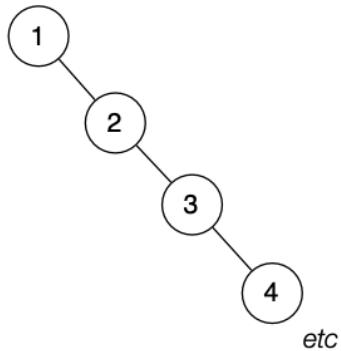
- Example 2:

Tree resulting from inserting: 5 6 2 3 4 7 1



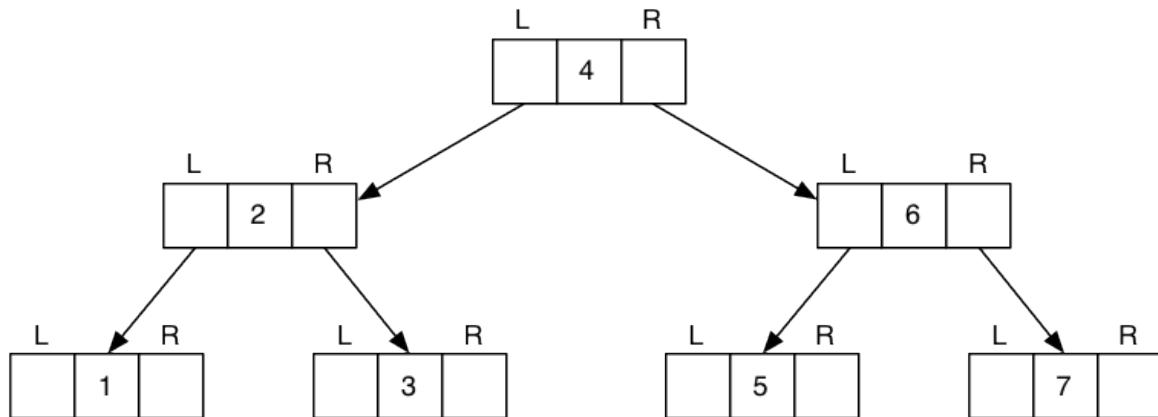
- Example 3:

Tree resulting from inserting: 1 2 3 4 5 6 7



Representing BSTs

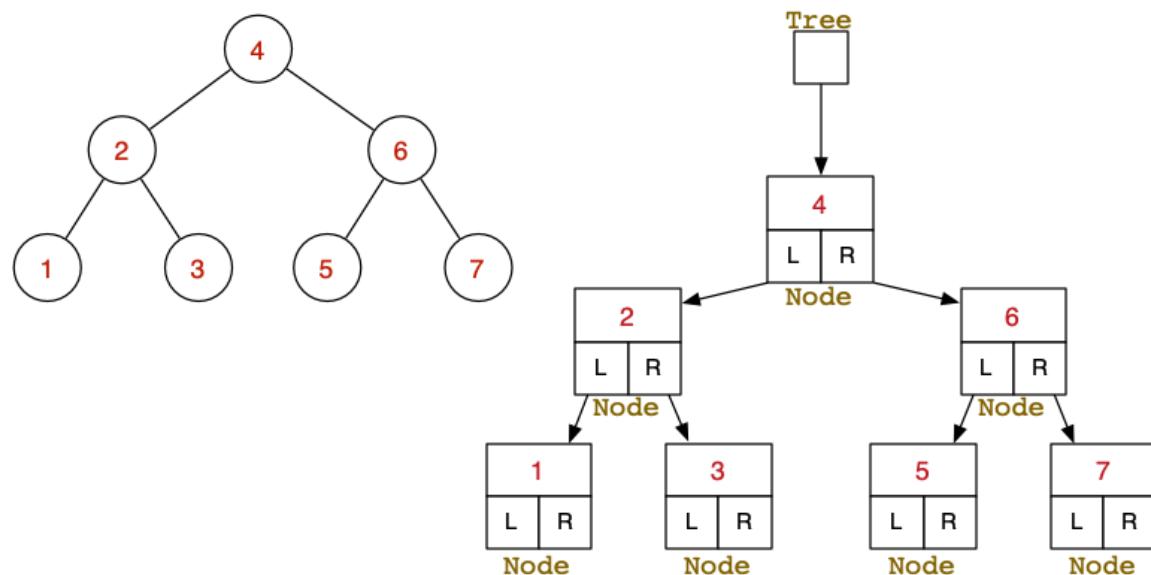
- Typically we use node structures



```
// a Tree is represented by a pointer to its root node
typedef struct Node *Tree;
```

```
// a Node contains its data, plus left and right subtrees
typedef struct Node {
    int data;
    Tree left, right;
} Node;
```

```
// some macros that we will use frequently
#define data(node) ((node)->data)
#define left(node) ((node)->left)
#define right(node) ((node)->right)
```



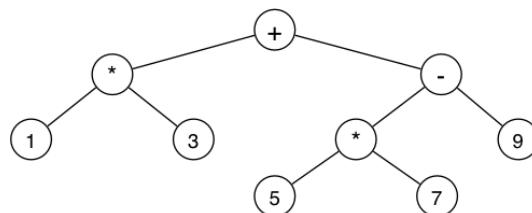
Tree Algorithms

```
TreeContains(tree, key):
| Input tree, key
| Output true if key found in tree, false otherwise
|
| if tree is empty then
|   return false
| else if key < data(tree) then
|   return TreeContains(left(tree), key)
| else if key > data(tree) then
|   return TreeContains(right(tree), key)
| else           // found
|   return true
| end if

TreeInsert(tree, item):
| Input tree, item
| Output tree with item inserted
|
| if tree is empty then
|   return new node containing item
| else if item < tree->data then
|   tree->left = TreeInsert(tree->left, item)
|   return tree
| else if item > tree->data then
|   tree->right = TreeInsert(tree->right, item)
|   return tree
| else
|   return tree      // avoid duplicates
| end if
```

Tree Traversal

- Iteration (traversal on)
 - Lists: Visit each value from first to last
 - Graphs: Visit each vertex, order determined by DFS/BFS
- For binary trees, several well-defined visiting orders exist (L – left, R – right, N – root):
 - Preorder (NLR)
 - Inorder (LNR)
 - Postorder (LRN)
 - Level-order: visit root, then all its children, then all their children
- For example:



- NLR: + * 1 3 - * 5 7 9 (prefix-order: useful for building tree)
LNR: 1 * 3 + 5 * 7 - 9 (infix-order: "natural" order)
LRN: 1 3 * 5 7 * 9 - + (postfix-order: useful for evaluation)
Level: + * - 1 3 * 9 5 7 (level-order: useful for printing tree)
- NLR (preorder traversal)

```
showBSTreePreorder(t):
```

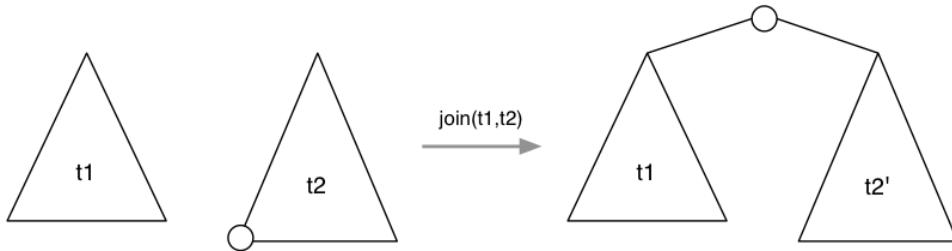
```

| Input tree t
|
| if t is not empty then
| | print data(t)
| | showBSTreePreorder(left(t))
| | showBSTreePreorder(right(t))
| end if
• LNR (inorder traversal)
showBSTreeInorder(t) :
| Input tree t
|
| if t is not empty then
| | showBSTreePreorder(left(t))
| | print data(t)
| | showBSTreePreorder(right(t))
| end if
• LRN (postorder traversal)
showBSTreePostorder(t) :
| Input tree t
|
| if t is not empty then
| | showBSTreePreorder(left(t))
| | showBSTreePreorder(right(t))
| | print data(t)
| end if
• Non-recursive NLR traversal
showBSTreePreorder(t) :
| Input tree t
|
| push t onto new stack S
| while stack is not empty do
| | t=pop(S)
| | print data(t)
| | if right(t) is not empty then
| | | push right(t) onto S
| | end if
| | if left(t) is not empty then
| | | push left(t) onto S
| | end if
| end while

```

Joining Two Trees

- $t = \text{TreeJoin}(t_1, t_2)$
- Pre-conditions:
 - Takes two BSTs; returns a single BST
 - $\text{Max}(\text{key}(t_1)) < \text{Min}(\text{key}(t_2))$
- Post-conditions:
 - Result is a BST (i.e. fully ordered)
 - Containing all items from t_1 and t_2
- Method is to make the min node of the right subtree as the new root of both trees

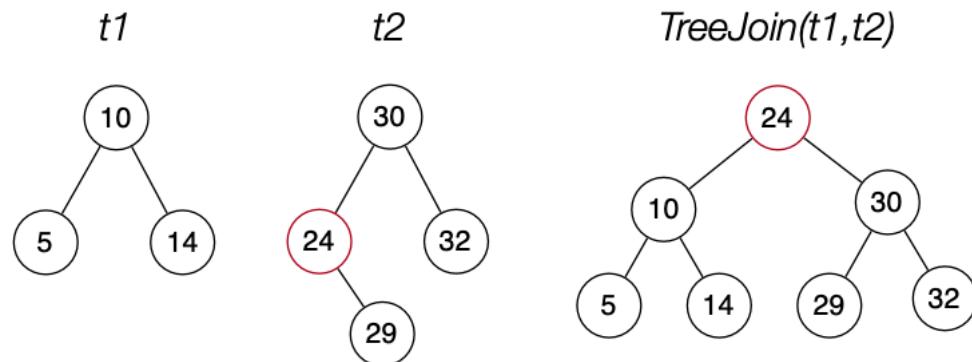


```

TreeJoin( $t_1, t_2$ ):
| Input trees  $t_1, t_2$ 
| Output  $t_1$  and  $t_2$  joined together
|
| if  $t_1$  is empty then return  $t_2$ 
| else if  $t_2$  is empty then return  $t_1$ 
| else
|   | curr= $t_2$ , parent=NULL
|   | while left(curr) is not empty do           // find min element in  $t_2$ 
|   |   | parent=curr
|   |   | curr=left(curr)
|   | end while
|   | if parent!=NULL then
|   |   | left(parent)=right(curr)    // unlink min element from parent
|   |   | right(curr)= $t_2$ 
|   | end if
|   | left(curr)= $t_1$ 
|   | return curr                      // curr is new root
| end if

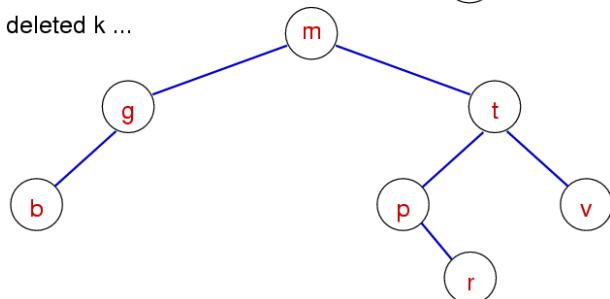
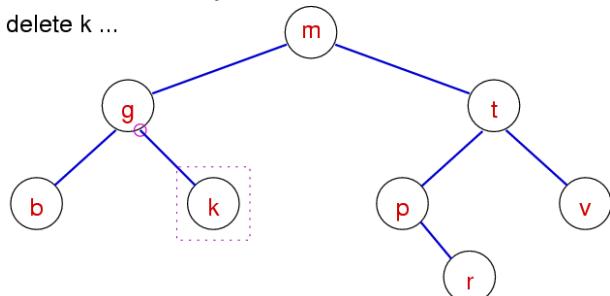
```

- Example:

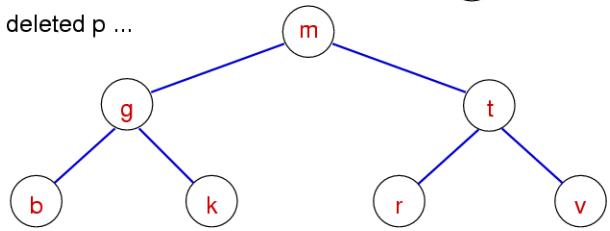
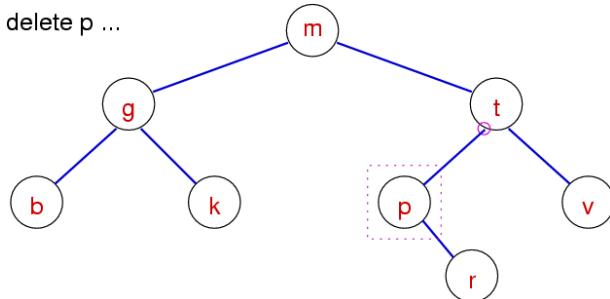


Deletion from BSTs

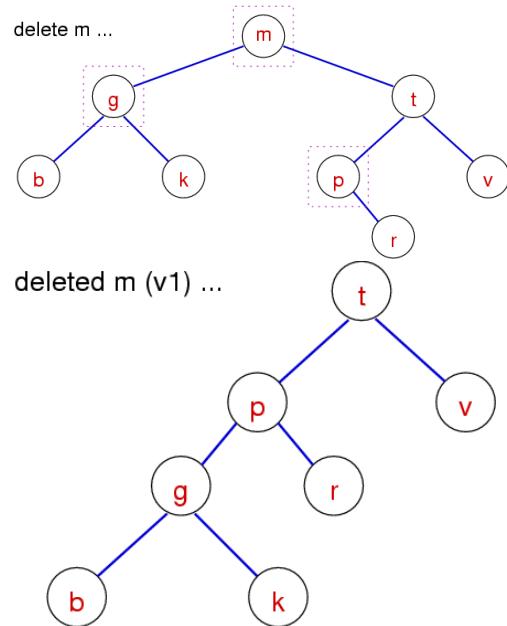
- Insertion into a binary search tree is easy
- Deletion from a binary search tree is harder
- Four cases to consider:
 - Empty tree, so new tree is also empty
 - Zero subtrees, so just delete the node
delete k ...



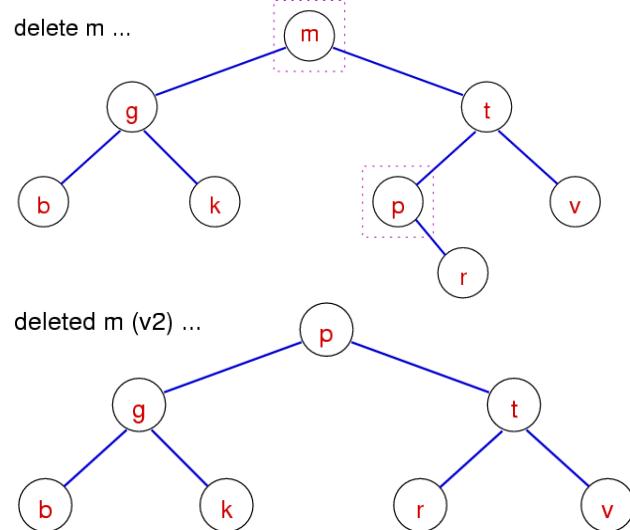
- One subtree, so replace by child
delete p ...



- Two subtrees, so replace by successor, join two subtrees
 - Version 1: right child comes new root, attach left subtree t min element of right subtree



- Version 2: Join left and right subtree



```

TreeDelete(t,item):
| Input tree t, item
| Output t with item deleted
|
| if t is not empty then           // nothing to do if tree is empty
| | if item < data(t) then        // delete item in left subtree
| | | left(t)=TreeDelete(left(t),item)
| | else if item > data(t) then  // delete item in left subtree
| | | right(t)=TreeDelete(right(t),item)
| | else                         // node 't' must be deleted
| | | if left(t) and right(t) are empty then
| | | | new=empty tree           // 0 children
| | | | else if left(t) is empty then
| | | | | new=right(t)          // 1 child
| | | | else if right(t) is empty then
| | | | | new=left(t)           // 1 child
| | | | else
| | | | | new=TreeJoin(left(t),right(t)) // 2 children
| | | | end if
| | | | free memory allocated for t
| | | | t=new
| | | end if
| end if
return t
  
```

3.2: Function Pointers in C

- C can pass functions by passing a pointer to them
- Function pointers are
 - references to memory address of functions
 - pointer values and can be assigned/passed
- Function pointer variables/parameters are declared as:
typeOfReturnValue (*fp)(typeOfArguments)
- Example: fp points to a function that returns int and have one argument of type int
`int (*fp) (int)`
- Example 1:

```
void traverse (list ls, void (*fp) (list)) {
    list curr = ls;
    while(curr != NULL) {
        // call function for the node
        fp(curr);
        curr = curr->next;
    }
}
```

Second argument is `fp`,
Pointer to a function like,
`void functionName(list n)`

```
void printNode(list n){
    if(n != NULL) {
        printf("%d->", n->data);
    }
}
```

```
void printGrade(list n){
    if(n != NULL) {
        if(n->data >= 50) {
            printf("Pass");
        } else {
            printf("Fail");
        }
    }
}
```

```
void traverse (list ls, void (*fp) (list));
//The second argument must have matching prototype
traverse(myList,    printNode);
traverse(myList,    printGrade);
```

- Example 2:

```
/* compare function
*/
int myCmp1(int val1, int val2){
    if(val1 < val2) { return -1;}
    else if (val1 > val2) { return 1; }
    else { return 0; }
}

/* Another compare function
*/
int myCmp2(int val1, int val2){
    if(val1 < val2) { return 1;}
    else if (val1 > val2) { return -1; }
    else { return 0; }
}
```

```
/*
* Pre: n > 0, array a with at least one element
* Post: ???
*/
int processA(int *a, int n, int (*fp) (int, int)){
    int cur = a[0];

    for(int i=1; i<n; i++){
        if( fp (cur, a[i]) < 0){
            cur = a[i];
        }
    }
    return cur;
}
```

```
int main(int argc, char *argv[]) {

    int marksA[] = {67, 34, 81, 44, 19};

    int ans1 = processA (marksA, 5, myCmp1);
    printf("ans1 is %d \n", ans1);
    printf("----- \n");

    int ans2 = processA (marksA, 5, myCmp2);
    printf("ans2 is %d \n", ans2);
    printf("----- \n");

    return 0;
}
```

Week 4

4.1: Balancing Search Trees

- The goal is to build a binary search tree which has:
 - minimum height \rightarrow minimum worst case search cost

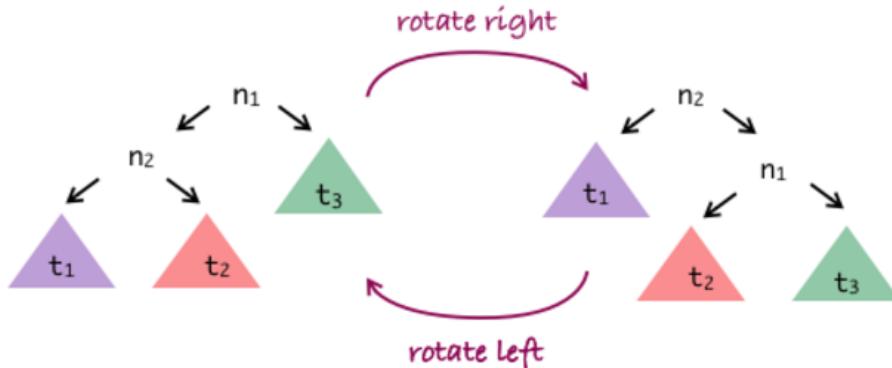
Perfectly Balanced Tree

- A perfectly balanced tree with N nodes has:
 - For all nodes,
 $|number_of_nodes(left\ tree) - number_of_nodes(right\ tree)| < 2$
 - Height of $\log_2 N$, so that the worst case search is $O(\log_2 N)$

Operations for Rebalancing

- Left rotation:
 - Move right child to root; rearrange links to retain order
- Right rotation
 - Move left child to root; rearrange links to retain order
- Insertion at root:
 - Each new item is added as the new root node
- Partition:
 - Rearrange tree around specified node (push it to root)

Tree Rotation



- Rotating right:
 - n_1 is the current root; n_2 is the root of n_1 's left subtree
 - n_1 gets new left subtree, which is n_2 's right subtree
 - n_1 becomes root of n_2 's new right subtree
 - n_2 becomes new root
 - n_2 's left subtree is unchanged
- Left rotation:
 - n_2 is the current root; n_1 is the root of n_2 's right subtree
 - n_2 gets new right subtree, which is n_1 's left subtree
 - n_2 becomes the root of n_1 's new left subtree
 - n_1 becomes new root
 - n_1 's right subtree is unchanged
- Rotation requires simple, localised pointer rearrangements
- Cost of tree rotation: $O(1)$

```

rotateRight(n1):
| Input tree n1
| Output n1 rotated to the right
|
| if n1 is empty V left(n1) is empty then
|   return n1
| end if
| n2= n1->left
| n1->left = n2->right
| n2->right =n1
| return n2

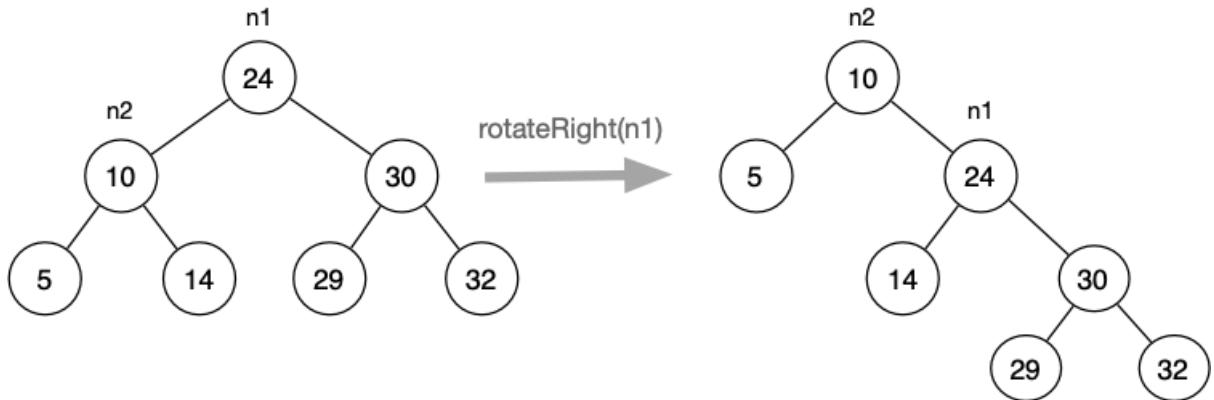
```

```

rotateLeft(n2):
| Input tree n2
| Output n2 rotated to the left
|
| if n2 is empty V right(n2) is empty then
|   return n2
| end if
| n1=n2->right
| n2->right = n1->left
| n1->left = n2
| return n1

```

- Example:



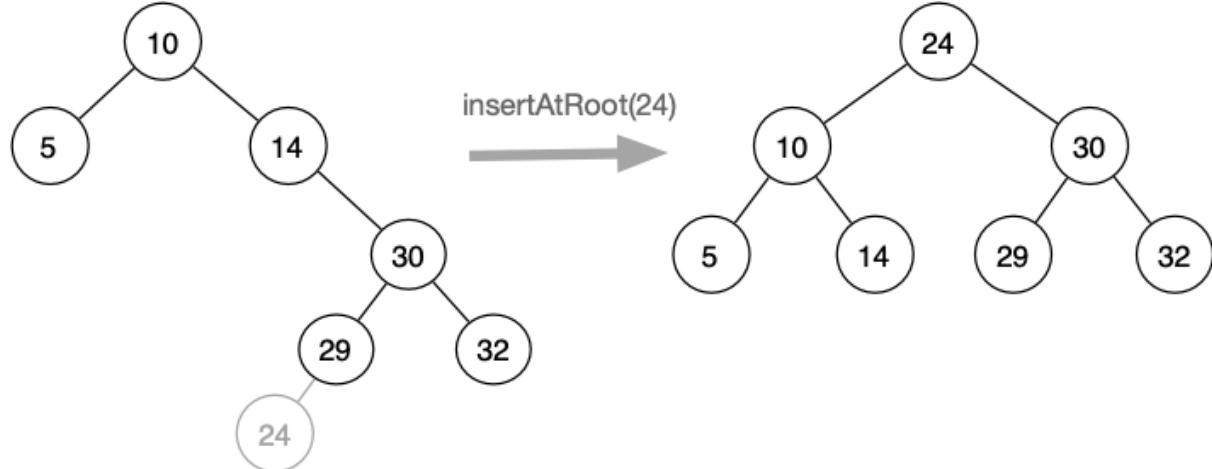
- Tree rotation is cheap: O(1)
- If applied appropriately, will tend to improve tree balance
- Sometimes rotation is applied from leaf to root, along one branch
 - Cost of this is O(height)
 - Payoff is improved balance which reduces height
- Reduced height pushes search cost to O(logN)

Insertion at Root

- Potential disadvantages:
 - Large-scale rearrangement of tree for each insert
- Potential advantages:
 - Recently inserted items are close to root
 - Lower cost if recent items more likely to be searched
- Method:
 - Base case:
 - Tree is empty; make new node and make it root
 - Recursive case:
 - Insert new node as root of appropriate subtree
 - Lift new node to root by rotation

```
insertAtRoot(t, it):
| Input tree t, item it to be inserted
| Output modified tree with item at root
|
| if t is empty tree then
|   t = new node containing item
| else if item < t->data then
|   t->left = insertAtRoot( t->left, it)
|   t = rotateRight(t)
| else if item > t->data then
|   t->right = insertAtRoot( t->right, it)
|   t = rotateLeft(t)
| end if
| return t;
```

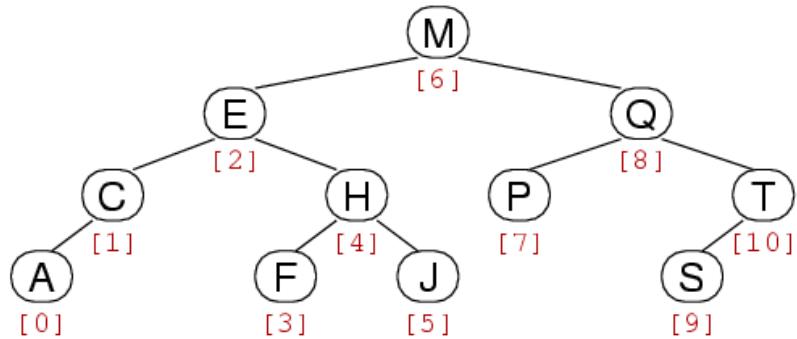
- Example:



- Same complexity as for insertion-at-leaf: $O(\text{height})$
 - But cost is effectively doubled ... traverse down, rotate up
 - Tendency to be balanced, but not guaranteed
- Benefit comes in searching
 - For some applications, search favours recently-added items
 - Insertion-at-root ensures these are close to root
- Could even consider 'move to root when found'

Tree Partitioning

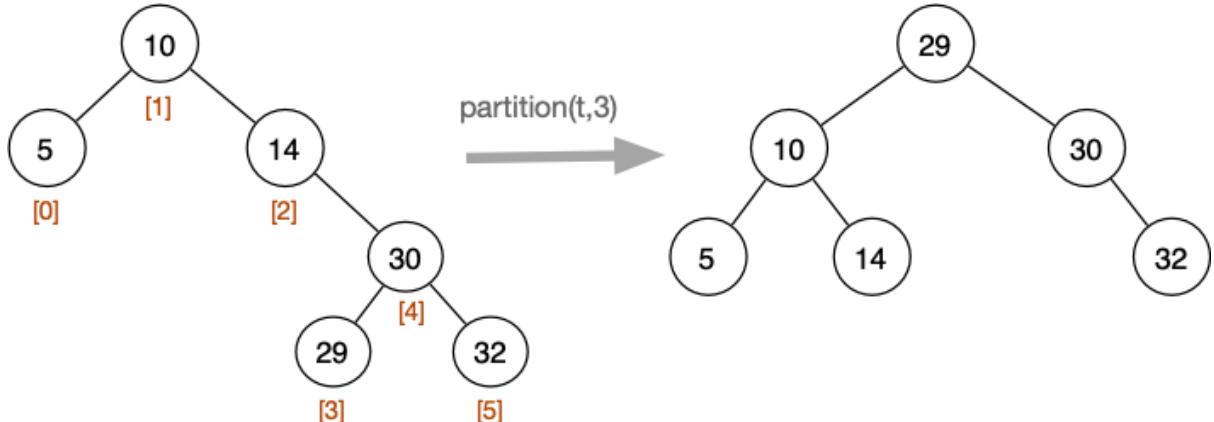
- Rearranges so that element with index i becomes root
- For three with N nodes, indices are $0 \dots N-1$, in LNR order



```

partition(tree, i):
|   Input tree with n nodes, index i
|   Output tree with  $i^{\text{th}}$  item moved to the root
|
|   m = number_of_nodes(left(tree))
|   if  $i < m$  then
|       tree->left = partition(tree->left, i)
|       tree = rotateRight(tree)
|   else if  $i > m$  then
|       tree->right = partition(tree->right, i-m-1)
|       tree = rotateLeft(tree)
|   end if
|   return tree
  
```

- Example:



- No requirement for search (using element index instead)
- After each recursive portioning step, one rotation
- Overall cost similar to insert-at-root
- Benefits:
 - Improves balance, so improves search cost

Periodic Rebalancing

- Insert at leaves as before; periodically, rebalance the tree every k insertions

```

Input tree, item
Output tree with item randomly inserted

|
| t=insertAtLeaf(tree,item)
| if #nodes(t) mod k = 0 then
|   t=rebalance(t)
| end if
| return t

```

- Problems:

- Has to traverse entire subtree to count number of nodes

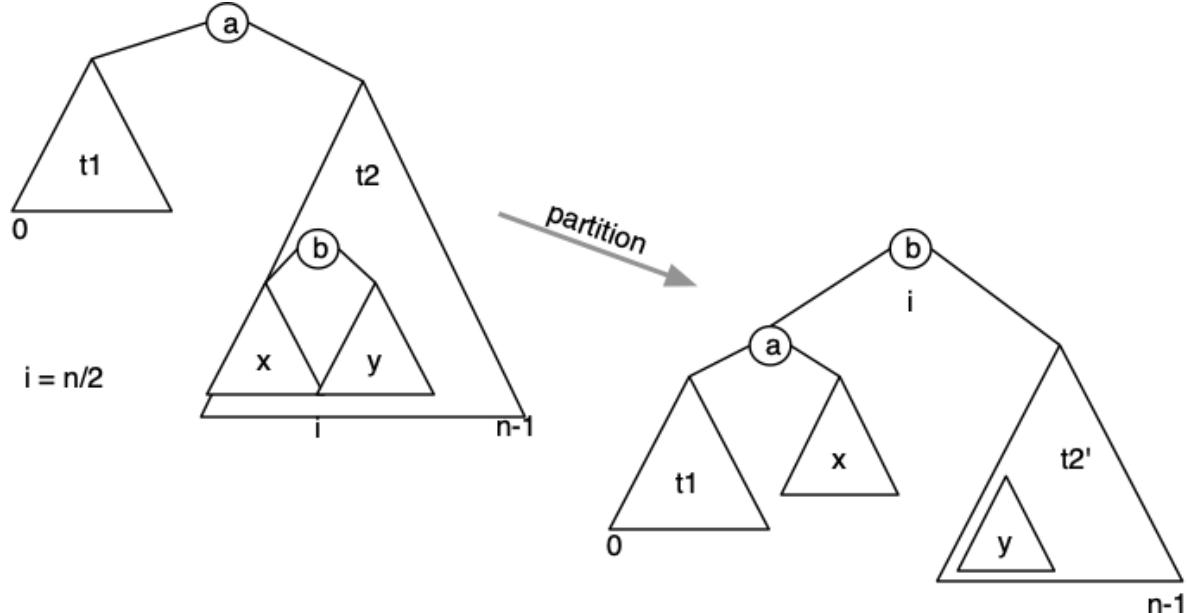
- Better to change nodes structure

```

typedef struct Node {
    int data;
    int nnodes;          // #nodes in my tree
    Tree left, right; // subtrees
} Node;

```

- How to rebalance a BST? Move median item to root



```

rebalance(t):
| Input tree t with n nodes
| Output t rebalanced
|
| if n≥3 then
|   // put node with median key at root
|   t=partition(t, n/2)
|   // then rebalance each subtree
|   t->left = rebalance(t->left)
|   t->right = rebalance(t->right)
| end if
| return t

```

Randomised BST Insertion

```
insertRandom(tree, item)
| Input tree, item
| Output tree with item randomly inserted
|
| if tree is empty then
|   return new node containing item
| end if
| // p/q chance of doing root insert
| if random() mod q < p then
|   return insertAtRoot(tree, item)
| else
|   return insertAtLeaf(tree, item)
| end if
```

- E.g. 30% chance, choose $p = 3, q = 10$
- Similar to cost for inserting keys in random order: $O(\log N)$
- Does not rely on keys being supplied in random order
- Approach can also be applied to deletion:
 - Promote inorder successor from right subtree OR left subtree

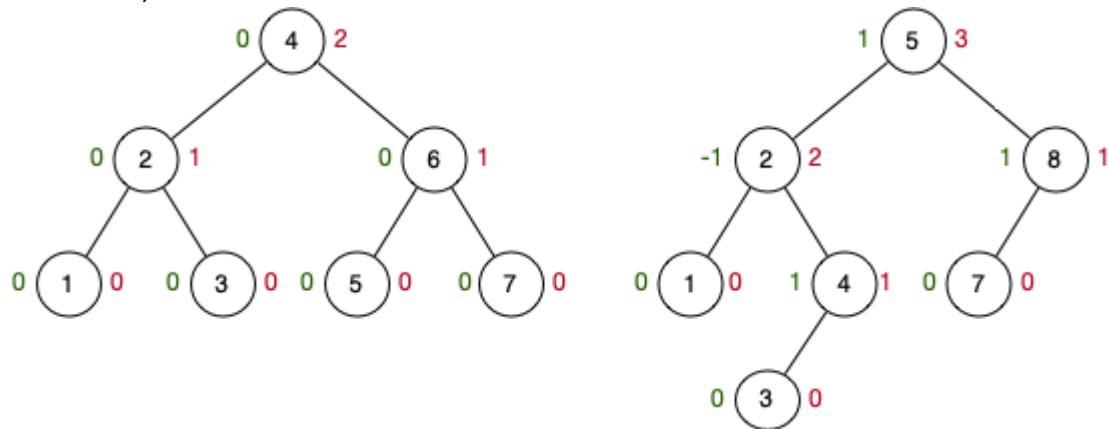
4.2: AVL Trees

Better Balanced Binary Search Trees

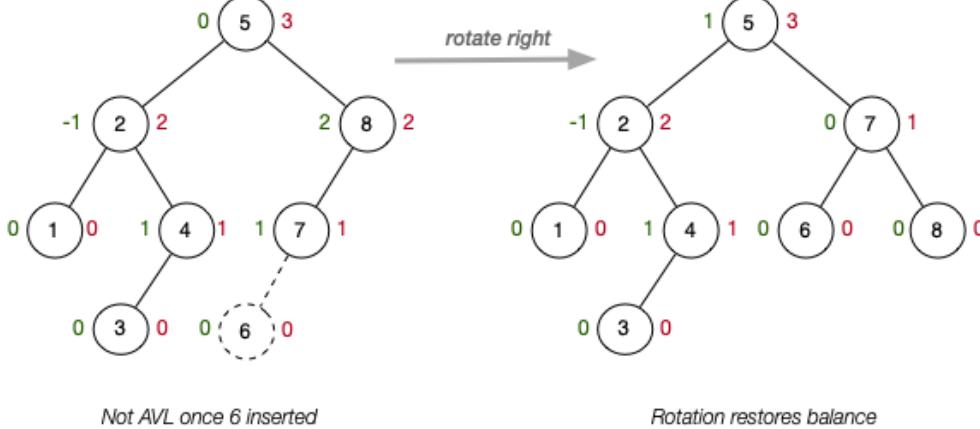
- So far, we have seen ...
 - Randomised trees ... make poor performance unlikely
 - Occasional rebalance ... fix balance periodically
 - Splay trees ... reasonable amortized performance
 - But all types still have $O(n)$ worst case
- Ideally we want both average/worst case to be $O(\log N)$
 - AVL trees ... fix imbalances as soon as they occur
 - 2-3-4 trees ... use varying sized nodes to assist balance
 - Red-black trees ... isomorphic to 2-3-4, but binary nodes

AVL Trees

- Goal:
 - Tree remains reasonably well-balanced $O(\log N)$
 - Cost of fixing imbalance is relatively cheap
- Approach:
 - Insertion (at leaves) may cause imbalance
 - Repair balance as soon as we notice imbalance
 - Repairs done locally, not by overall tree restructure
- A tree is unbalanced when $|height(left subtree) - height (right subtree)| > 1$
- This can be repaired by rotation:
 - If left subtree too deep, rotate right
 - If right subtree too deep, rotate left
- Problem: Determining height/depth of subtrees is expensive
 - Need to traverse whole subtree to find longest path
- Solution: Store balance data in each node
- Real numbers are height, green numbers are balance (difference in height between left and right subtree)



- Example of AVL tree rebalancing:



AVL Insertion

```

insertAVL(tree, item):
| Input tree, item
| Output tree with item AVL-inserted

| if tree is empty then
|   return new node containing item
| else if item = data(tree) then
|   return tree
| else
|   | if item < data(tree) then
|   |   left(tree) = insertAVL(left(tree), item)
|   | else if item > data(tree) then
|   |   right(tree) = insertAVL(right(tree), item)
|   | end if
|   | LHeight = height(left(tree))
|   | RHeight = height(right(tree))
|   | if (LHeight - RHeight) > 1 then
|   |   | if item > data(left(tree)) then
|   |   |       left(tree) = rotateLeft(left(tree))
|   |   |   end if
|   |   |   tree=rotateRight(tree)
|   |   | else if (RHeight - LHeight) > 1 then
|   |   |       if item < data(right(tree)) then
|   |   |           right(tree) = rotateRight(right(tree))
|   |   |       end if
|   |   |       tree=rotateLeft(tree)
|   |   | end if
|   |   return tree
| end if

```

Performance of AVL Trees

- Analysis of AVL trees:
 - Trees are height-balanced; subtree depths differ by +/- 1
 - Average/worst-case search performance of $O(\log N)$
 - Require extra data to be stored in each node (efficiency)
 - Require extra data to be maintained during insertion
 - May not be weight-balanced; subtree sizes may differ

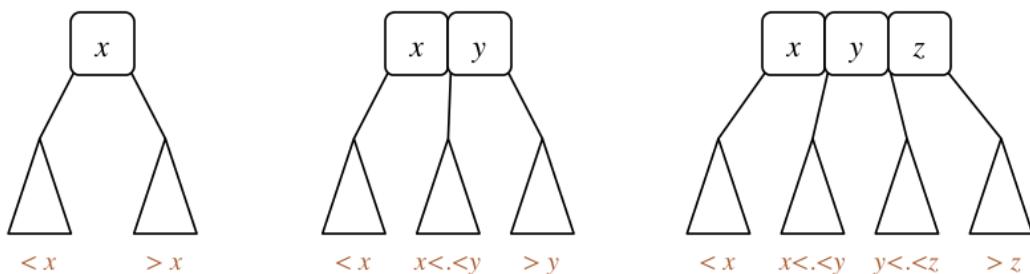
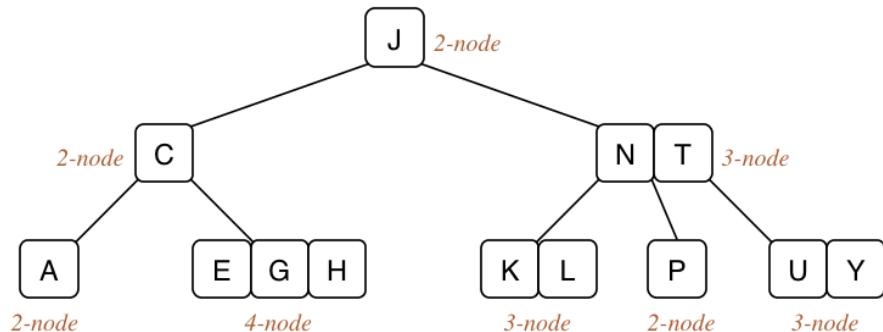
4.3: 2-3-4 Trees

Search Cost

- Critical factor determining search cost in BSTs
 - Worst case: Length of longest path
 - Average case: < average path length
- Either way, path length (tree depth) is a critical factor
- What if branching factor is > 2?
- $\log_2 4096 = 12$, $\log_4 4096 = 6$, $\log_8 4096 = 4$

2-3-4 Trees

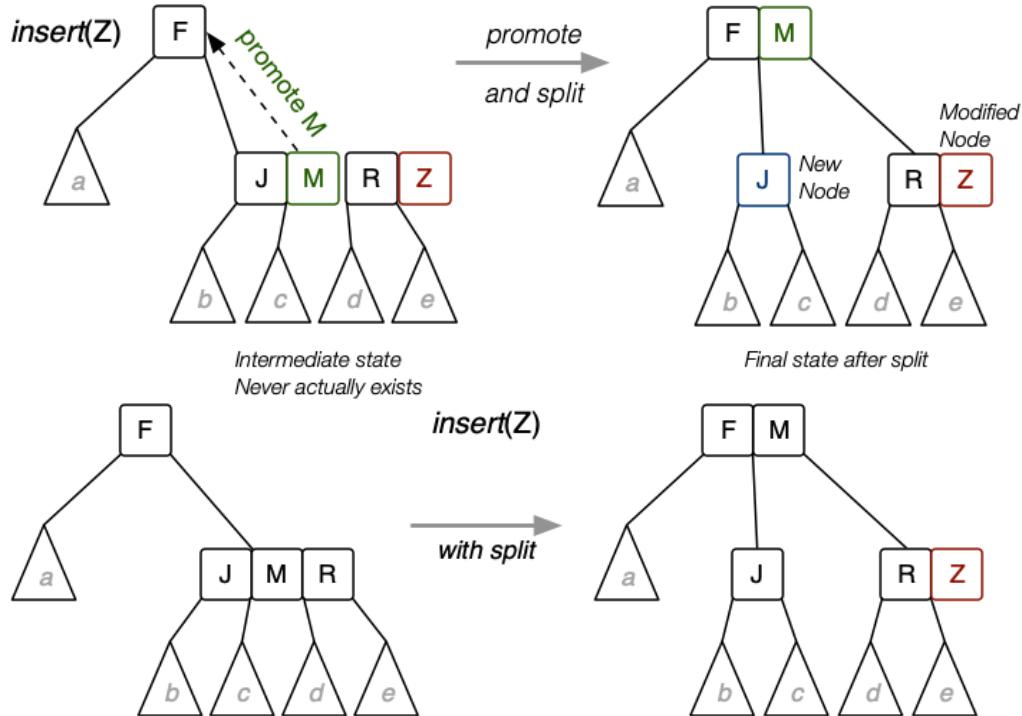
- 2-nodes, with two children (same as BSTs)
- 3-nodes, two values, three children
- 4-nodes, three values and four children



- In a balanced 2-3-4 tree:
 - All leaves are at same distance from the root
- 2-3-4 trees grow ‘upwards’ from the leaves, via node splitting

Node Splitting

- Insertion into a full node causes a split
- The middle value is propagated to the parent node
- Values in the original node split across original node and new node



Searching in 2-3-4 Trees:

```

Search(tree,item):
| Input tree, item
| Output address of item if found in 2-3-4 tree
|   NULL otherwise

| if tree is empty then
|   return NULL
| else
|   | scan tree.data to find i such that
|   |   tree.data[i-1] < item ≤ tree.data[i]
|   | if item=tree.data[i] then // item found
|   |   return address of tree.data[i]
|   | else           // keep looking in relevant subtree
|   |   return Search(tree.child[i],item)
|   | end if
| end if

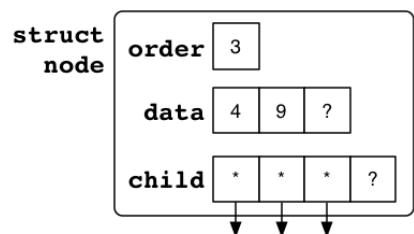
```

Possible Data Structure

```

typedef struct node {
    int          order;      // 2, 3 or 4
    int          data[3];    // items in node
    struct node *child[4];  // links to subtrees
} node;

```

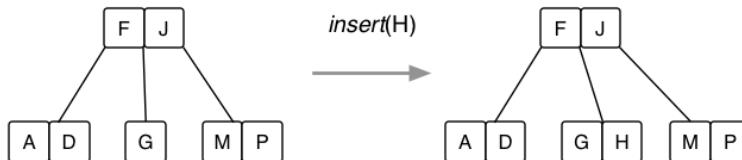
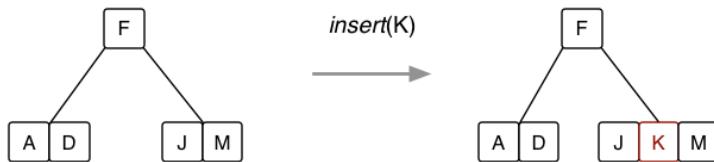


Search Cost Analysis

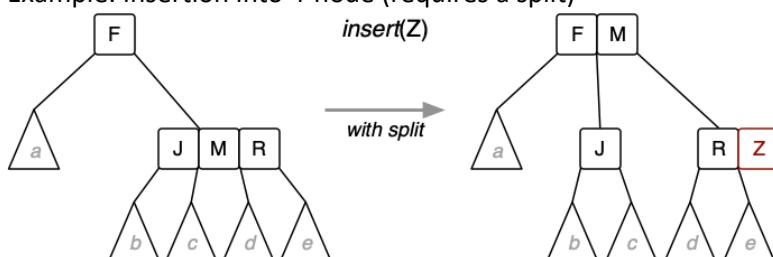
- In other trees, worst case is determined by height
- 2-3-4 trees are always balanced, so height is $O(\log N)$
- Worst case for height: All nodes are 2-nodes: $O(\log N)$
- Best case for height: All nodes are 4-nodes: $O(\log_4 N)$

Insertion into 2-3-4 Trees

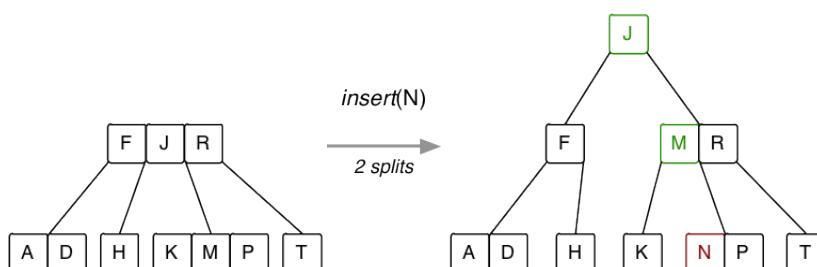
- Find a leaf node where the item belongs
- If not full (i.e. order < 4)
 - Insert item in this node, order++
- If node is full (i.e. contains 3 items)
 - Split into two 2-nodes as leaves
 - Promote middle element to parent
 - Insert item into appropriate leaf 2-node
 - If parent is a 4-node
 - Continue split/promote upwards
 - If promote to root, and root is a 4-node
 - Split root node and add new root
- Example: Insertion into a 2-node or 3-node:



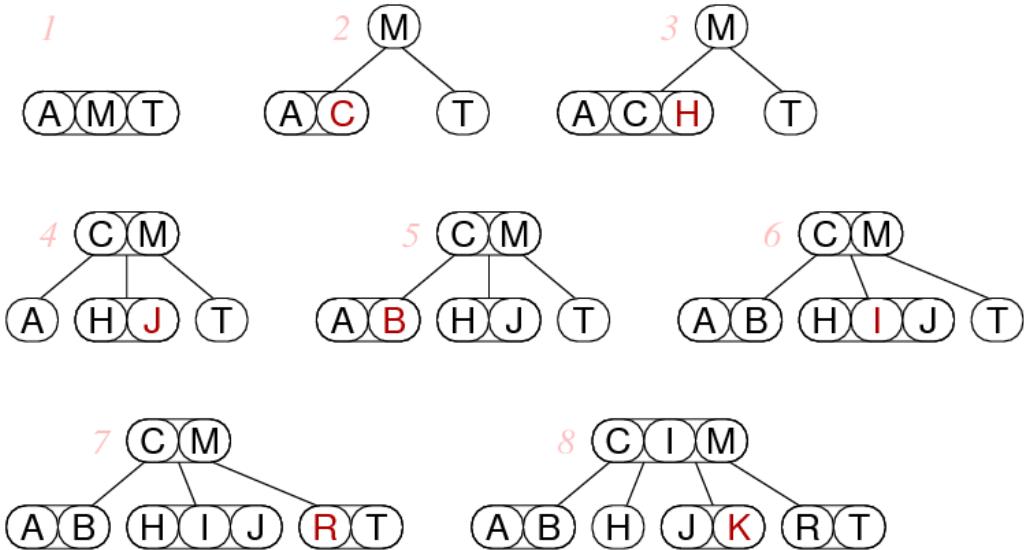
- Example: Insertion into 4-node (requires a split)



- Example: Splitting the root



- Example:



Insertion Algorithm

```

insert(tree, item):
| Input 2-3-4 tree, item
| Output tree with item inserted

| if tree is empty then
| | return new node containing item
| end if
| node=Search(tree, item)
| parent=parent of node
| if node.order < 4 then
| | insert item into node
| | increment node.order
| else
| | | promote = node.data[1]      // middle value
| | | nodeL   = new node containing data[0]
| | | nodeR   = new node containing data[2]
| | | delete node
| | | if item < promote then
| | | | insert(nodeL, item)
| | | else
| | | | insert(nodeR, item)
| | | end if
| | | insert(parent, promote)
| | while parent.order=4 do
| | | continue promote/split upwards
| | end while
| | if parent is root  $\wedge$  parent.order=4 then
| | | split root, making new root
| | end if
| end if

```

Performance Analysis

- Searching for leaf node in which to insert = $O(\log N)$
- If node not full, insert item into node = $O(1)$
- If node full, promote middle, create two new nodes = $O(1)$
- If promotion propagates...
 - Best case: Update parent = $O(1)$
 - Worst case: Propagate to root = $O(\log N)$
- Overall insertion cost = $O(\log N)$

2-3-4 Variations

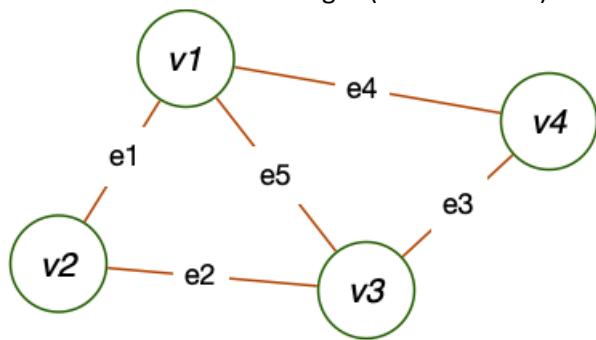
- Why stop at 4? Why no 2-3-4-5 trees? Or M-way trees?
- Or don't have variable sized nodes, simulate 2-3-4 trees with binary trees -> red black trees

Week 5

5.1: Graph Basics

- A graph $G = (V, E)$:

- V is a set of vertices
- E is a set of edges (subset of $V \times V$)



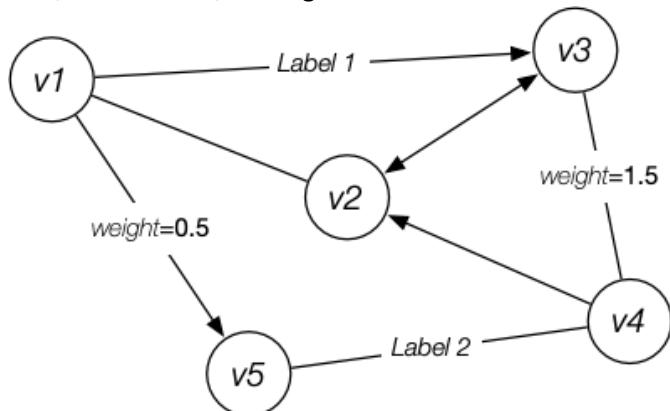
$$V = \{ v1, v2, v3, v4 \}$$

$$E = \{ e1, e2, e3, e4, e5 \}$$

or

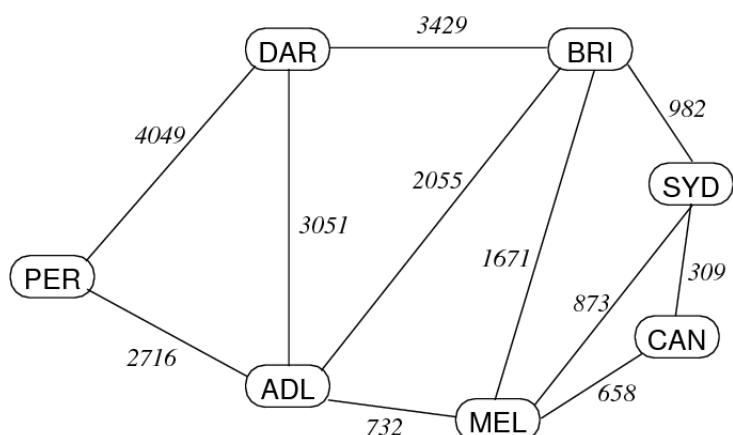
$$E = \{ (v1, v2), (v2, v3), (v3, v4), (v1, v4), (v1, v3) \}$$

- Nodes are distinguished by a unique identifier
- Edges may be directed, labelled and/or weighted



- Example:

Distance	Adelaide	Brisbane	Canberra	Darwin	Melbourne	Perth	Sydney
Adelaide	-	2055	1390	3051	732	2716	1605
Brisbane	2055	-	1291	3429	1671	4771	982
Canberra	1390	1291	-	4441	658	4106	309
Darwin	3051	3429	4441	-	3783	4049	4411
Melbourne	732	1671	658	3783	-	3448	873
Perth	2716	4771	4106	4049	3448	-	3972
Sydney	1605	982	309	4411	873	3972	-



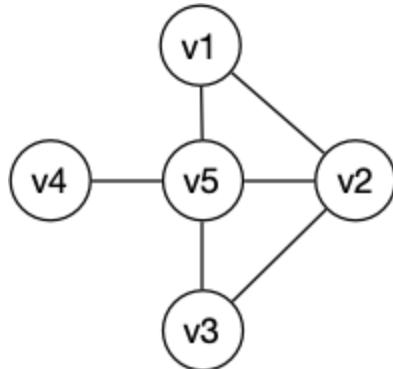
- Graph algorithms are generally more complex than tree/list ones:
 - No implicit order of items
 - Graphs may contain cycles
 - Concentric representation is less obvious
 - Algorithm complexity depends on connection complexity

Properties of Graphs

- A graph with V vertices has maximum $V(V-1)/2$ edges
- The ratio $E:V$ can vary
 - If E is closer to V^2 , the graph is dense
 - If E is closer to V , the graph is sparse
- Knowing if the graph is sparse or dense is important
 - May affect choice of data structures to represent the graph
 - May affect choice of algorithms to process graph

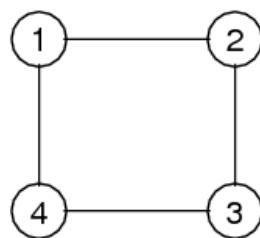
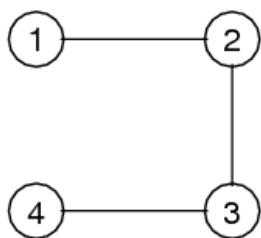
Graph Terminology

- For an edge e that connects vertices v and w
 - v and w are adjacent (neighbours)
 - e is incident on both v and w
- Degree of a vertex v
 - Number of edges incident on v (how many edges the vertex v has)



$\text{degree}(v1) = 2$
 $\text{degree}(v2) = 3$
 $\text{degree}(v3) = 2$
 $\text{degree}(v4) = 1$
 $\text{degree}(v5) = 4$

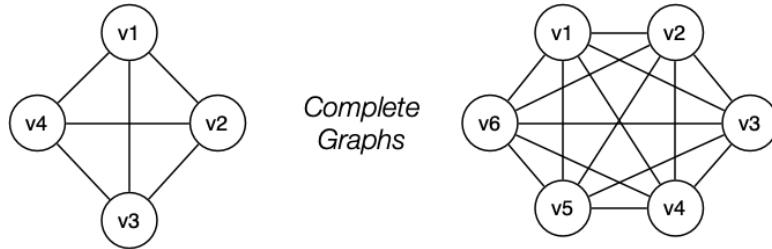
- Synonyms:
 - Vertex = Node
 - Edge = Arc = Link (note: some people use arc for directed edges)
- Path: A sequence of vertices where each vertex has an edge to its predecessor
- Cycle: A path where last vertex in path is same as first vertex in path
- Length of path or cycle = #edges



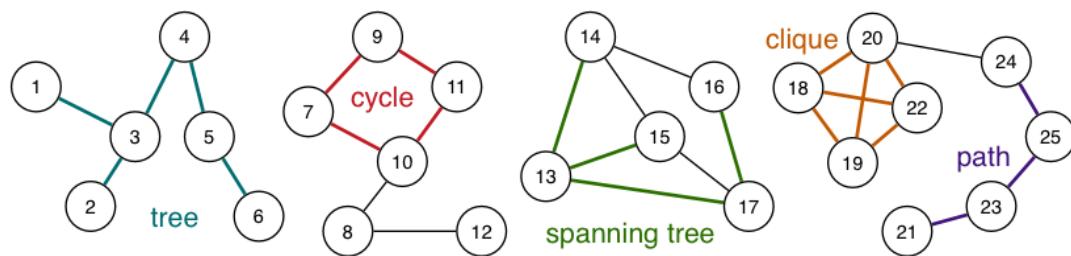
Path: 1–2, 2–3, 3–4

Cycle: 1–2, 2–3, 3–4, 4–1

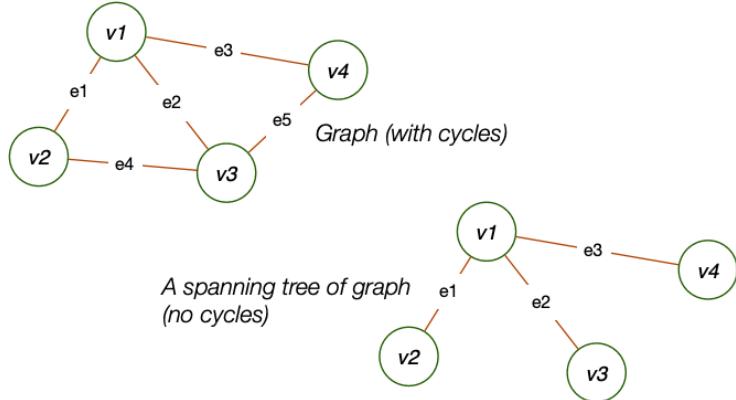
- Connected graph: There is a path from each vertex to every other vertex
- Complete graph: There is an edge from each vertex to every other vertex



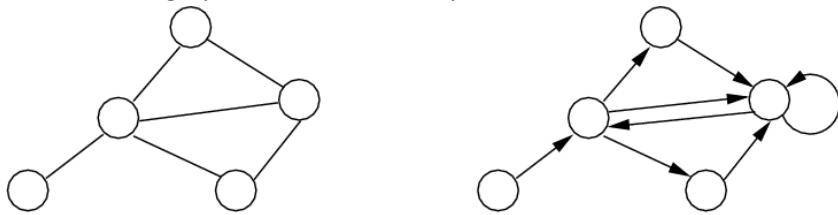
- Tree: Connected (sub)graph with no cycles
- Spanning tree: Tree containing all vertices
- Clique: Complete subgraph
- Consider the following as a single graph
 - This graph has 25 edges, 32 edges and 4 connected components



- A spanning tree: A subgraph of G containing all of V and is a single tree (connected, no cycles)
- A spanning forest: A subgraph of G containing all of V and is a set of trees (not connected, no cycles)



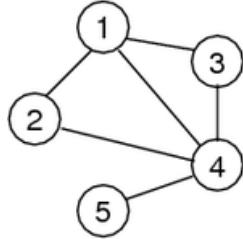
- Undirected graph: No self loops
- Directed graph: Can have self loops



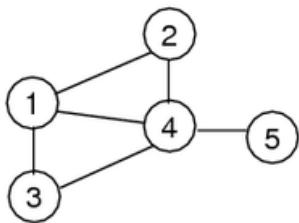
- Weighted graph: Each edge has an associated value (weight)
- Multi-graph: Allow multiple edges between two vertices
- Labelled graph: Edges have associated labels

5.1: Graph Representations

- Four representations of the same graph:



(a)



(b)

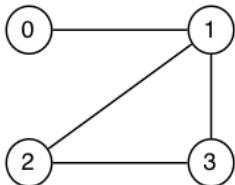
1–2	1–3	1–4	1–3
2–4			2–4
3–4			4–1 4–3
4–5			5–4

(c)

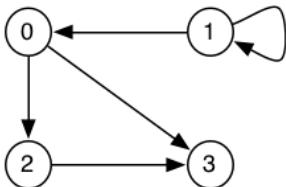
(d)

Array of Edges

- Edges are representations of Edge values (= pairs of vertices)
 - Space efficient
 - Adding and deleting edges is slightly complex
 - Undirected: Order of vertices in an Edge doesn't matter
 - Directed: Order of vertices in an Edge encodes direction



[(0,1), (1,2), (1,3), (2,3)]



[(1,0), (1,1), (0,2), (0,3), (2,3)]

- Assume edges is numbered 0 – (V – 1)

- Graph intialisation:

```
newGraph(V) :
| Input number of nodes V
| Output new empty graph (no edges)
|
| g.nV = V      // #vertices (numbered 0..V-1)
| g.nE = 0      // #edges
| allocate enough memory for g.edges[]
| return g
```

- Assumes struct Graph { int nV; int nE; Edge edges[]; }

- Edge insertion:

```
insertEdge(g, (v,w)) :
| Input graph g, edge (v,w)
| Output graph g containing (v,w)
|
| i=0
| while i < g.nE  $\wedge$  g.edges[i]  $\neq$  (v,w) do
|   i=i+1
| end while
| if i=g.nE then      // (v,w) not found
|   g.edges[i]=(v,w)
|   g.nE=g.nE+1
| end if
```

- Edge removal

```
removeEdge (g, (v,w)) :
| Input graph g, edge (v,w)
| Output graph g without (v,w)
|
| i=0
| while i < g.nE  $\wedge$  g.edges[i]  $\neq$  (v,w) do
|   i=i+1
| end while
| if i < g.nE then // (v,w) found
|   g.edges[i]=g.edges[g.nE-1]
|   // replace by last edge in array
|   g.nE=g.nE-1
| end if
```

- Print a list of edges

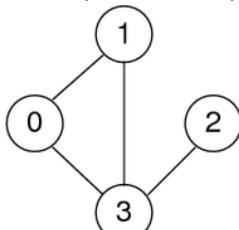
```
showEdges (g) :
| Input graph g
|
| for all i=0 to g.nE-1 do
|   | (v,w)=g.edges[i]
|   | print v"--"w
| end for
```

- Array-of-edges Cost Analysis

- Storage cost: $O(E)$
- Cost of operations:
 - Initialisation: $O(1)$
 - Insert edge: $O(E)$
 - Delete edge: $O(E)$
- If array is full on insert
 - Allocate space for a bigger array, copy edges across -> $O(E)$
- If we maintain edges in order
 - Use binary search to find edge -> $O(\log E)$

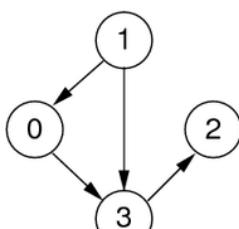
Adjacency Matrix Representation

- Represented by $V \times V$ matrix



Undirected graph

A	0	1	2	3
0	0	1	0	1
1	1	0	0	1
2	0	0	0	1
3	1	1	1	0



Directed graph

A	0	1	2	3
0	0	0	0	1
1	1	0	0	1
2	0	0	0	0
3	0	0	1	0

- Advantages:
 - Easily implemented as 2-dimensional array
 - Can represent graphs, digraphs and weighted graphs
 - Graphs: Symmetric Boolean matrix
 - Digraphs: Non-symmetric Boolean matrix
 - Weighted: Non-symmetric matrix of weight values

- Disadvantages:
 - If few edges (sparse) -> memory inefficient ($O(V^2)$ space)

- Graph initialisation

```
newGraph(V) :
| Input number of nodes V
| Output new empty graph
|
| g.nV = V    // #vertices (numbered 0..V-1)
| g.nE = 0    // #edges
| allocate memory for g.edges[] []
| for all i,j=0..V-1 do
|   g.edges[i][j]=0  // false
| end for
| return g
```

- Edge insertion

```
insertEdge(g, (v,w)) :
| Input graph g, edge (v,w)
| Output graph g containing (v,w)
|
| if g.edges[v][w] = 0 then // (v,w) not in graph
|   g.edges[v][w]=1        // set to true
|   g.edges[w][v]=1
|   g.nE=g.nE+1
| end if
```

- Edge removal

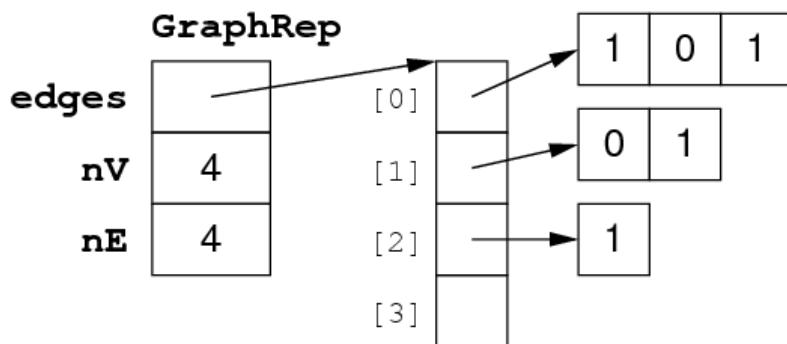
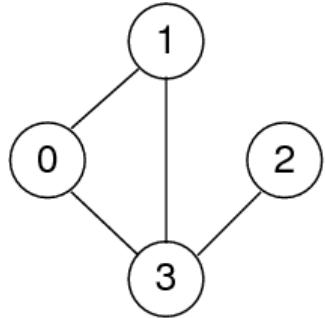
```
removeEdge(g, (v,w)) :
| Input graph g, edge (v,w)
| Output graph g without (v,w)
|
| if g.edges[v][w] ≠ 0 then // (v,w) in graph
|   g.edges[v][w]=0        // set to false
|   g.edges[w][v]=0
|   g.nE=g.nE-1
| end if
```

- Print a list of edges

```
showEdges(g) :
| Input graph g
|
| for all i=0 to g.nV-1 do
|   for all j=i+1 to g.nV-1 do
|     if g.edges[i][j] ≠ 0 then
|       print i"-j
|     end if
|   end for
| end for
```

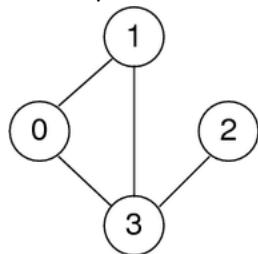
- Cost Analysis
 - Storage cost: $O(V^2)$
 - If the graph is sparse, most storage is wasted
 - Cost of operations:
 - Initialisation: $O(V^2)$
 - Insert edge: $O(1)$
 - Delete edge: $O(1)$

Adjacency List Representation



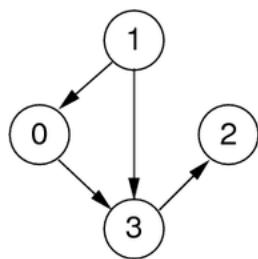
Undirected graph

- New storage cost: $V-1$ int ptrs + $V(V+1)/2$ ints
- Requires us to always use edges (v, w) such that $v < w$



$A[0] = \langle 1, 3 \rangle$
 $A[1] = \langle 0, 3 \rangle$
 $A[2] = \langle 3 \rangle$
 $A[3] = \langle 0, 1, 2 \rangle$

Undirected graph



$A[0] = \langle 3 \rangle$
 $A[1] = \langle 0, 3 \rangle$
 $A[2] = \langle \rangle$
 $A[3] = \langle 2 \rangle$

Directed graph

- Advantages:
 - Relatively easy to implement in languages like C
 - Can represent graphs and digraphs
 - Memory efficient if $E:V$ relatively small
- Disadvantages:
 - One graph has many possible representations

- Graph initialisation:

```
newGraph(V) :
| Input number of nodes V
| Output new empty graph
|
| g.nV = V    // #vertices (numbered 0..V-1)
| g.nE = 0    // #edges
| allocate memory for g.edges[]
| for all i=0..V-1 do
|   g.edges[i]= newList() // empty list
| end for
| return g
```

- Edge insertion:

```
insertEdge(g, (v,w)) :
| Input graph g, edge (v,w)
| Output graph g containing (v,w)
|
| if not ListMember(g.edges[v],w) then
|   // (v,w) not in graph
|   ListInsert(g.edges[v],w)
|   ListInsert(g.edges[w],v)
|   g.nE=g.nE+1
| end if
```

- Edge removal:

```
removeEdge(g, (v,w)) :
| Input graph g, edge (v,w)
| Output graph g without (v,w)
|
| if ListMember(g.edges[v],w) then
|   // (v,w) in graph
|   ListDelete(g.edges[v],w)
|   ListDelete(g.edges[w],v)
|   g.nE=g.nE-1
| end if
```

- Print a list of edges

```
showEdges(g) :
| Input graph g
|
| for all i=0 to g.nV-1 do
|   for all v in g.edges[i] do
|     if i < v then
|       print i"--"v
|     end if
|   end for
| end for
```

- Adjacency List Cost Analysis

- Storage cost: $O(V+E)$
- Cost of operations:
 - Initialisation: $O(V)$
 - Inserted edge: $O(V)$
 - Delete edge: $O(V)$

Comparison of Graph Representations

	array of edges	adjacency matrix	adjacency list
space usage	E	V^2	$V+E$
initialise	1	V^2	V
insert edge	E	1	E
remove edge	E	1	E

- Other operations:

	array of edges	adjacency matrix	adjacency list
disconnected(v)?	E	V	1
isPath(x,y)?	$E \cdot \log V$	V^2	$V+E$
copy graph	E	V^2	$V+E$
destroy graph	1	V	$V+E$

5.3: Graph ADT

- Data: set of edges, set of vertices
- Operations:
 - Building: create graph, add edge
 - Deleting: remove edge, drop whole graph
 - Scanning: Check if graph contains a given edge
- Things to note:
 - Set of vertices is fixed when graph initialised
 - We treat vertices as Ints but could be arbitrary Items

Graph.h

```
// graph representation is hidden
typedef struct GraphRep *Graph;

// vertices denoted by integers 0..N-1
typedef int Vertex;

// edges are pairs of vertices (end-points)
typedef struct Edge { Vertex v; Vertex w; } Edge;

// operations on graphs
Graph newGraph(int V); // new graph with V vertices
void insertEdge(Graph, Edge);
void removeEdge(Graph, Edge);
bool adjacent(Graph, Vertex, Vertex);
    // is there an edge between two vertices?
void freeGraph(Graph);
```

Graph ADT (Array of Edges)

```
typedef struct GraphRep {
    Edge *edges; // array of edges
    int nV;      // #vertices (numbered 0..nV-1)
    int nE;      // #edges
    int n;        // size of edge array
} GraphRep;
```

```
Graph newGraph(int V) {
    assert(V >= 0);
    Graph g = malloc(sizeof(GraphRep));
    assert(g != NULL);
    g->nV = V; g->nE = 0;
    // allocate enough memory for edges
    g->n = Enough;
    g->edges = malloc(g->n*sizeof(Edge));
    assert(g->edges != NULL);
    return g;
}
```

```
// check if two edges are equal
bool eq(Edge e1, Edge e2) {
    return ( (e1.v == e2.v && e1.w == e2.w)
            || (e1.v == e2.w && e1.w == e2.v) );
```

```

}

// check if vertex is valid in a graph
bool validV(Graph g, Vertex v) {
    return (g != NULL && v >= 0 && v < g->nV);
}

// check if an edge is valid in a graph
bool validE(Graph g, Edge e) {
    return (g != NULL && validV(e.v) && validV(e.w));
}

void insertEdge(Graph g, Edge e) {
    // ensure that g exists and array of edges isn't full
    assert(g != NULL && g->nE < g->n && isValidE(g,e));
    int i = 0; // can't define in for [...]
    for (i = 0; i < g->nE; i++)
        if (eq(e,g->edges[i])) break;
    if (i == g->nE) // edge e not found
        g->edges[g->nE++] = e;
}

void removeEdge(Graph g, Edge e) {
    // ensure that g exists
    assert(g != NULL && validE(g,e));
    int i = 0;
    while (i < g->nE && !eq(e,g->edges[i]))
        i++;
    if (i < g->nE) // edge e found
        g->edges[i] = g->edges[--g->nE];
}

bool adjacent(Graph g, Vertex x, Vertex y) {
    assert(g != NULL && validV(g,x) && validV(g,y));
    Edge e;
    e.v = x; e.w = y;
    for (int i = 0; i < g->nE; i++) {
        if (eq(e,g->edges[i])) // edge found
            return true;
    }
    return false; // edge not found
}

```

```

void insertEdge(Graph g, Edge e) {
    // ensure that g exists
    assert(g != NULL && validE(g,e));
    int i = 0;
    for (i = 0; i < g->nE; i++)
        if (eq(e,g->edges[i])) break;
    if (i == g->nE) { // edge e not found
        if (g->n == g->nE) { // array full; expand
            g->edges = realloc(g->edges, 2*g->n);
            assert(g->edges != NULL);
            g->n = 2*g->n;
        }
        g->edges[g->nE++] = e;
    }
}

```

```

void freeGraph(Graph g) {
    assert(g != NULL);
    free(g->edges); // free array of edges
    free(g); // remove Graph object
}

```

Graph ADT (Adjacency Matrix)

```

typedef struct GraphRep {
    int **edges; // adjacency matrix
    int nV; // #vertices
    int nE; // #edges
} GraphRep;

```

```

Graph newGraph(int V) {
    assert(V >= 0);
    Graph g = malloc(sizeof(GraphRep));
    assert(g != NULL);
    g->nV = V; g->nE = 0;
    // allocate array of pointers to rows
    g->edges = malloc(V * sizeof(int *));
    assert(g->edges != NULL);
    // allocate memory for each column and initialise with 0
    for (int i = 0; i < V; i++) {
        g->edges[i] = calloc(V, sizeof(int));
        assert(g->edges[i] != NULL);
    }
    return g;
}

```

```

void insertEdge(Graph g, Edge e) {
    assert(g != NULL && validE(g,e));

    if (!g->edges[e.v][e.w]) { // edge e not in graph
        g->edges[e.v][e.w] = 1;
        g->edges[e.w][e.v] = 1;
        g->nE++;
    }
}

```

```

void removeEdge(Graph g, Edge e) {
    assert(g != NULL && validE(g,e));

    if (g->edges[e.v][e.w]) { // edge e in graph
        g->edges[e.v][e.w] = 0;
        g->edges[e.w][e.v] = 0;
        g->nE--;
    }
}

```

```

bool adjacent(Graph g, Vertex x, Vertex y) {
    assert(g != NULL && validV(g,x) && validV(g,y));

    return (g->edges[x][y] != 0);
}

```

```

void freeGraph(Graph g) {
    assert(g != NULL);
    for (int i = 0; i < g->nV; i++)
        // free one row of matrix
        free(g->edges[i]);
    free(g->edges); // free array of row pointers
    free(g);         // remove Graph object
}

```

Graph ADT (Adjacency Lists)

```

typedef struct GraphRep {
    Node **edges; // array of lists
    int nV;       // #vertices
    int nE;       // #edges
} GraphRep;

```

```

typedef struct Node {
    Vertex v;
    struct Node *next;
} Node;

```

```

bool inLL(Node *L, Vertex v) {
    while (L != NULL) {
        if (L->v == v) return true;
        L = L->next;
    }
    return false;
}

```

```

Graph newGraph(int V) {
    assert(V >= 0);
    Graph g = malloc(sizeof(GraphRep));
    assert(g != NULL);
    g->nV = V; g->nE = 0;
    // allocate memory for array of lists
    g->edges = malloc(V * sizeof(Node *));
    assert(g->edges != NULL);
    for (int i = 0; i < V; i++)

```

```

        g->edges[i] = NULL;
    return g;
}

void insertEdge(Graph g, Edge e) {
    assert(g != NULL && validE(g,e));
    if (!inLL(g->edges[e.v], e.w)) { // edge e not in graph
        g->edges[e.v] = insertLL(g->edges[e.v], e.w);
        g->edges[e.w] = insertLL(g->edges[e.w], e.v);
        g->nE++;
    }
}
void removeEdge(Graph g, Edge e) {
    assert(g != NULL && validE(g,e));
    if (inLL(g->edges[e.v], e.w)) { // edge e in graph
        g->edges[e.v] = deleteLL(g->edges[e.v], e.w);
        g->edges[e.w] = deleteLL(g->edges[e.w], e.v);
        g->nE--;
    }
}

bool adjacent(Graph g, Vertex x, Vertex y) {
    assert(g != NULL && validV(g,x) && validV(g,y));
    return inLL(g->edges[x], y);
}

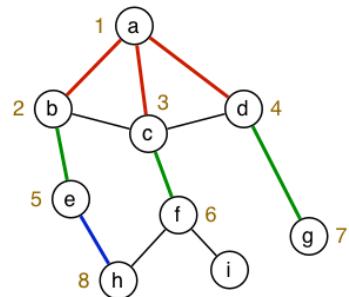
void freeGraph(Graph g) {
    assert(g != NULL);
    for (int i = 0; i < g->nV; i++)
        freeLL(g->edges[i]); // free one list
    free(g->edges); // free array of list pointers
    free(g); // remove Graph object
}

```

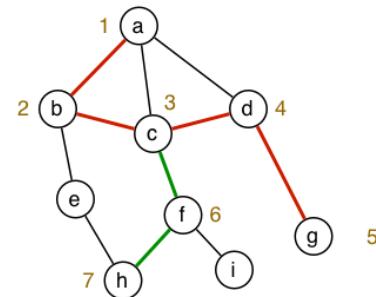
5.4: Graph Traversal

Graph Traversal

- Depth-first search (DFS)
 - Favours following path rather than neighbours
 - Implemented recursively or iteratively (via stack)
 - Full traversal produces a dept-first spanning tree
- Breadth-first search (BFS)
 - Favours neighbours rather than path following
 - Can be implemented iteratively (via queue)
 - Full traversal produces a breadth-first spanning tree



Breadth-first Search



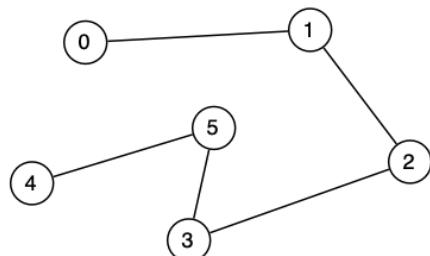
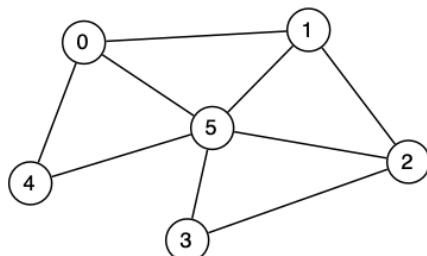
Depth-first Search

Spanning Tree

- A spanning tree of a graph includes all vertices, using a subset of edges, without cycles

Original graph

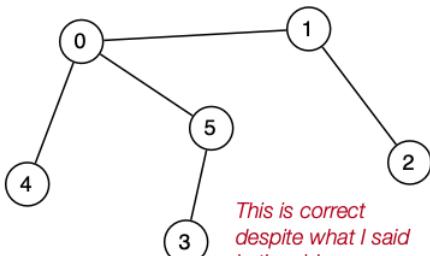
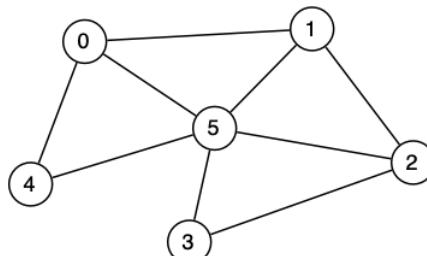
Spanning tree



DFS Traversal: 0 -> 1 -> 2 -> 3 -> 5 -> 4

Original graph

Spanning tree



BFS Traversal: 0 -> 1,4,5; 1 -> 2; 5 -> 3

Depth-first search

```

visited = {}

depthFirst(G, v):
| visited = visited U {v}
| for all (v,w) ∈ edges(G) do
| | if w ∉ visited then
| | | depthFirst(G, w)
| | end if
| end for

```

- Path checking:

```

visited = {}

hasPath(G, src, dest):
| Input graph G, vertices src,dest
| Output true if there is a path from src to dest,
|           false otherwise
|
| return dfsPathCheck(G, src, dest)

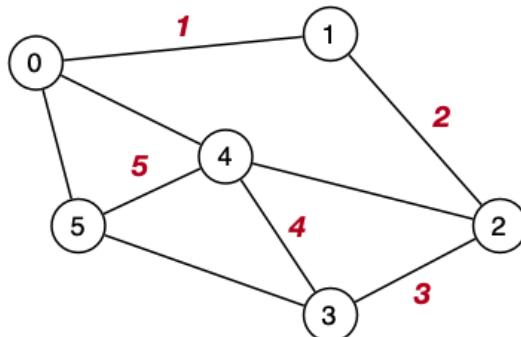
```

```

dfsPathCheck(G, v, dest):
| visited = visited U {v}
| for all (v,w) ∈ edges(G) do
| | if w=dest then // found edge to dest
| | | return true
| | else if w ∉ visited then
| | | if dfsPathCheck(G, w, dest) then
| | | | return true // found path via w to dest
| | | end if
| | end if
| end for
| return false // no path from v to dest

```

- Path check example: 0 → 5



- Cost analysis:

- Each vertex visited at most once, cost = $O(V)$
- Visit all edges incident on visited vertices, cost = $O(E)$
 - Assuming an adjacency list representation

- Time complexity of DFS: $O(V+E)$ (adjacency list representation)

Path Finding

- Depth-first search can find a path to a node

```

visited[] // store previously visited node
          // for each vertex 0..nV-1

findPath(G,src,dest):
| Input graph G, vertices src,dest
|
| for all vertices v $\in$ G do
|   visited[v]=-1
| end for
| visited[src]=src // starting node of the path
| if dfsPathCheck(G,src,dest) then
|   // show path in dest..src order
|   v=dest
|   while v $\neq$ src do
|     print v"-"
|     v=visited[v]
|   end while
|   print src
| end if

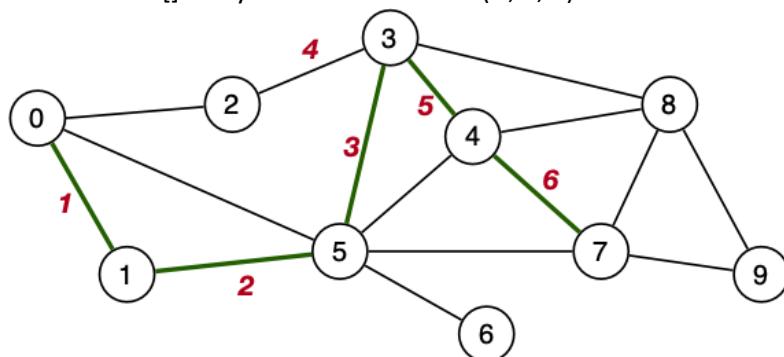
```

```

dfsPathCheck(G,v,dest):
| for all (v,w)  $\in$  edges(G) do
|   if visited[w] = -1 then
|     visited[w] = v
|     if w = dest then // found edge from v to dest
|       return true
|     else if dfsPathCheck(G,w,dest) then
|       return true // found path via w to dest
|     end if
|   end if
| end for
| return false // no path from v to dest

```

- The visited[] array after dfsPathCheck(G, 0, 7) succeeds



visited	0	0	3	5	3	1	5	4	-1	-1
	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]	[8]	[9]

Breadth-first Search

- Visit and mark current vertex
- Visit all neighbours of current vertex
- Then consider neighbours of neighbours
- Uses queue

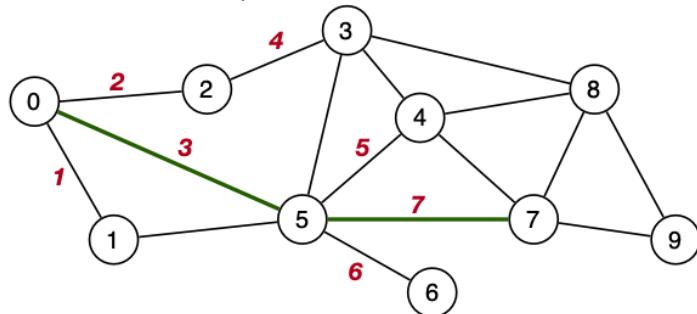
```

visited[] // store previously visited node
           // for each vertex 0..nV-1

findPathBFS (G, src, dest) :
| Input graph G, vertices src,dest
|
| for all vertices v $\in$ G do
|     visited[v]=-1
| end for
| found=false
| visited[src]=src
| enqueue src into queue Q
| while not found  $\wedge$  Q is not empty do
| | dequeue v from Q
| | if v=dest then
| | | found=true
| | else
| | | for each (v,w)  $\in$  edges(G) with visited[w]=-1 do
| | | | visited[w]=v
| | | | enqueue w into Q
| | | end for
| | end if
| end while
| if found then
| | display path in dest..src order
| end if

```

- The visited[] array after findPathBFS(G, 0 , 7) succeeds



visited	0	0	0	2	5	0	5	5	-1	-1
	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]	[8]	[9]

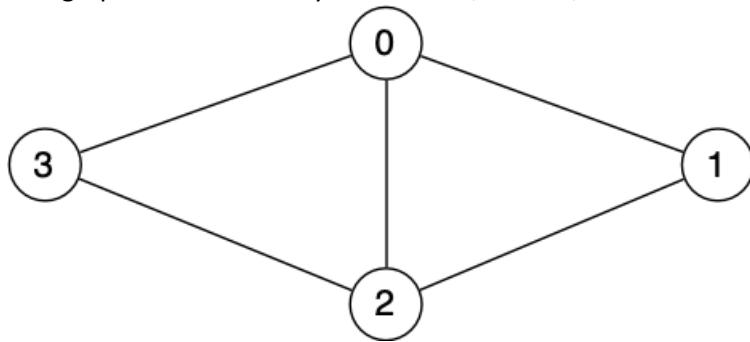
https://www.cse.unsw.edu.au/~cs2521/19T2/lecs/week04b/dfs_bfs_trace/Graph_BFS.htm

- Cost Analysis:
 - $O(V+E)$ same as DFS

5.5: Graph Algorithms

Cycle Checking

- A graph has a cycle if:
 - It has a path of length > 2
 - With start vertex *src* = end vertex *dest*
 - And without using any edge more than once
- This graph has 3 distinct cycles: 0-1-2-0, 2-3-0-2, 0-1-2-3-0



```
hasCycle(G):
| Input graph G
| Output true if G has a cycle, false otherwise
|
| choose any vertex v ∈ G
| return dfsCycleCheck(G, v)

dfsCycleCheck(G, v):
| mark v as visited
| for each (v, w) ∈ edges(G) do
| | if w has been visited then // found cycle
| | | return true
| | else if dfsCycleCheck(G, w) then
| | | return true
bool dfsCycleCheck(Graph g, Vertex v, Vertex u) {
    visited[v] = true;
    for (Vertex w = 0; w < g->nV; w++) {
        if (adjacent(g, v, w)) {
            if (!visited[w]) {
                if (dfsCycleCheck(g, w, v))
                    return true;
            }
            else if (w != u)
                return true;
        }
    }
    return false;
}
```

```
Vertex *visited;

bool hasCycle(Graph g, Vertex s) {
    bool result = false;
    visited = calloc(g->nV, sizeof(int));
```

```
for (int v = 0; v < g->nV; v++) {
    for (int i = 0; i < g->nV; i++)
        visited[i] = -1;
    if dfsCycleCheck(g, v, v)) {
        result = true;
        break;
    }
}
free(visited);
return result;
}
```

- But what about non-connected components of a graph?
 - Add new fields to the GraphRep structure

```
typedef struct GraphRep *Graph;

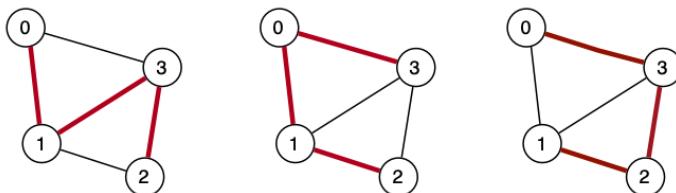
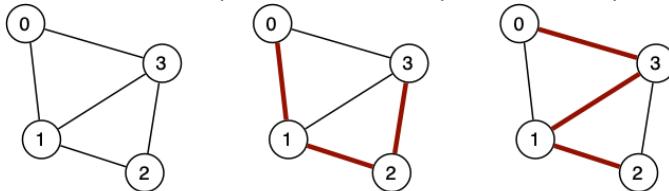
struct GraphRep {
    ...
    int nC; // # connected components
    int *cc; // which component each vertex is contained in
    ... // i.e. array [0..nV-1] of 0..nC-1
}
```

```
// How many connected subgraphs are there?
int nConnected(Graph g) {
    return g->nC;
}

// Are two vertices in the same connected subgraph?
bool inSameComponent(Graph g, Vertex v, Vertex w) {
    return (g->cc[v] == g->cc[w]);
}
```

Hamiltonian Path and Circuit

- Find a simple path connected two vertices v, w in a graph G
- Such that the path includes every vertex exactly once



- Approach:
 - Generate all possible simple paths (using e.g. DFS)
 - Keep a counter of vertices visited in current path
 - Stop when find a path containing V vertices
- Can be expressed via a recursive DFS algorithm

```
visited[] // array [0..nV-1] to keep track of visited vertices
```

```
hasHamiltonianPath(G, src, dest):
| Input graph G, plus src/dest vertices
| Output true if Hamiltonian path src...dest,
|           false otherwise
|
| for all vertices v ∈ G do
|     visited[v]=false
| end for
| return false // no cycle at v
```

```

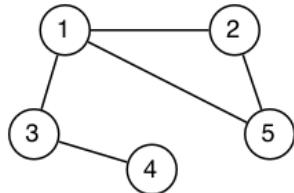
| return hamiltonR(G,src,dest,#vertices(G)-1)
hamiltonR(G,v,dest,d):
| Input G      graph
|         v      current vertex considered
|         dest   destination vertex
|         d      distance "remaining" until path found
|
| if v=dest then
|     if d=0 then return true else return false
| else
|     | visited[v]=true
|     | for each (v,w)  $\in$  edges(G) where not visited[w] do
|     |     if hamiltonR(G,w,dest,d-1) then
|     |         return true
|     |     end if
|     | end for
| end if
| visited[v]=false           // reset visited mark
| return false

```

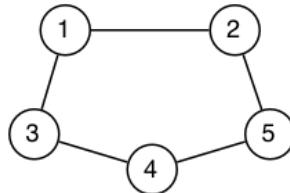
- Analysis: Worst case requires $(V-1)!$ paths to be examined

Euler Path and Circuit

- Find a path connected two vertices v, w in graph G
- Such that the path includes each edge exactly once



Euler Path: 4-3-1-5-2-1



Euler Circuit: 1-2-5-4-3-1

```

hasEulerPath(G,src,dest):
| Input graph G, vertices src,dest
| Output true if G has Euler path from src to dest
|         false otherwise
|
| if src $\neq$ dest then
|     if degree(G,src) is even  $\wedge$  degree(G,dest) is even then
|         return false
|     end if
| end if
| for all vertices v  $\in$  G do
|     if v $\neq$ src  $\wedge$  v $\neq$ dest  $\wedge$  degree(G,v) is odd then
|         return false
|     end if
| end for
| return true

```

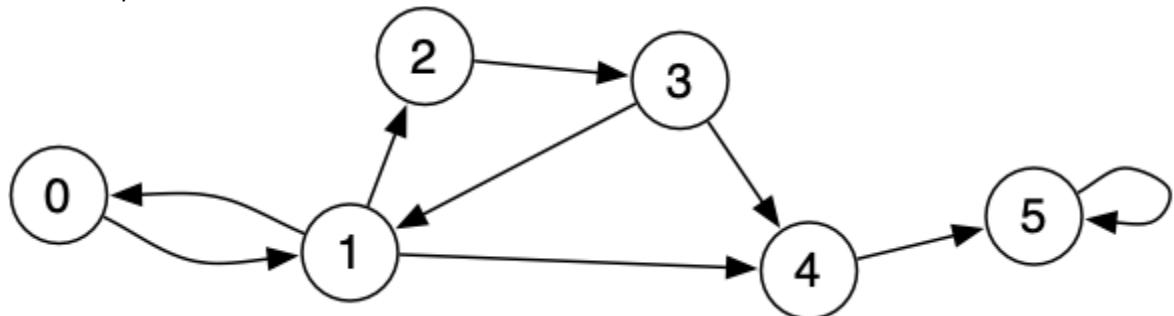
- Overall cost is $O(V^2)$

Week 7

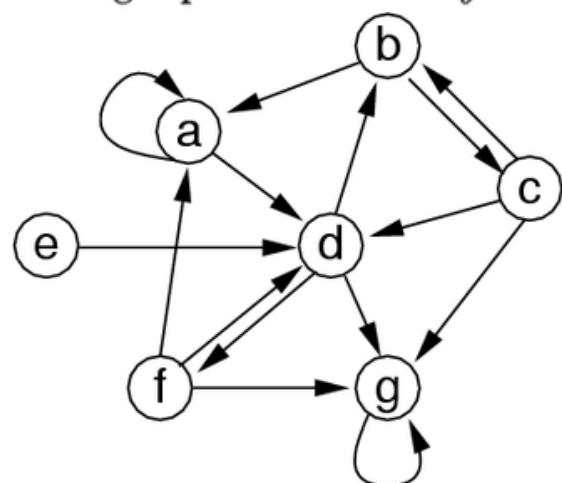
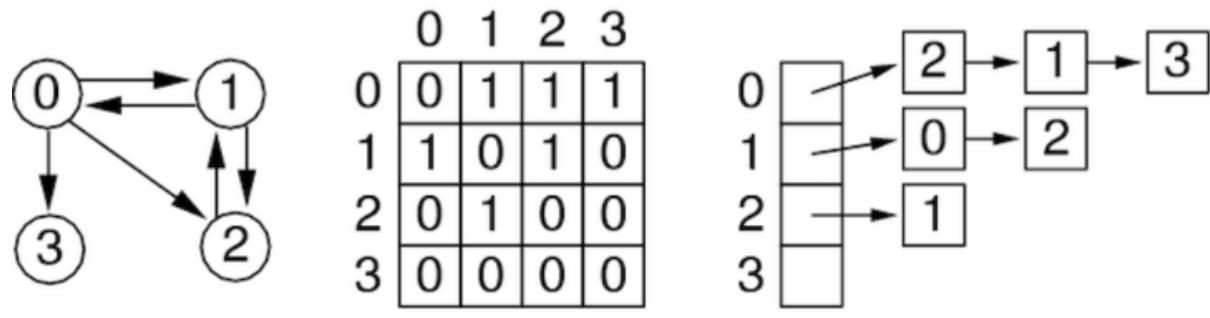
7.1: Directed/Weighted Graphs

Directed Graphs (Digraphs)

- Example:



- Some properties of:
 - Edges 1-2-3 form a cycle, edges 1-3-4 do not form a cycle
 - Vertex 5 has a self-referencing edge (5, 5)
 - Vertices 0 and 1 reference each other, i.e. (0, 1) and (1, 0)
 - There are no paths from 5 to any other nodes
 - Paths from -> 5: 0-1-2-3-4-5, 0-1-4-5, 0-1-2-3-1-4-5
- Outdegree: Number of edges coming out of a node
- Indegree: Number of edges coming into the node



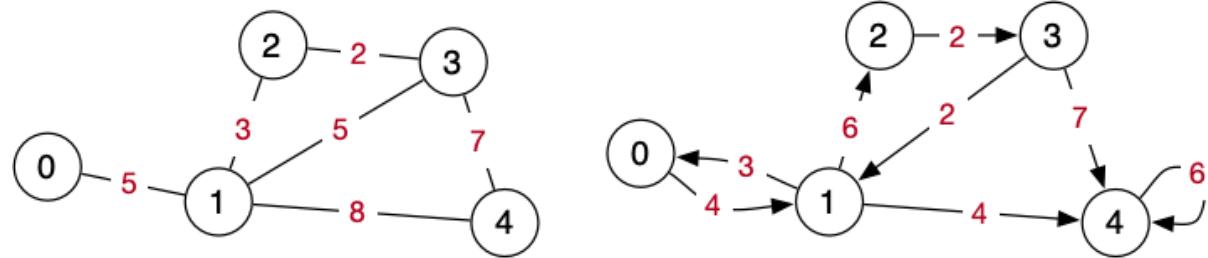
	a	b	c	d	e	f	g
a	1	0	0	1	0	0	0
b	1	0	1	0	0	0	0
c	0	1	0	1	0	0	1
d	0	1	0	0	0	1	1
e	0	0	0	1	0	0	0
f	1	0	0	1	0	0	1
g	0	0	0	0	0	0	1

Costs of representations: (where degree $\text{deg}(v) = \#\text{edges leaving } v$)

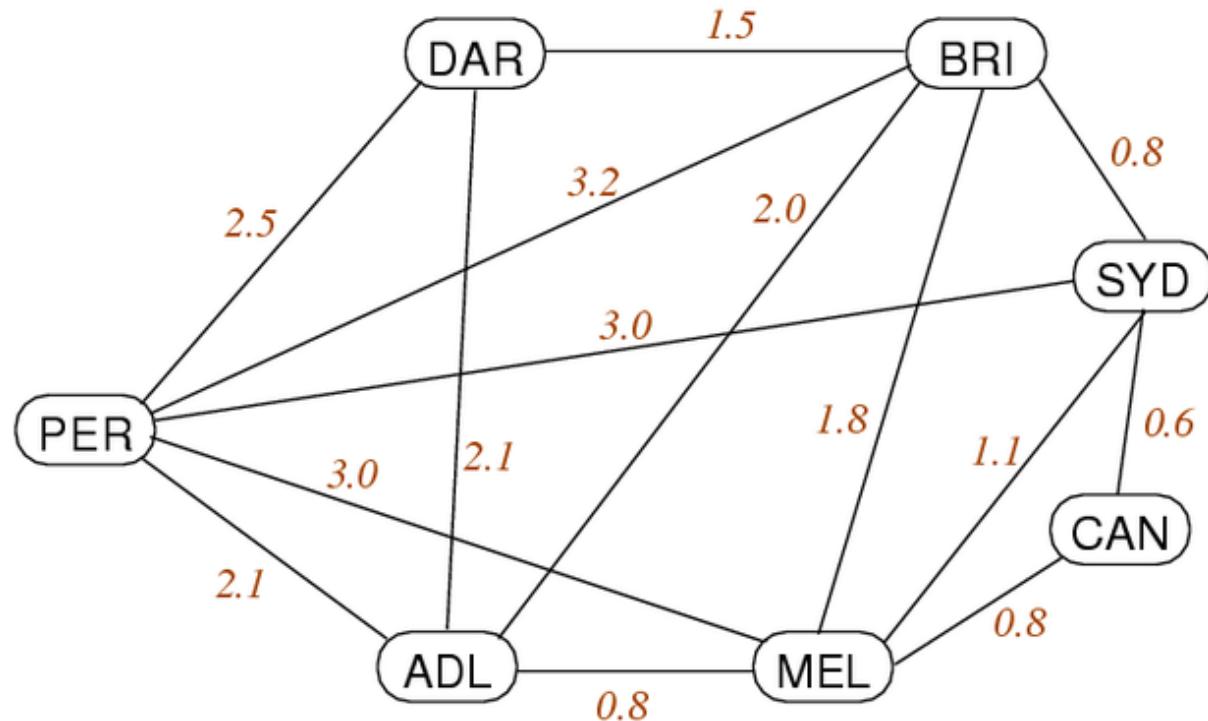
	array of edges	adjacency matrix	adjacency list
space usage	E	V^2	$V+E$
insert edge	E	1	1
exists edge $(v,w)?$	E	1	$\text{deg}(v)$
get edges leaving v	E	V	$\text{deg}(v)$

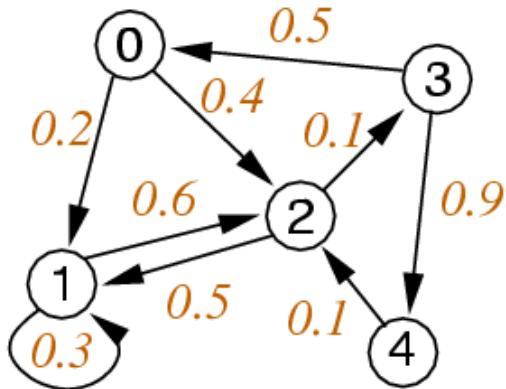
Overall, adjacency list representation is best

Weighted Graphs



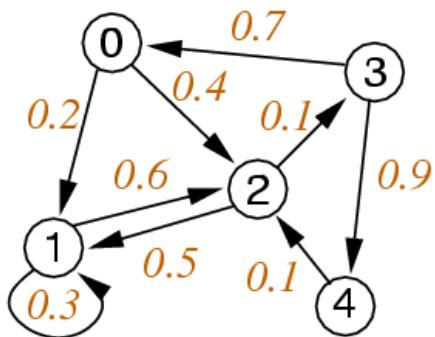
Weighted Graph



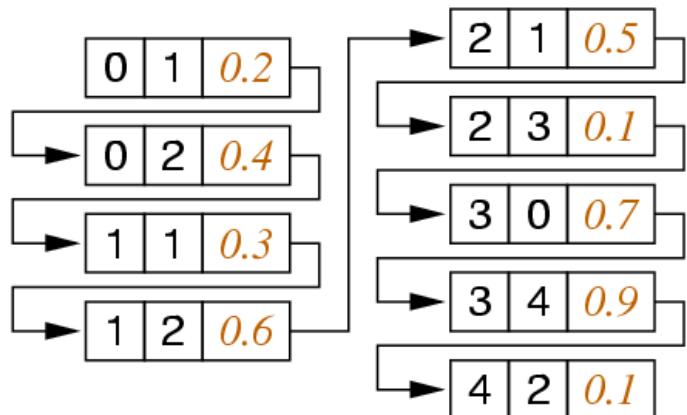


	0	1	2	3	4
0	*	0.2	0.4	*	*
1	*	0.3	0.6	*	*
2	*	0.5	*	0.1	*
3	0.5	*	*	*	0.9
4	*	*	0.1	*	*

Weighted Digraph



Adjacency Matrix



Weighted Digraph

Edge List

```
// edges are pairs of vertices (end-points) plus weight
typedef struct Edge {
    Vertex v;
    Vertex w;
    int    weight;
} Edge;

// returns weight, or 0 if vertices not adjacent
int adjacent(Graph, Vertex, Vertex);
```

```
typedef struct GraphRep {
    int **edges; // adjacency matrix storing weights
                  // 0 if nodes not adjacent
    int    nV;   // #vertices
    int    nE;   // #edges
} GraphRep;

bool adjacent(Graph g, Vertex v, Vertex w) {
    assert(valid graph, valid vertices)
    return (g->edges[v][w] != 0);
}
```

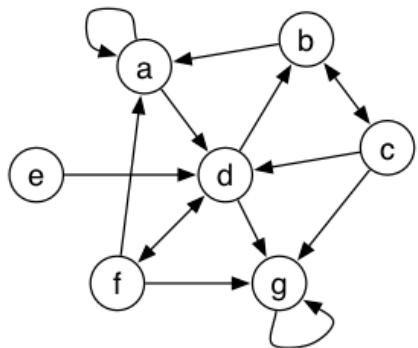
```
void insertEdge(Graph g, Edge e) {
    assert(valid graph, valid edge)
    // edge e not already in graph
    if (g->edges[e.v][e.w] == 0) g->nE++;
    // may change weight of existing edge
    g->edges[e.v][e.w] = e.weight;
    g->edges[e.w][e.v] = e.weight;
}

void removeEdge(Graph g, Edge e) {
    assert(valid graph, valid edge)
    // edge e not in graph
    if (g->edges[e.v][e.w] == 0) return;
    g->edges[e.v][e.w] = 0;
    g->edges[e.w][e.v] = 0;
    g->nE--;
}
```

7.2: Digraph Algorithms

Transitive Closure

- Goal: Determine reachability, is there a path from t to s?
- Transitive closure:



	a	b	c	d	e	f	g
a	1	0	0	1	0	0	0
b	1	0	1	0	0	0	0
c	0	1	0	1	0	0	1
d	0	1	0	0	0	1	1
e	0	0	0	1	0	0	0
f	1	0	0	1	0	0	1
g	0	0	0	0	0	0	1

adjacency matrix

	a	b	c	d	e	f	g
a	1	1	1	1	0	1	1
b	1	1	1	1	0	1	1
c	1	1	1	1	0	1	1
d	1	1	1	1	0	1	1
e	1	1	1	1	0	1	1
f	1	1	1	1	0	1	1
g	0	0	0	0	0	0	1

reachability matrix

- Warshall's Algorithm:

```

makeTC(G):
| tc[][][] = edges[][][]
| for all i ∈ vertices(G) do
| | for all s ∈ vertices(G) do
| | | for all t ∈ vertices(G) do
| | | | if tc[s][i]=1 ∧ tc[i][t]=1 then
| | | | | tc[s][t]=1
| | | | end if
| | | end for
| | end for
| end for
  
```

- Cost: $O(V^3)$

```

depthFirst(G,v):
| mark v as visited
| for each (v,w) ∈ edges(G) do
| | if w has not been visited then
| | | depthFirst(w)
| | end if
| end for

breadthFirst(G,v):
| enqueue v
| while queue not empty do
| | curr=dequeue
| | if curr not already visited then
| | | mark curr as visited
| | | enqueue each w where (curr,w) ∈ edges(G)
| | end if
  
```

Page Rank

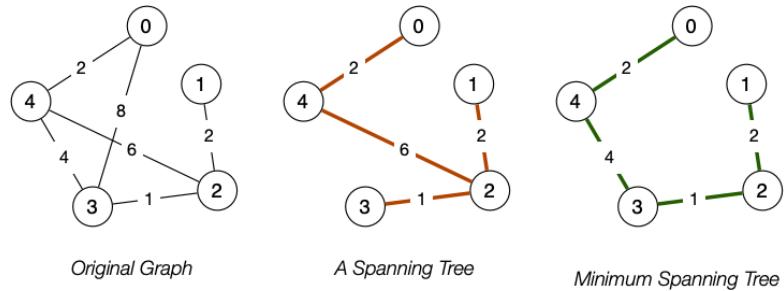
- Goal: To determine which Web pages are ‘important’
- Approach: Ignore page contents; focus on hyperlinks
 - Treat Web as a graph: page = vertex, hyperlink = edge

```
PageRank(myPage) :  
| rank=0  
| for each page in the Web do  
| | if linkExists(page,myPage) then  
| | | rank=rank+1  
| | end if  
| end for
```

- Random web surfer strategy:
 - Each age typically has many outbound hyperlinks ...
 - Choose one at random, without a visited[] check
 - Follow link and repeat above process on destination page
 - If no visited check, need a way to (mostly) avoid loops
- Important property of this strategy:
 - If we randomly follow links in the webpage
 - ... more likely to re-discover pages with many inbound links

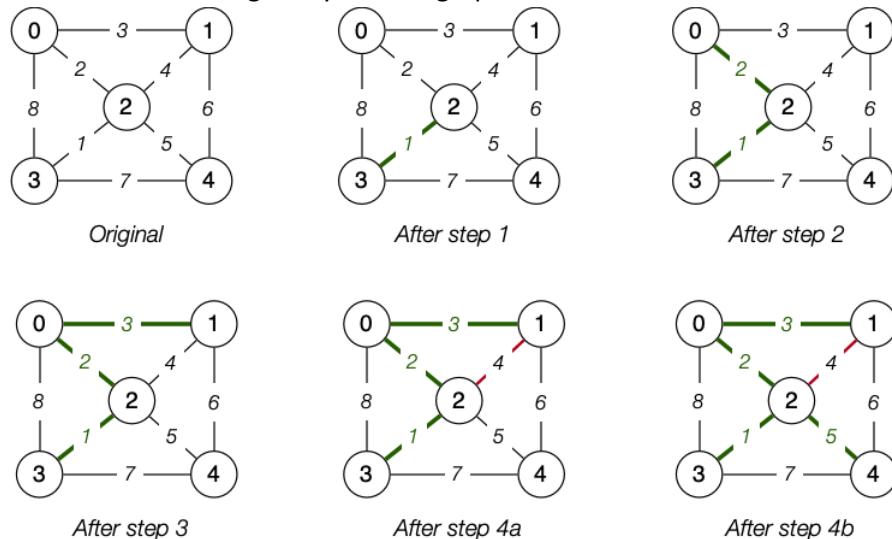
```
curr=random page, prev=null  
for a long time do  
| if curr not in array rank[] then  
| | rank[curr]=0  
| end if  
| rank[curr]=rank[curr]+1  
| if random(0,100) < 85 then // with 85% chance ...  
| | prev=curr // ... keep crawling  
| | curr=choose hyperlink from curr  
| else  
| | curr=random page, not prev // avoid getting stuck  
| | prev=null  
| end if  
end for
```

7.3: Minimum Spanning Tree



Kruskal's Algorithm

- Start with empty MST
- Consider edges in increasing weight order
 - Add edge if it does not form a cycle in MST
- Repeat until $V-1$ edges are added
- Critical operations:
 - Iterating over edges in weight order
 - Checking for cycles in a graph



KruskalMST (G) :

```

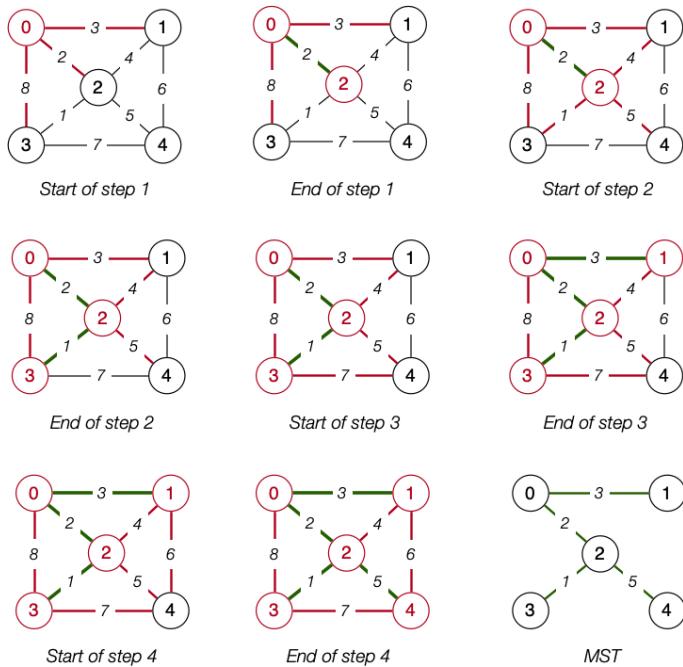
| Input graph G with n nodes
| Output a minimum spanning tree of G
|
| MST=empty graph
| sort edges(G) by weight
| for each e ∈ sortedEdgeList do
|   | MST = MST ∪ {e} // add edge
|   | if MST has a cyle then
|   |   | MST = MST \ {e} // drop edge
|   | end if
|   | if MST has n-1 edges then
|   |   | return MST
|   | end if
| end for

```

- Complexity analysis:
 - Sort edge list is $O(E \times \log E)$
 - At least V iterations over sorted edges
 - On each iteration:
 - Getting next lowest cost edge is $O(1)$
 - Checking whether adding it forms a cycle: cost = $O(V^2)$

Prim's Algorithm

- Start from any vertex V and empty MST
- Choose edge not already in MST to add to MST; must be:
 - Node must not already be connected to MST
 - Minimal weight of all such edges
- Repeat until MST covers all vertices



PrimMST(G) :

```

| Input graph G with n nodes
| Output a minimum spanning tree of G
|
| MST=empty graph
| usedV={ 0 }
| unusedE=edges (g)
| while |usedV| < n do
|   | find e=(s,t,w) ∈ unusedE such that {
|   |   s ∈ usedV and t ∉ usedV
|   |   and w is min weight of all such edges
|   |
|   |   MST = MST ∪ {e}
|   |   usedV = usedV ∪ {t}
|   |   unusedE = unusedE \ {e}
| end while
| return MST

```

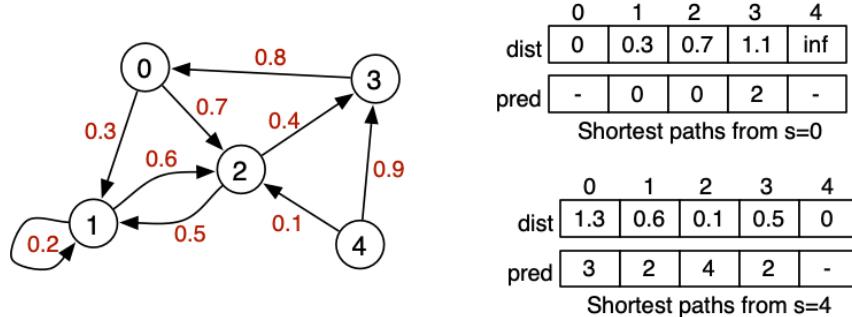
- Complexity analysis:

- V iterations of outer loop
- In each iteration, finding min-weighted edge...
 - With set of edges is $O(E)$ $\rightarrow O(V \times E)$ overall
 - With priority queue is $O(\log E)$ $\rightarrow O(V \times \log E)$ overall

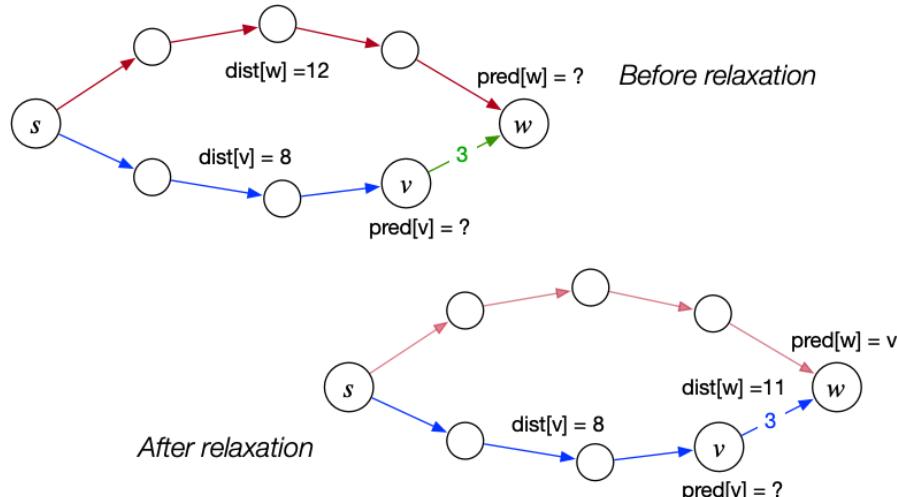
7.4: Shortest Path

Single-source Shortest Path (SSSP)

- Shortest paths from s to all other vertices
- $\text{dist}[]$ V-indexed array of cost of shortest path from s
- $\text{pred}[]$ V-indexed array of predecessor in shortest path from s



Edge Relaxation



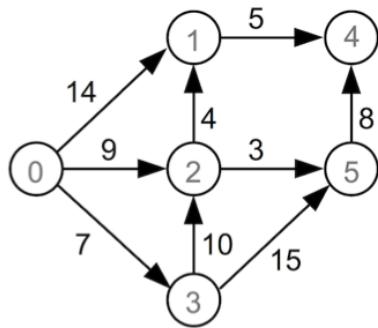
Dijkstra's Algorithm

```

dist[] // array of cost of shortest path from s
pred[] // array of predecessor in shortest path from s
vSet // vertices whose shortest path from s is unknown

dijkstrassSP(G,source) :
| Input graph G, source node
|
| initialise all dist[] to ∞
| dist[source]=0
| initialise all pred[] to -1
| vSet=all vertices of G
| while vSet ≠ ∅ do
| | find v ∈ vSet with minimum dist[v]
| | for each (v,w,weight) ∈ edges(G) do
| | | relax along (v,w,weight)
| | end for
| | vSet=vSet \ {v}
| end while

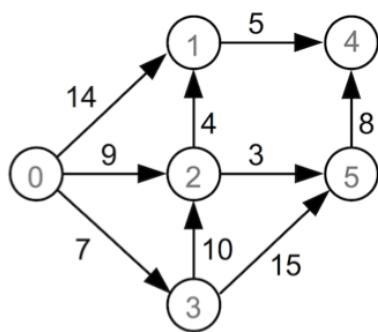
```



```

while vSet not empty do
    find v in vSet
    with min dist[v]
    for each (v,w,weight) in E do
        relax along (v,w,weight)
    end for
    vSet = vSet \ {v}
end while

```



```

while vSet not empty do
    find v in vSet
    with min dist[v]
    for each (v,w,weight) in E do
        relax along (v,w,weight)
    end for
    vSet = vSet \ {v}
end while

```

Initially

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	inf	inf	inf	inf	inf
pred	-	-	-	-	-	-

First iteration, v=0

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	14	9	7	inf	int
pred	-	0	0	0	-	-

Second Iteration, v=3

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	14	9	7	inf	22
pred	-	0	0	0	-	3

Third iteration, v=2

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	13	9	7	inf	12
pred	-	2	0	0	-	2

Fourth iteration, v=5

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	13	9	7	20	12
pred	-	2	0	0	5	2

Fifth iteration, v=1

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	13	9	7	18	12
pred	-	2	0	0	1	2

Sixth iteration,

	[0]	[1]	[2]	[3]	[4]	[5]
dist	0	13	9	7	18	12
pred	-	2	0	0	1	2

Completed, vSet is empty

- Time complexity analysis:
 - Each edge needs to be considered once $\rightarrow O(E)$
 - Outer loop has $O(V)$ iterations

Week 8

8.1: Sorting

Concrete Framework

```
// we deal with generic Items
typedef SomeType Item;

// abstractions to hide details of Items
#define key(A) (A)
#define less(A,B) (key(A) < key(B))
#define swap(A,B) {Item t; t = A; A = B; B = t;}

// Sorts a slice of an array of Items, a[lo..hi]
void sort(Item a[], int lo, int hi);

// Check for sortedness (to validate functions)
int isSorted(Item a[], int lo, int hi);

bool isSorted(Item a[], int lo, int hi)
{
    for (int i = lo; i < hi; i++) {
        if (!less(a[i], a[i+1])) return false;
    }
    return true;
}
```

Sorts on Linux

- The sort command:
 - Sorts a file of text, understands fields in a line
 - Can sort alphabetically, numerically, reverse, random
- The qsort command:
 - qsort(void *a, int n, int size, int (*cmp)())
 - Sorts any kind of array
 - Requires the user to supply a comparison function
 - Sorts list of items using the order given by cmp()

8.2: $O(n^2)$ Sorts

$O(n^2)$ Sorting Algorithms

- Selection sort: Simple, non-adaptive sort
- Bubble sort: Simple, adaptive sort
- Insertion sort: Simple, adaptive sort
- Shell sort: Improved version of insertion sort
- For small arrays, the above methods are adequate
- Adaptive: Time complexity depends on how ‘sorted’ the original list is

Selection Sort

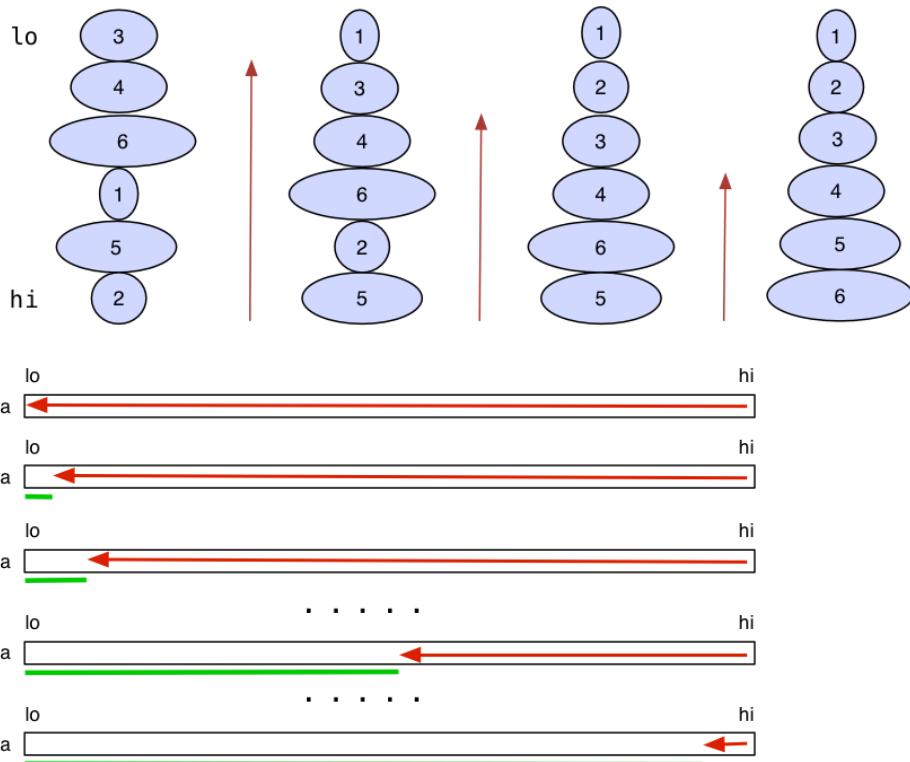
- Simple, non-adaptive method:
 - Find the smallest element, put it into first array slot
 - Find second smallest element, put it into second array sort
 - Repeat until all elements in the correct position



```
void selectionSort(int a[], int lo, int hi)
{
    int i, j, min;
    for (i = lo; i < hi-1; i++) {
        min = i;
        for (j = i+1; j <= hi; j++) {
            if (less(a[j], a[min])) min = j;
        }
        swap(a[i], a[min]);
    }
}
```

Bubble Sort

- Simple adaptive method:
 - Make multiple passes from the highest to lowest
 - On each pair passed, swap if they are out of order



```

S O R T E X A M P L E
A S O R T E X E M P L
A E S O R T E X L M P
A E E S O R T L X M P
A E E L S O R T M X P
A E E L M S O R T P X
A E E L M O S P R T X
A E E L M O P S R T X
A E E L M O P R S T X
... no swaps ⇒ done ...
A E E L M O P R S T X

```

```

void bubbleSort(int a[], int lo, int hi)
{
    int i, j, nswaps;
    for (i = lo; i < hi; i++) {
        nswaps = 0;
        for (j = hi; j > i; j--) {
            if (less(a[j], a[j-1])) {
                swap(a[j], a[j-1]);
                nswaps++;
            }
        }
        if (nswaps == 0) break;
    }
}

```

Insertion Sort

- Simple adaptive method:
 - Take first element and treat as sorted array
 - Take next element and insert into sorted part of array so that order is preserved
 - Above increases length of sorted path by one
 - Repeat until whole array is sorted

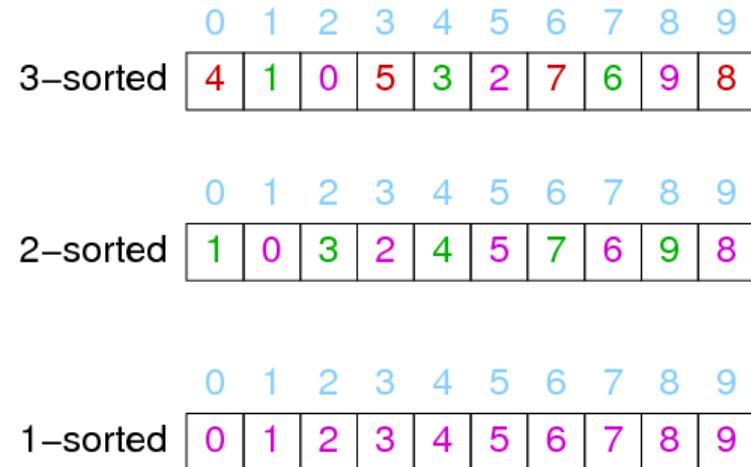


```
S O R T E X A M P L E
S O R T E X A M P L E
O S R T E X A M P L E
O R S T E X A M P L E
O R S T E X A M P L E
E O R S T X A M P L E
E O R S T X A M P L E
A E O R S T X M P L E
A E M O R S T X P L E
A E M O P R S T X L E
A E L M O P R S T X E
A E E L M O P R S T X
```

```
void insertionSort(int a[], int lo, int hi)
{
    int i, j, val;
    for (i = lo+1; i <= hi; i++) {
        val = a[i];
        for (j = i; j > lo; j--) {
            if (!less(val,a[j-1])) break;
            a[j] = a[j-1];
        }
        a[j] = val;
    }
}
```

ShellSort: Improving Insertion Sort

- Array is h-sorted if taking every h'th element yields a sorted array
- A h-sorted array is made up of n/h interleaved sorted arrays
- Shell sort: h-sort array for progressively smaller h, ending with 1-sorted



```
void shellSort(int a[], int lo, int hi)
{
    int hvals[8] = {701, 301, 132, 57, 23, 10, 4, 1};
    int g, h, start, i, j, val;
    for (g = 0; g < 8; g++) {
        h = hvals[g];
        start = lo + h;
        for (i = start+1; i <= hi; i++) {
            val = a[i];
            for (j = i; j >= start; j -= h) {
                if (!less(val, a[j-h])) break;
                a[j] = a[j-h];
            }
            a[j] = val;
        }
    }
}
```

- Complexity may be $O(n^{1.5})$ or $O(n^{4/3})$

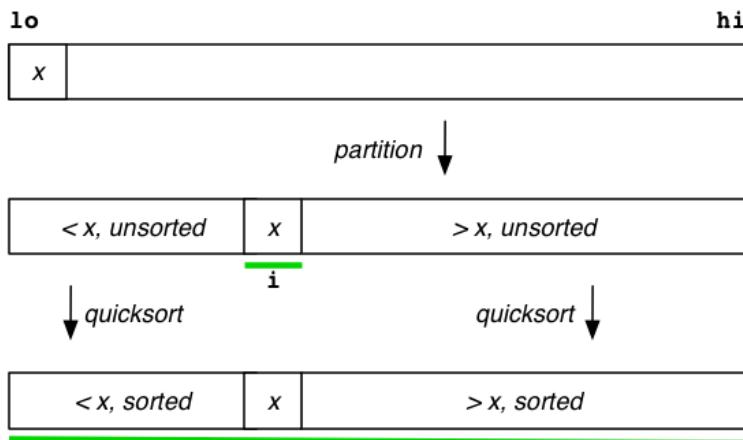
Sorting Comparison

	#compares			#swaps			#moves		
	min	avg	max	min	avg	max	min	avg	max
Selection sort	n^2	n^2	n^2	n	n	n	.	.	.
Bubble sort	n	n^2	n^2	0	n^2	n^2	.	.	.
Insertion sort	n	n^2	n^2	.	.	.	n	n^2	n^2
Shell sort	n	$n^{4/3}$	$n^{4/3}$.	.	.	1	$n^{4/3}$	$n^{4/3}$

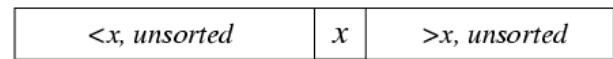
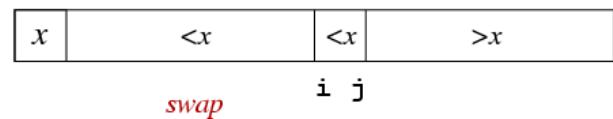
8.3: Quicksort

Quicksort

- Choose an item to be a pivot
- Rearrange (partition) the array so that:
 - All elements to left of pivot are smaller than pivot
 - All elements of pivot are greater than pivot
- Recursively sort each of the partitions



```
void quicksort(Item a[], int lo, int hi)
{
    int i; // index of pivot
    if (hi <= lo) return;
    i = partition(a, lo, hi);
    quicksort(a, lo, i-1);
    quicksort(a, i+1, hi);
}
```



```

int partition(Item a[], int lo, int hi)
{
    Item v = a[lo]; // pivot
    int i = lo+1, j = hi;
    for (;;) {
        while (less(a[i],v) && i < j) i++;
        while (less(v,a[j]) && j > i) j--;
        if (i == j) break;
        swap(a,i,j);
    }
    j = less(a[i],v) ? i : i-1;
    swap(a,lo,j);
    return j;
}

```

Quicksort Performance

- Best case: $O(n \log n)$
- Worst case: $O(n^2)$

Quicksort Improvements

- Choice of pivot can have significant effect
 - Always choosing largest/smallest -> worst case
 - Find intermediate value by median of tree

```

void medianOfThree(Item a[], int lo, int hi)
{
    int mid = (lo+hi)/2;
    if (less(a[mid],a[lo])) swap(a, lo, mid);
    if (less(a[hi],a[mid])) swap(a, mid, hi);
    if (less(a[mid],a[lo])) swap(a, lo, mid);
    // now, we have a[lo] < a[mid] < a[hi]
    // swap a[mid] to a[lo+1] to use as pivot
    swap(a, mid, lo+1);
}
void quicksort(Item a[], int lo, int hi)
{
    if (hi <= lo) return;
    medianOfThree(a, lo, hi);
    int i = partition(a, lo+1, hi-1);
    quicksort(a, lo, i-1);
    quicksort(a, i+1, hi);
}

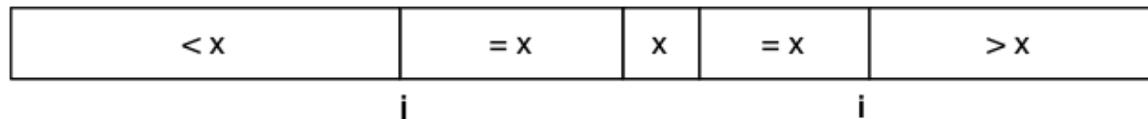
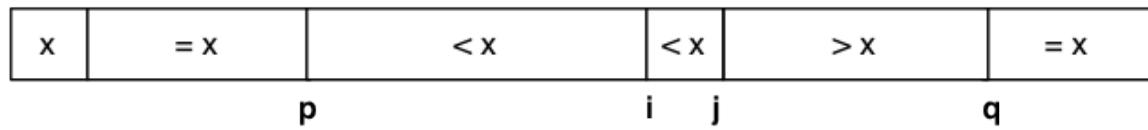
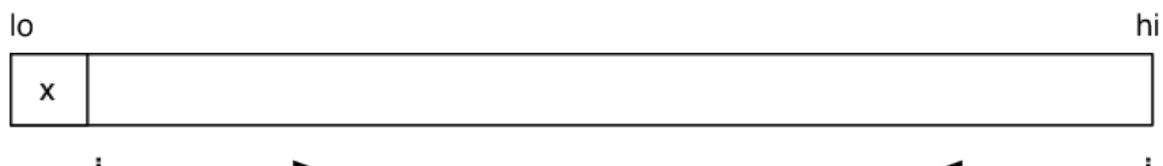
```

- Little benefit when partitioning when size < 5

- So use another sort, e.g. insertionSort

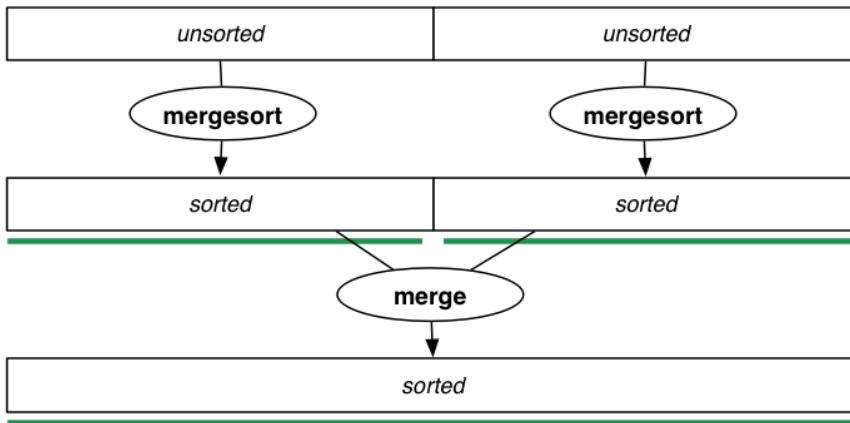
```
void quicksort(Item a[], int lo, int hi)
{
    if (hi-lo < Threshold) {
        insertionSort(a, lo, hi);
        return;
    }
    medianOfThree(a, lo, hi);
    int i = partition(a, lo+1, hi-1);
    quicksort(a, lo, i-1);
    quicksort(a, i+1, hi);
}
```

Three Way Partitioning



8.4: Mergesort

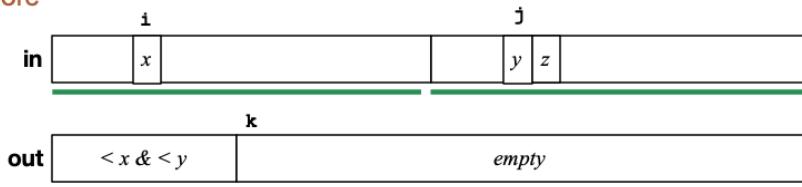
- Split the array into two equal-sized partitions
- Recursively sort each of the partitions
- Merge the two partitions into a new sorted array
- Copy back into original array



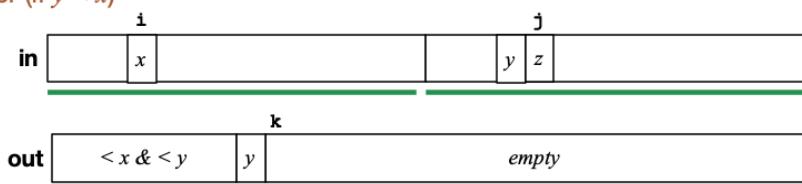
```

void mergesort(Item a[], int lo, int hi)
{
    int mid = (lo+hi)/2; // mid point
    if (hi <= lo) return;
    mergesort(a, lo, mid);
    mergesort(a, mid+1, hi);
    merge(a, lo, mid, hi);
}
  
```

Before



After (if $y < x$)



```

void merge(Item a[], int lo, int mid, int hi)
{
    int i, j, k, nitems = hi-lo+1;
    Item *tmp = malloc(nitems*sizeof(Item));

    i = lo; j = mid+1; k = 0;
    // scan both segments, copying to tmp
    while (i <= mid && j <= hi) {
        if (less(a[i],a[j]))
            tmp[k++] = a[i++];
        else
            tmp[k++] = a[j++];
    }
    // copy items from unfinished segment
    while (i <= mid) tmp[k++] = a[i++];
    while (j <= hi) tmp[k++] = a[j++];

    //copy tmp back to main array
    for (i = lo, k = 0; i <= hi; i++, k++)
        a[i] = tmp[k];
    free(tmp);
}

```

- Time complexity $O(N \log N)$

Bottom-up Mergesort

- Non-recursive merge sort does not require a stack
 - Partition boundaries can be computed iteratively
- Bottom-up mergesort:
 - On each pass, array contains sorted runs of length m
 - At start, treat as N sorted runs of length 1
 - 1st pass merges adjacent elements into runs of length 2
 - 2nd pass merges adjacent 2-runs into runs of length 4
 - Continue until a single sorted run of length N

```

#define min(A,B) ((A < B) ? A : B)

void mergesort(Item a[], int lo, int hi)
{
    int m;      // m = length of runs
    int len;    // end of 2nd run
    Item c[];  // array to merge into
    for (m = 1; m <= lo-hi; m = 2*m) {
        for (int i = lo; i <= hi-m; i += 2*m) {
            len = min(m, hi-(i+m)+1);
            merge(&a[i], m, &a[i+m], len, c);
            copy(&a[i], c, m+len);
        }
    }
}

```

```

// merge arrays a[] and b[] into c[]
// aN = size of a[], bN = size of b[]
void merge(Item a[], int aN, Item b[], int bN, Item c[])
{
    int i; // index into a[]
    int j; // index into b[]
    int k; // index into c[]
    for (i = j = k = 0; k < aN+bN; k++) {
        if (i == aN)
            c[k] = b[j++];
        else if (j == bN)
            c[k] = a[i++];
        else if (less(a[i],b[j]))
            c[k] = a[i++];
        else
            c[k] = b[j++];
    }
}

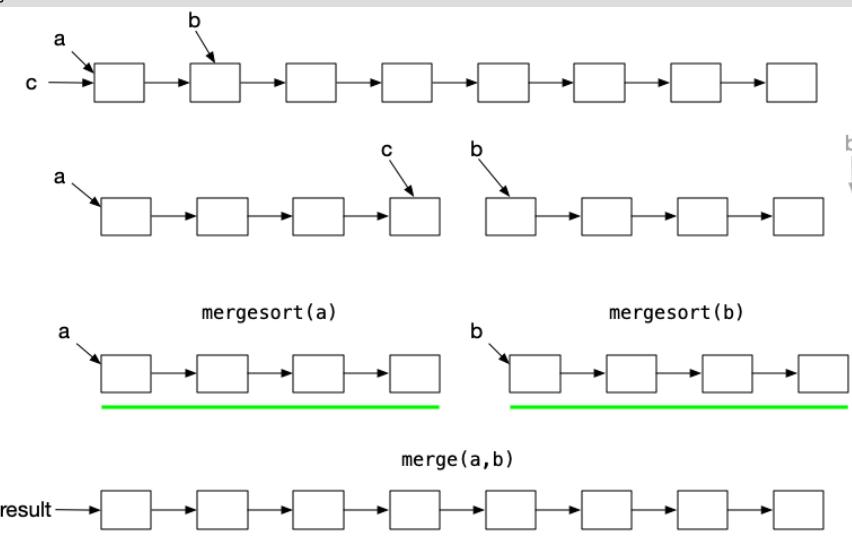
```

Mergesort and Linked Lists

```

List merge(List a, List b)
{
    List new = newList();
    while (a != NULL && b != NULL) {
        if (less(a->item, b->item))
            { new = ListAppend(new, a->item); a = a->next; }
        else
            { new = ListAppend(new, b->item); b = b->next; }
    }
    while (a != NULL)
        { new = ListAppend(new, a->item); a = a->next; }
    while (b != NULL)
        { new = ListAppend(new, b->item); b = b->next; }
    return new;
}

```



```
List mergesort(List c)
{
    List a, b;
    if (c == NULL || c->next == NULL) return c;
    a = c; b = c->next;
    while (b != NULL && b->next != NULL)
        { c = c->next; b = b->next->next; }
    b = c->next; c->next = NULL; // split list
    return merge(mergesort(a), mergesort(b));
}
```