



Lecture Notes: Syllabus, I/O

1. 13. 25

Standard I/O:

\$ program 2> error #redirect err
 \$ program > output 2>&1 #redirect both

Programming Challenges I/O:

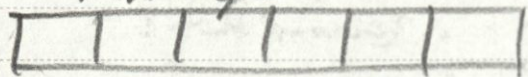
1. output is line oriented
2. input may have sentinels
3. input may specify # items to read
4. Normally do not have to handle
 bad input

* must correctly format output
exactly for grading

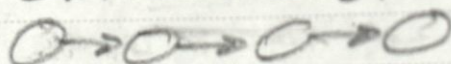
see exercise 01-4

Lecture Notes: Arrays, Lists

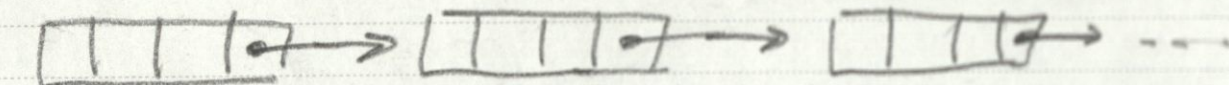
1.15.25

Arrays

- contiguous container of elements
- contiguous - next to each other in mem
- random access: $O(1)$
- poor insertions and deletions
- static amount of space

Linked List

- sequence of nodes that point to each other
- efficient insertions
- bad random access $O(n)$ lookup
- not contiguous
- more memory efficient?
- overhead for pointers

Deque :

random
access

locality

combines benefits of
both arrays and linked
lists

Whiteboarding: Tips

1. Ask questions to resolve ambiguity
What are the data types? assumptions?
2. Design an algorithm
What is the complexity? What trade-offs?
3. Write pseudocode first
with the goal of converting to real code later

Lecture Notes: Complexity

1.17.25

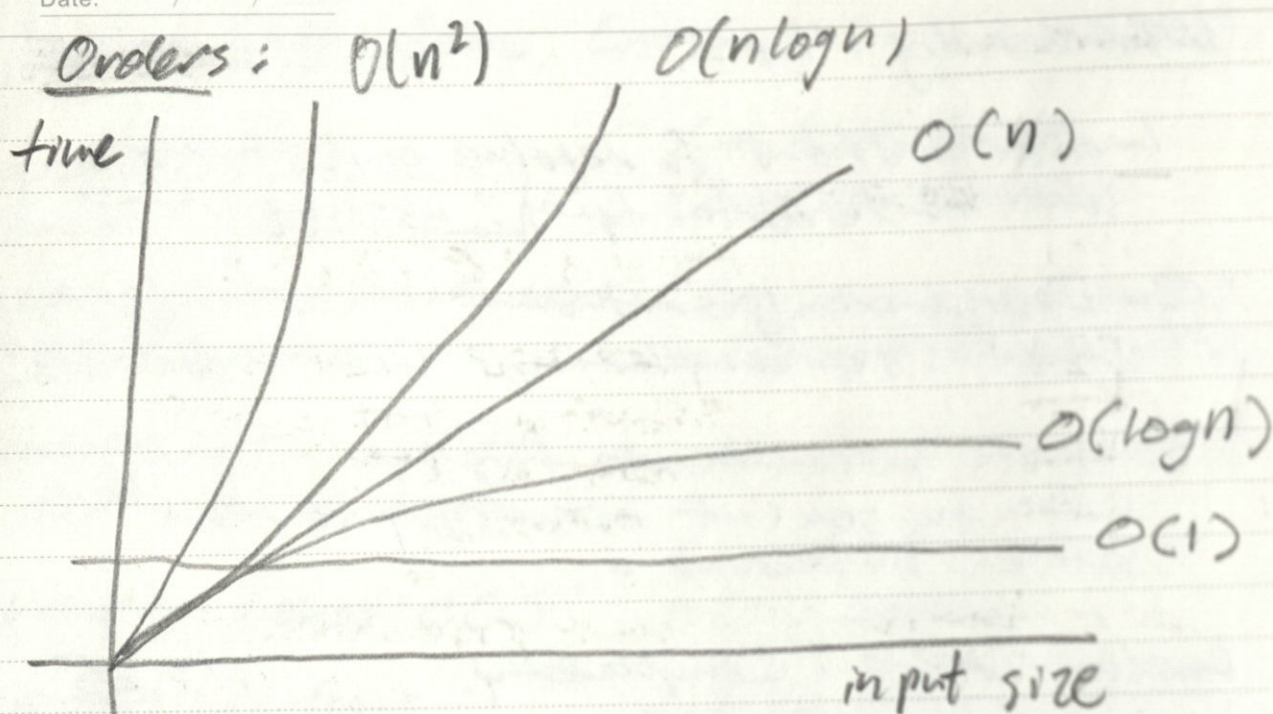
Big-O: describes the efficiency of an algorithm in terms of the amount of data it must process

time complexity: # of operations an algorithm performs relative to input size

space complexity: amount of additional memory an algorithm requires relative to amount of input

Def. a computation is $O(f(n))$ time if for large n , the computation time is at most $kf(n)$ for some k

Date. / /



Coding Styles:

Linters: tools to help with coding style

Lecture Notes:

1.22.25

Stacks:

45817

7
1
8
5
4

LIFO

"last in first out"
used in DFS
backtracking
function calls

Queues:

4	5	8	1	7
---	---	---	---	---

FIFO

"first in first out"

used in BFS

Scheduling
threading
unix pipeline

Implementation:

stack = []

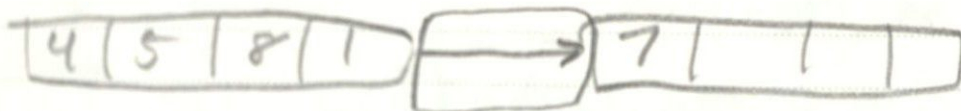
- pop()
- append()

queue = []

- popleft()
- append()

Deque:

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import collections

q = collections.deque()

- amortized constant time insertion and deletion from front and back
- constant time random access
- some cache locality

Lecture Notes:

1.24.25

Testing: Overview

- determines whether code is correct.
- like proofs for math

functional
• checking
correctness

performance
• tests time
or resource
constraints

unit
• check
individual
components

penetration, integration, load

Lecture Notes:

1.27.25

linear search: go through list
one-by-one in $O(N)$ time

binary search: split search space
in half each time in $O(\log N)$ time

- need a way of determining
which half of search space
our target is

algorithm:

1. compute mid point
2. check if midpoint is target
3. target < midpoint: go left
4. target > midpoint: go right
5. repeat until found or boundary

Python standard library:Linear Search:

```
if target in container:
    print("Found it")
```

Binary Search:

```
import bisect
index = bisect.bisect_left(data, target)
if index < len(data) and
    data[index] == target:
    print("Found it")
```

Lecture Notes:

1.29.25

Benefits of Sorting:

1. Min/Max values
2. Median
3. Frequency / duplicates

 $O(n \log n)$ sorting:

```
V = sorted(V) // returns a copy
V.sort() // in-place
```

↪ uses Timsort $\leftarrow$ Mergesort

C++: `sort(V.begin(), V.end())`
 ↪ introsort $\leftarrow$ quicksort

Reversing Directions:C++

```
sort(v.begin(),
      v.end())
```

Python

```
sorted(v, reverse=True)
reversed(sorted(v))
v.sort(reverse=True)
```

Stable Sorting: maintaining relative order of equivalent elements

4, 6, 6, 15 $\rightarrow$ 1, 4, 5, 6, 6, 15

timsort, insertion sort, and merge sort are all stable

Lecture Notes:

2.3.25

set: collection of unique items

map: collection of key: value pairs

can be used to check membership, counting, memoization

built with hashing

Set:

insert (value)
 remove (value)
 search (value)
 membership
 Unix Permissions

Map:

insert (key, value)
 remove (key)
 search (key)
 Table of k/v pairs
 File System
 Javascript

Sets and Maps: Ordered vs Unordered

- Sets are unordered: means no random access
- Maps are ordered: maintain order you insert them

Ordered:

- each element needs a comparison operator defined
- implemented with a BST

Unordered:

- each element needs to be immutable
- typically implemented with a hash table

Sets and Maps: Python

```
my_set = set()
```

```
my_set = {}
```

```
my_set.add(1)
```

```
my_dict = dict()
```

```
my_dict = {}
```

```
my_dict[key] = value
```

Implementation:Binary Search Tree:Time: $O(\log n)$ Space: $O(k \cdot n)$ Hash Table:Time: $O(1)$ Space: $O(n)$ $O(k(m+n))$ Techniques:

Counting: keeping track of frequencies

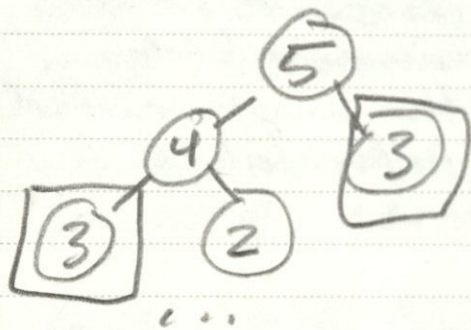
Memorization: Caching subproblem results

• Fibonacci

History: keeping track of info over time

Ex. Fibonacci sequence

for problems with overlapping subproblems,
 can store result of previous subproblems
 for future re-use



2.5.25

Lecture Notes:Sets and Maps: Techniques

1. Counting: Using a set or map to count frequencies
2. Memorization: using a map to code subproblems
3. History: using a set or map to track seen information

see ex is_palindrome.py
- counting

see ex memo-fib.py
- memorization

Lecture Notes:

2.10.25

Complete Search: sometimes the best approach is to try every possibility i.e. brute force

- guarantees it will find the right answer
- very inefficient
- first pass solution

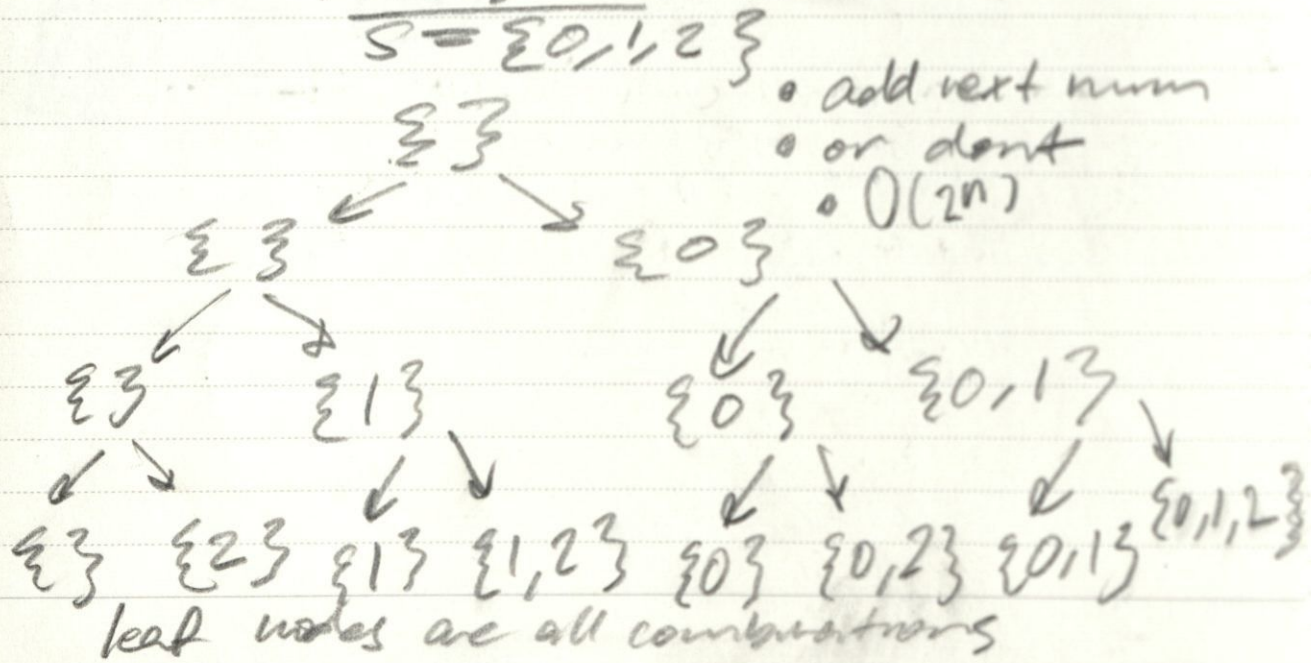
Combinations: $O(2^n)$ Permutations: $O(n!)$

Subsets:

- itertools.combinations
- bitmask
- recursive solution

Recursive: Combinations

$S = \{0, 1, 2\}$



Lecture Notes:

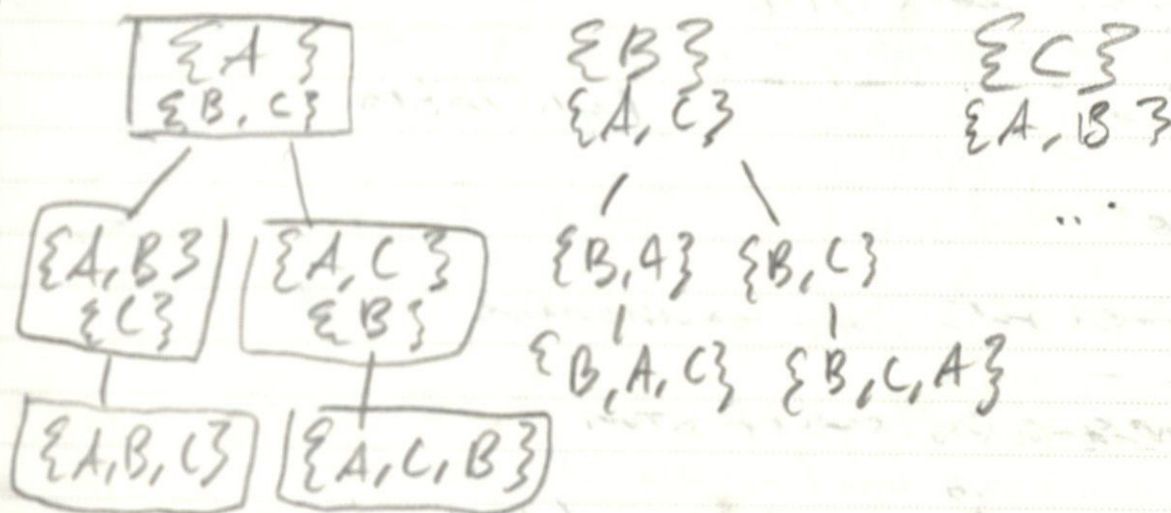
2.12.25

Complete Search:

Permutations: order matters

Combinations: order does not matter

itertools . permutations (same as combinations)

Permutations Diagram: $S = \{A, B, C\}$ 

note, permutations are all the same length as the initial set.

ie, no empty set etc.

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Backtracking:

- start w/ empty sol
- add 1 item and see if sol violates constraints
- if it does not, recursively add another item
- otherwise back-up to previous stage and try alternatives

Lecture Notes:

2.17.25

Review: set ADT

Set Implementation:

BST:

Time: $O(\log N)$

Space: $O(N)$

Hash Table:

Time: $O(1)$

Space: $O(N)$

• ordered set

• unordered set

Bitsets: instead of storing elements, we can store whether an element is in a set by marking a bit

574 $\rightarrow$

7	6	5	4	3	2	1	0	index
1	0	1	1	0	0	0	0	bit

note. not counting, just checking membership

- use ints as sets
- use logical bit operators to set and clear individual bits

Bitsets: Subsets

To generate subsets using bitsets,
Start with $\emptyset$, and then increment by 1

Each bit represents membership
of element in subset

<u>Bitset</u>	<u>A</u>	<u>B</u>	<u>C</u>	<u>Subset</u>
0	0	0	0	$\{\}$
1	0	0	1	$\{C\}$
2	0	1	0	$\{B\}$
3	0	1	1	$\{B, C\}$
4	1	0	0	$\{A\}$
5	1	0	1	$\{A, C\}$
6	1	1	0	$\{A, B\}$
7	1	1	1	$\{A, B, C\}$

Lecture Notes: Greedy Algs

2.19.25

A greedy algorithm constructs a solution by always taking the best local choice (choice at the moment)

directly constructs final solution and never takes back a choice

Usually very efficient

Locally optimized solutions are globally optimal

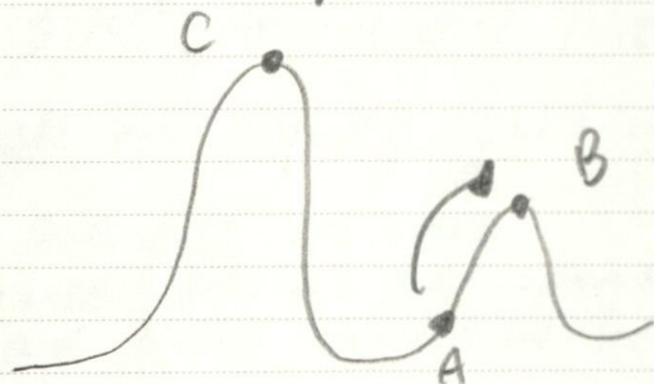
Ex. Making Change

$$17¢ = 10 + 5 + 1 + 1$$

$$36¢ = 25 + 10 + 1$$

$$64¢ = 25 + 25 + 10 + 1 + 1 + 1 + 1$$

Limitations: not guaranteed to find the optimal solution



Lecture Notes :

2.24.25

Dynamic Programming combines the correctness of complete search and the efficiency of greedy algorithms

- cache/memorize results of subproblems in memory to prevent repeated computations
- Efficiently solves a set of search and optimization problems

Caching : Memory Computations

- identify overlapping subproblems
- store results in lookup table

Steps :

1. identify recursive structure
2. memorize recursive step in cache
3. return memorized solution

```

graph TD
    7 --> 2
    7 --> 6
    2 --> 7
    2 --> 9
    6 --> 0
    6 --> 1
    6 --> 2
  
```

5

Sackward ✓



Lecture Notes:

Making Change: (1, 3, 4)

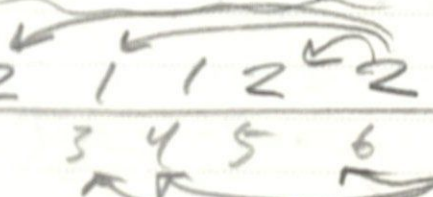
Greedy Alg: $6 \rightarrow (4, 1, 1)$

Correct Sol: $6 \rightarrow (3, 3)$

i.e. greedy failed

Use DP:

Bottom Up (Backward):



coins	0	1	2	1	1	2	2	2	
AMT	0	1	2	3	4	5	6	7	8

use $\min(i-1, i-3, i-4) + 1$

$= \min(\min(5, 3, 2), 1, 1) + 1$

; note. initialize coins to 0!

Bottom Up (Forward):



Coins	0	1	2	1	1	2			
AMT	0	1	2	3	4	5	6	7	8

$\text{Table}[i + \text{coin}] = \min(\text{Table}[i + \text{coin}], \text{Table}[i] + 1)$

note initialize coins to ∞ !

Date. / /

Squirrel Hunting:

PAD w/ 03 for boards

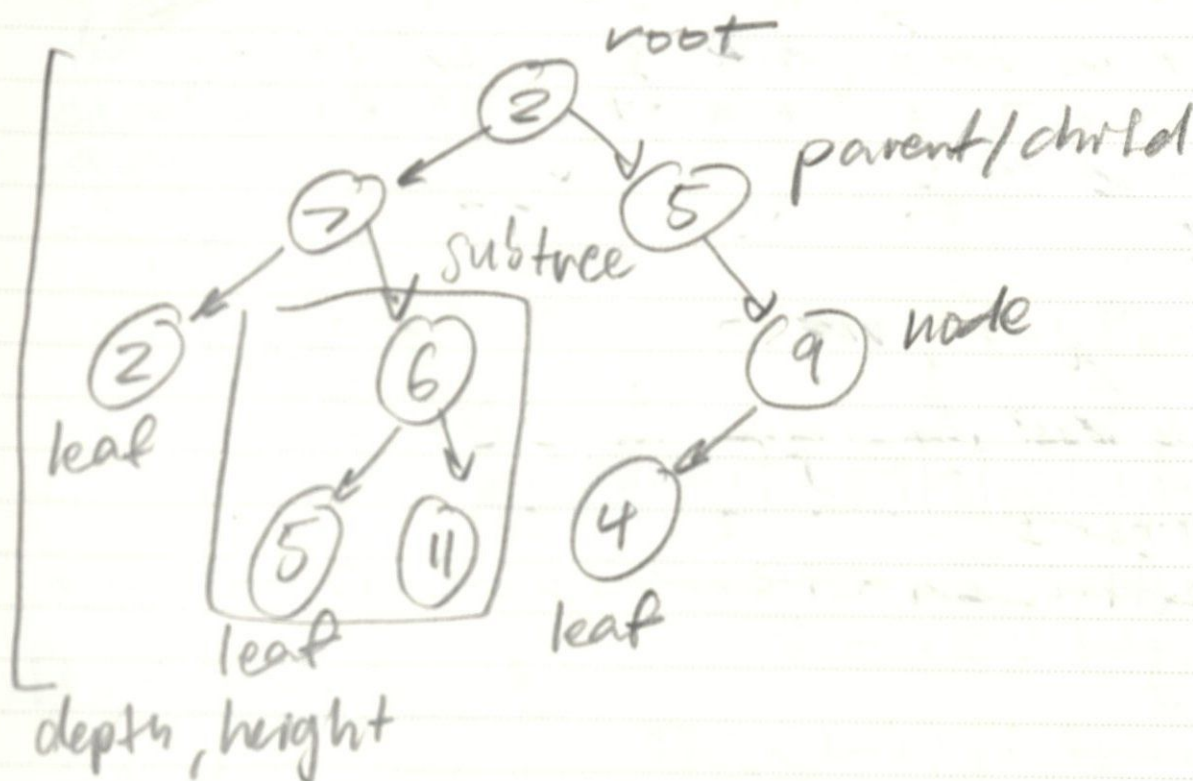
1	5	2	3	6	0	0	0	0	0	0
4	3	2	1	2	0	5	9	11	12	19
3	8	4	2	1	0	8	17	21	23	24
0	5	2	3	4	0	8	22	24	27	31
3	1	4	2	1	0	11	23	28	30	32

$\text{Max}(\text{left}, \text{up}) + \text{current_cell}$

How did you get to 32?

Follow squirrel that correlates w/
accum_cell - curr_cell

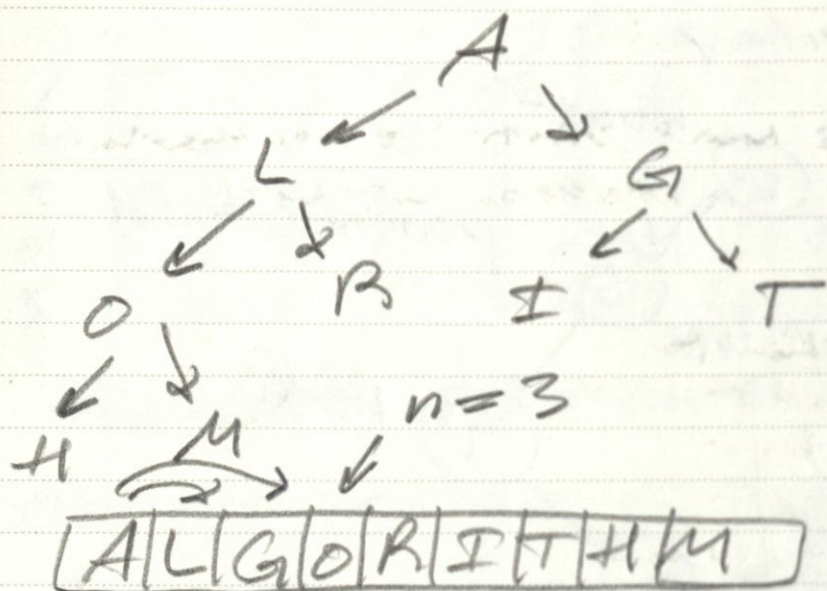
3.17.25

Lecture Notes:Trees: Vocabulary

full: every node has 0 or 2 children
 complete: every level has max children
 except last (left $\rightarrow$ right)

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Representations:



BFS
Rep

Left: $2n + 1$

Right: $2n + 2$

Parent: $\lfloor (i-1)/2 \rfloor$

Binary Heap: binary tree where
a parent is larger than all of
its children or smaller

→ used in priority queues

must also be complete

BST: for every parent node

- 1) left subtree must be less than parent
- 2) right subtree must be greater

* must be balanced *

Lecture Notes

3.19.25

Tree Traversals :

traversing a tree means visiting each node exactly once in a deterministic order

- Breadth-first
- Depth-first
 - pre-order
 - in-order
 - post-order

* implementation in slides

Divide and Conquer :

top-down technique that consists of dividing the problem into smaller subproblems and combining the partial solutions

* exs of this in slides

Lecture Notes :

3/24/25

Graphs :

collection of vertices and edges

Representation :

- Adjacency List (more used)
- Adjacency Matrix

Traversals :

- BFS
- DFS
- Dijkstra's - computes shortest path

Vocabulary :

edge: connection between vertices

directed: has a direction

weighted: edge has an associated cost

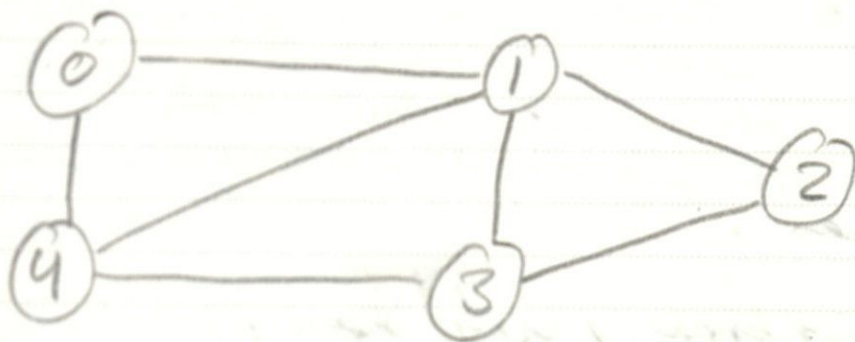
vertices: circles with labels

cyclic vs acyclic

connected: can reach any node from any node (vs disconnected)

simple: only one edge b/w 2 nodes
no self loops

Representation:



Adjacency List:

graph = {
 0: [1, 4]
 1: [0, 2, 3, 4]
 2: [1, 3]
 3: [1, 2, 4]
 4: [0, 1, 3]
}

Adjacency Matrix:

graph = [[0, 1, 0, 0, 1],
 [1, 0, 1, 1, 1],
 [0, 1, 0, 1, 0],
 [0, 1, 1, 0, 1],
 [1, 1, 0, 1, 0]]

Lecture Notes:

3.28.25

Dijkstra's: single source shortest path algorithm

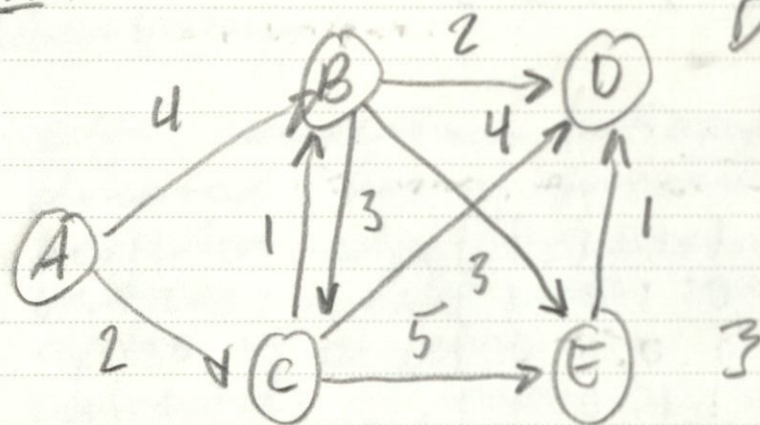
determines the shortest path between two vertices

using our traversal alg., we repeatedly remove a vertex from the frontier until we find the shortest path

to choose next vertex, select the cheapest via minheap

• based on total distance

Ex.



Distances = ∞

A: 0
B: ~~∞~~ 4 3
C: ~~∞~~ 2
D: ~~∞~~ 6 5
E: ~~∞~~ 7 6

Frontier = { B, C, D, E }
Visited = { A, C, B, D, E }

Pseudo-Code:

Dijkstra(g, v):
 frontier = PQ
 visited = $\{\}$

while frontier:
 $v = \text{frontier.pop}()$

if v in visited:
 continue

for u in $g[v]$:
 frontier.append(u)

for frontier we use a priority Q to store vertex (cost, name, prev) and order by cost

for visited, we use a map to record either

- node $\rightarrow$ prev
- node $\rightarrow$ t-cost

we can terminate loop when we reach target vertex

Lecture Notes:

3.31.25

Spanning Tree: acyclic graph that has
a path b/w any two vertices

Minimum Spanning Tree: spanning tree
with the least cost

Prim's Algorithm: local expansion

Kruskal's Algorithm: global expansion

MST:

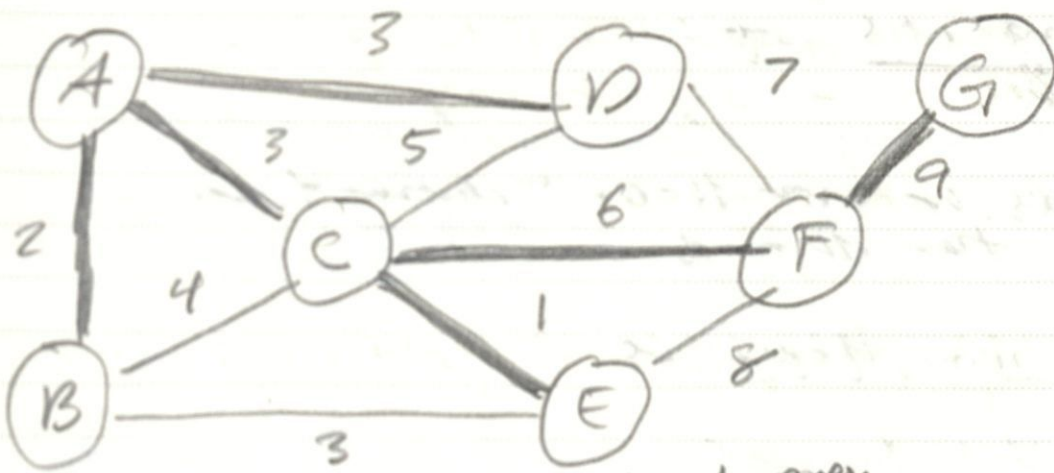
A spanning tree of a graph $G=(V, E)$
is a subset of edges forming a tree
connecting all vertices

- Connected
- Undirected

MST is a spanning tree with the least
total cost

• sum of edge weights is a minimum

Ex.



~~(0, A, A)~~
 cost ↑ Target ↑ Prev

~~(2, B, A)~~

~~(3, C, A)~~

(3, D, A)

(6, C, B)

(5, E, B)

(8, D, C)

(4, E, C)

target prev

A: A

B: A

C: A

D: A

E: C

F: C

G: F

dijkstra numbers

want isolated cost not total cost

Prim's MST:

1) Use dijkstra's and pick any node as start node

2) Only add edge's cost to PQ
 don't accumulate

Lecture Notes

4.2.25

Topological Sort:

models a large collection of actions, objects, or events that depend on each other through a DAG

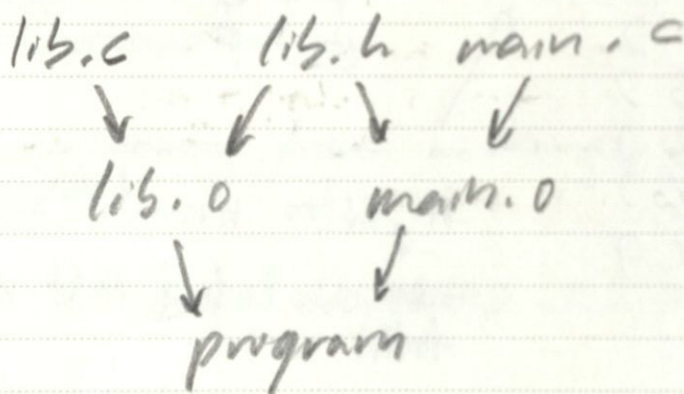
Motivation: Makefiles

target: sources

lib.o: lib.c lib.h

main.o: main.c lib.h

program: lib.o main.o



frontier: queue, any data structure

visited: set

degrees: dictionary

Ex. frontier:
 lib.c, lib.h, main.c

visited:
 lib.c, lib.h

}

degrees:
 lib.c: 0
 lib.h: 0
 main.c: 0
 lib.o: 2 + 0
 main.o: 2 + 1
 program: 2

}

degree = # incoming edges

1. init frontier w/ all nodes of degree 0
2. pop frontier add to visited
3. update neighbor's degrees
4. add node from degrees to frontier if degree is now 0

Lecture Notes:

4.7.25

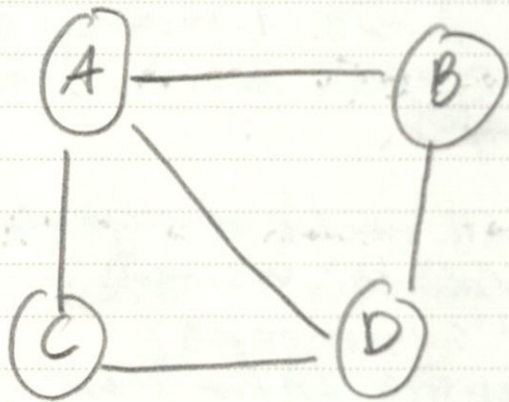
Eulerian Path: goes through every edge at least once

general bridge problem

Hamiltonian Path: goes through every vertex at least once

traveling salesman problem

Ex.



Eulerian: $A \rightarrow B \rightarrow D \rightarrow C \rightarrow A \rightarrow D$

Hamiltonian: $A \rightarrow C \rightarrow D \rightarrow B$

* there can be multiple paths

Circuit. Path that starts and ends on the same node

Rules.

Eulerian:

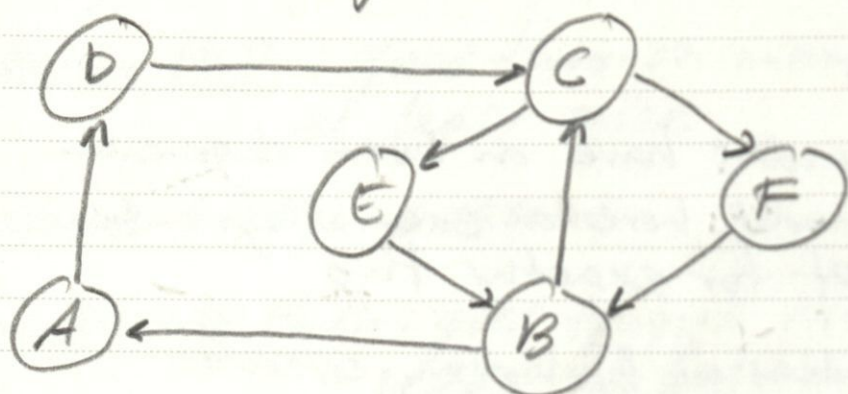
- 1) All nodes have an even degree
- 2) All nodes have an even degree except for exactly two

(1) guarantees an Eulerian circuit

Hamiltonian: NP-hard

- 1) connected
- 2) for each pair of vertices either $\text{degree}(u) + \text{degree}(v) \geq n$ and u and v are adjacent
- 3) each vertex has $\text{degree} \geq n/2$

Hierholzer's Algorithm:



- 1) build a circuit
- 2) mark all edges used as used
- 3) look for extra edges connected to circuit
- 4) build a subcircuit with extra edges
- 5) add subcircuit to circuit
- 6) repeat until all edges visited

Lecture Notes:

4.9.25

Flows.

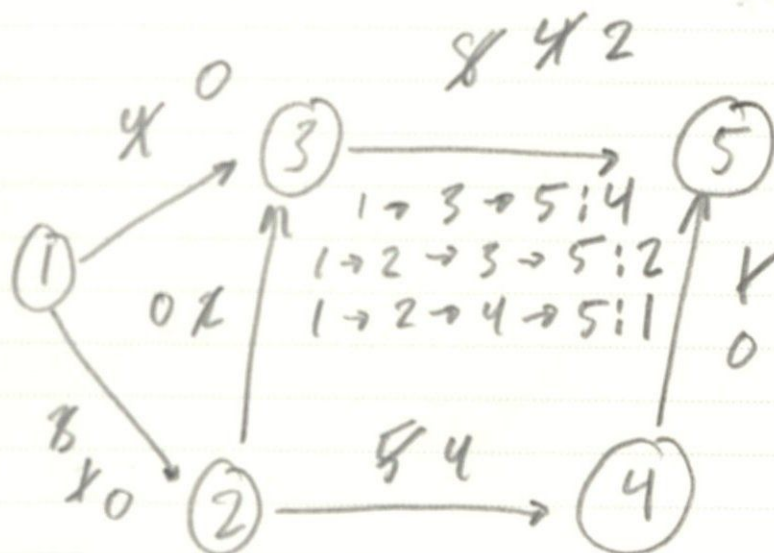
Source: vertex with only outgoing edges

Sink: vertex with only incoming edges

Max flow: maximum amount of "stuff" that can flow from source node to sink node using weighted edges

Flow Overview:

- 1) BFS Traversal $\rightarrow$ Paths
- 2) Find smallest edge in paths
- 3) Subtract weight from edges in paths
- 4) Don't take 0 edges

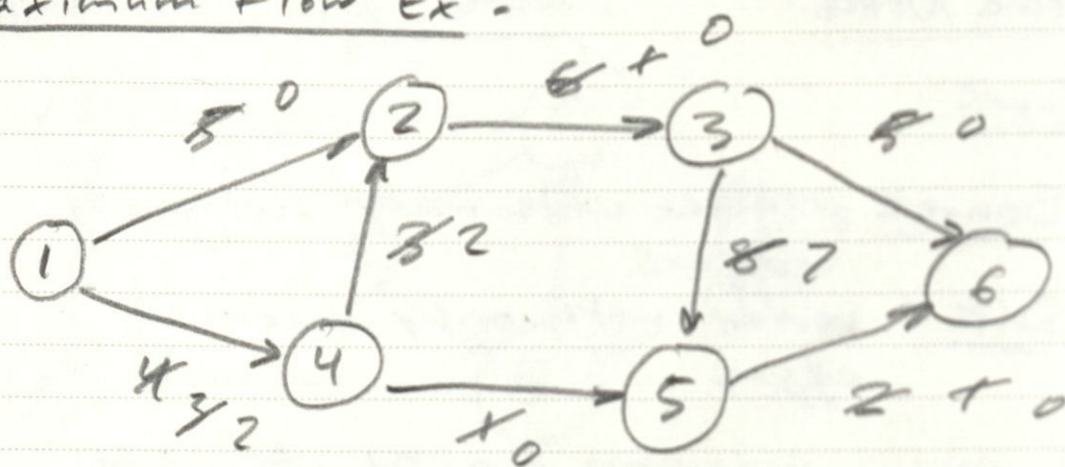


Max Flow: 7

Min Cut: 7

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Maximum Flow Ex :



Max Flow = 7

1) BFS : 1 2 4 3 5 6

paths: $1 \rightarrow 2 \rightarrow 3 \rightarrow 6 : 5$

$1 \rightarrow 4 \rightarrow 5 \rightarrow 6 : 1$

$1 \rightarrow 4 \rightarrow 2 \rightarrow 3 \rightarrow 5 \rightarrow 6 : 1$

$5 + 1 + 1 = 7 = \text{Max Flow}$

Ford - Fulkerson: algorithm we just
run using DFS

Lecture Notes:

Prime #: only divisible by 1 and itself

Check:

- Brute force
- Sieve of Eratosthenes

GCD: largest positive integer that divides both of the integers

computed w/ Euclidean Alg