

# Analysis of Algorithms

## CS 477/677

---

Instructor: Monica Nicolescu

Lecture 27

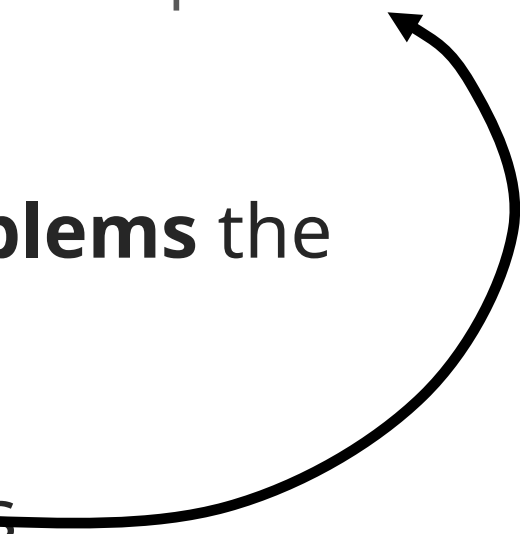
# Final Exam

---

- Tuesday, May 14, 12:45-2:45pm, in class
- Exam structure:
  - TRUE/FALSE questions
  - short questions on the topics discussed in class
  - homework-like problems
- Questions will be from the entire semester, with emphasis on material after midterms

# General Advice for Study

---

- **Understand** how the algorithms are working
    - Work through the examples we did in class
    - “Narrate” for yourselves the main steps of the algorithms in a few sentences
  - Know **when** or **for what problems** the algorithms are applicable
  - **Do not memorize** algorithms
- 

# Topics Covered after Midterms

---

- Dynamic programming (recurrence only)
- Greedy algorithms (greedy choice + proof)
- Graph algorithms
  - Search (BFS, DFS)
  - Topological Sort
  - Minimum spanning trees
  - Shortest paths (single source)
- NP-completeness (general questions only)
- Definitions
  - Do not say "An X is **when** [story here]"
  - Do say: "An X is **a/an** [noun phrase here]"
- Material prior to midterms
  - True/false, short questions/mini-problems, extra-credit/grad

# Dynamic Programming

---

- Used for **optimization problems**
  - A set of choices must be made to get an optimal solution
  - Find a solution with the optimal value (minimum or maximum)
- Applicability:
  - Subproblems are not independent, i.e., subproblems share subsubproblems
  - A divide-and-conquer approach would repeatedly solve the common subproblems
  - Dynamic programming solves every subproblem just once and stores the answer in a table

# Elements of Dynamic Programming

---

- Optimal Substructure
  - An optimal solution to a problem contains within it an optimal solution to subproblems
  - Optimal solution to the entire problem is build in a bottom-up manner from optimal solutions to subproblems
- Overlapping Subproblems
  - If a recursive algorithm revisits the same subproblems again and again  $\Rightarrow$  the problem has overlapping subproblems

# Exercise

Give an  $O(n^2)$  algorithm to find the longest monotonically increasing sequence in a sequence of  $n$  numbers.

- Take an example: (5, 2, 8, 7, 3, 1, 6, 4)
- Define:  $s_i$  = the length of the longest sequence ending with the  $i$ -th character

$s_0 = 0$

5	2	8	7	3	1	6	4
1	1	2	2	2	1	3	3

$$s_i = \max_{0 < j < i, \text{seq}[j] < \text{seq}[i]} \{s_j\} + 1$$

$s_i$  = 1 more than the greatest value for a previous number that is smaller than  $\text{seq}[i]$

# Greedy Algorithms

---

- Similar to dynamic programming, but simpler approach
  - Also used for optimization problems
- **Idea:**
  - When we have a choice to make, make the one that looks best right now in hope of getting a globally optimal solution
- **Problems:**
  - Greedy algorithms don't always yield an optimal solution
- When the problem has certain general characteristics, greedy algorithms give optimal solutions



# Correctness of Greedy Algorithms

---

- Greedy Choice Property
  - A globally optimal solution can be arrived at by making a locally optimal (greedy) choice
- Optimal Substructure Property
  - Optimal solution to subproblem + greedy choice  
⇒ optimal solution for the original problem

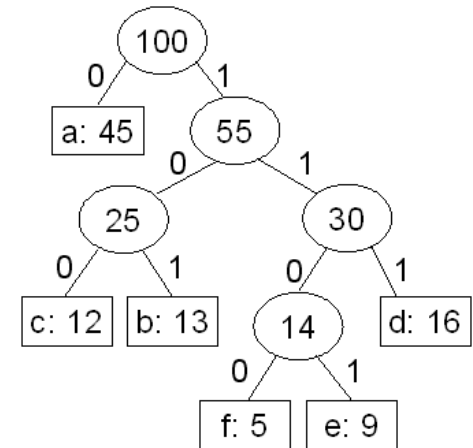
# Dynamic Programming vs. Greedy Algorithms

---

- Dynamic programming
  - We make a choice at each step
  - The choice depends on solutions to subproblems
  - Bottom up solution, from smaller to larger subproblems
- Greedy algorithm
  - Make the greedy choice and THEN
  - Solve the subproblem arising after the choice is made
  - The choice we make may depend on previous choices, but not on solutions to subproblems
  - Top down solution, problems decrease in size

# Huffman Codes

- Technique for data compression
- **Idea:**
  - Represent each character as a binary string
  - **Variable length code:** assign short codewords to frequent characters
  - Represent the codes as full binary trees whose leaves are the given characters
- Constructing an optimal prefix code (Huffmann code)
  - Start with a set of  $|C|$  leaves
  - At each step, merge the two least frequent objects: the frequency of the new node = sum of two frequencies



$$B(T) = \sum_{c \in C} f(c) d_T(c)$$

# Adj. List - Adj. Matrix Comparison

---

Graph representation: adjacency list, adjacency matrix

Comparison	Better
Faster to test if $(x, y)$ exists?	matrices
Faster to find vertex degree?	lists
Less memory on sparse graphs?	lists $(m+n)$ vs. $n^2$
Faster to traverse the graph?	lists $(m+n)$ vs. $n^2$

Adjacency list representation is better for most applications

# BFS vs. DFS

## BFS

- **Input:**
  - A graph  $G = (V, E)$  (directed or undirected)
  - A **source** vertex  $s \in V$
- **Idea:**
  - Explore the edges of  $G$  to “discover” every vertex reachable from  $s$ , **taking the ones closest to  $s$  first**
- **Output:**
  - $d[v]$  = distance (smallest # of edges, or shortest path) from  $s$  to  $v$ , for all  $v \in V$
  - BFS tree

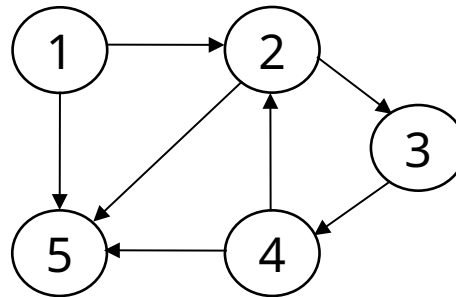
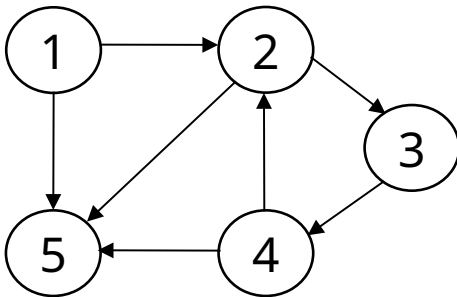
## DFS

- **Input:**
  - A Graph  $G = (V, E)$  (directed or undirected)
  - No source vertex given!
- **Idea:**
  - Explore the edges of  $G$  to “discover” every vertex in  $V$  **starting at the most recently visited node**
  - Search may be repeated from multiple sources
- **Output:**
  - 2 **timestamps** on each vertex:  $d[v]$ ,  $f[v]$
  - DFS forest

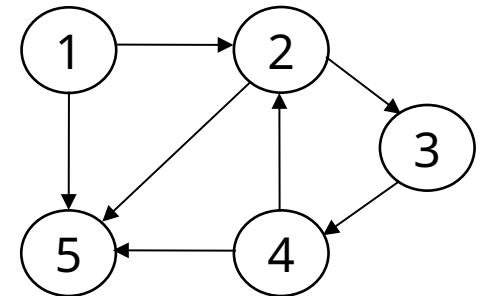
# Example - Question

- $G = (V, E)$ . True or false?

All DFS forests (for traversal starting at different vertices) will have the same number of trees.



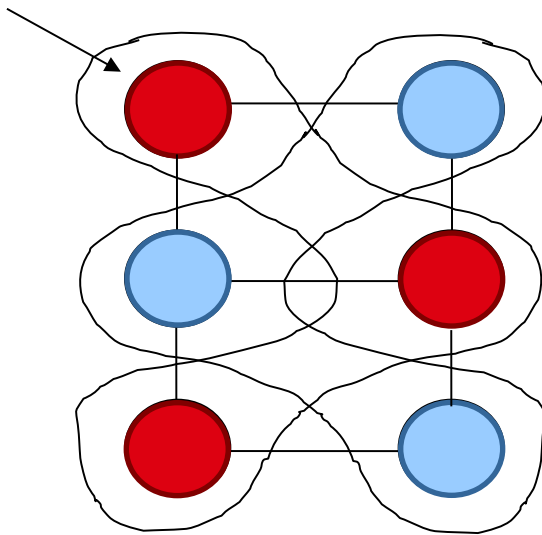
$\{1, 2, 5\}, \{3, 4\}$



$\{2, 5\}, \{1\}, \{4\}, \{3\}$

# Exercise

- Show how you can detect whether a graph is bipartite.
- Idea:
  - Nodes in the same set will never be adjacent to each other



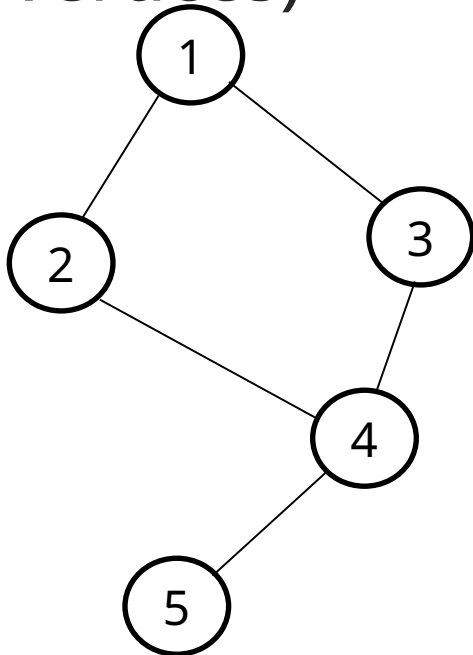
While doing a BFS, color nodes in two colors (red, blue), always alternating from a parent to a child

The graph is bipartite if all adjacent nodes have different colors

# Exercise

---

- Give an  $O(n)$  algorithm to test if an **undirected graph** has a cycle ( $n = \#$  of vertices)



- **Idea:**

- The graph has a cycle if it has more than  $n - 1$  edges
- Checking this takes  $O(n)$



# Sample Problem

---

- Give an  $\Theta(n+m)$  algorithm that tests whether an **undirected graph** is connected. The graph is given in adjacency list representation and has  $n$  vertices and  $m$  edges.
  - Run DFS using any vertex as a start vertex and then check whether all nodes were visited.
  - The graph is connected if and only if all nodes were visited, that is, if and only if all nodes are reachable from the start vertex.

# Properties of DFS

---

- Descendant – ancestor relationships
    - Relations on discovery – finish times for descendant, parent (ancestor) nodes
  - Parenthesis Theorem
  - White Path Theorem
- Review these properties!

# Topological Sort

---

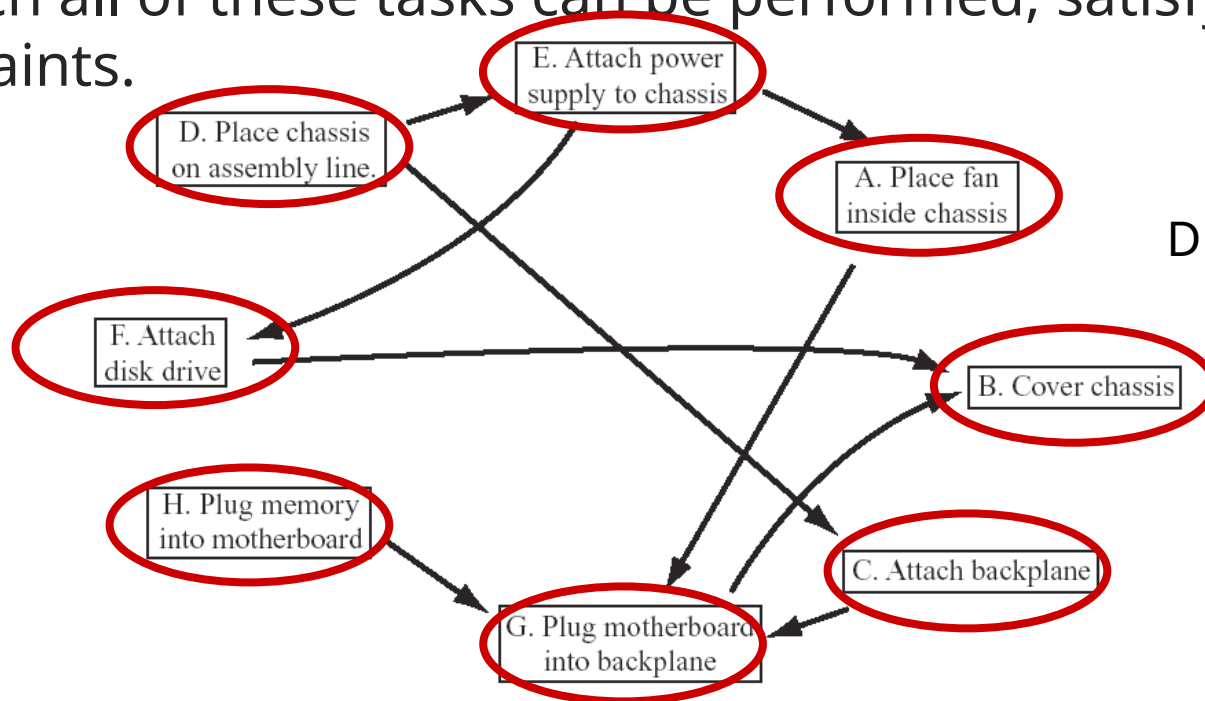
**Topological sort** of a directed acyclic graph  $G = (V, E)$ : a linear order of vertices such that if there exists an edge  $(u, v)$ , then  $u$  appears before  $v$  in the ordering.

TOPOLOGICAL-SORT( $V, E$ )

1. Call DFS( $V, E$ ) to compute finishing times  $f[v]$  for each vertex  $v$
2. When each vertex is finished, insert it onto the front of a linked list
3. Return the linked list of vertices

# Sample Problem

- In the following graph, boxes represent tasks that must be performed in the assembly of a computer, and arrows represent constraints that one task must be performed before another (for instance, the disk drive must be attached before the chassis can be covered). Write down a sequence in which all of these tasks can be performed, satisfying all constraints.



D E A F H C G B

# Minimum Spanning Trees

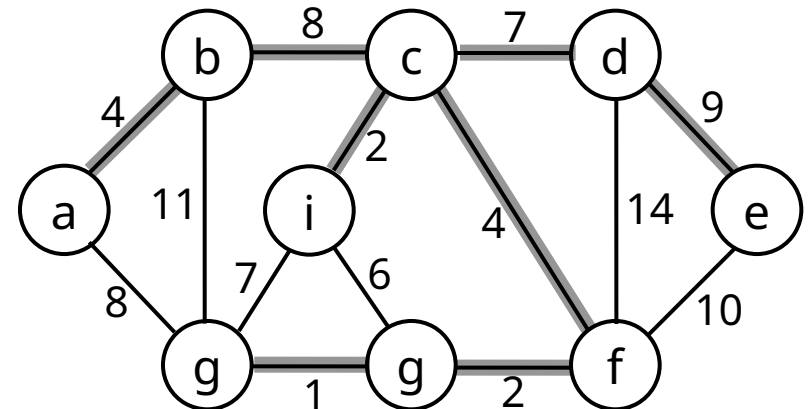
---

Given:

- A connected, undirected, weighted graph  $G = (V, E)$

A minimum spanning tree:

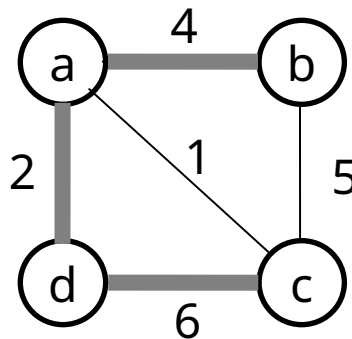
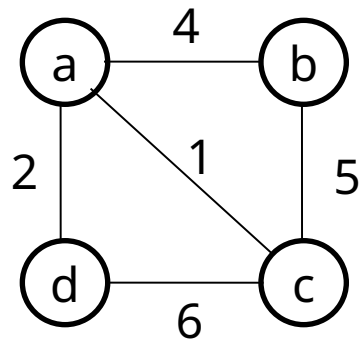
1.  $T$  connects all vertices
2.  $w(T) = \sum_{(u,v) \in T} w(u, v)$  is minimized



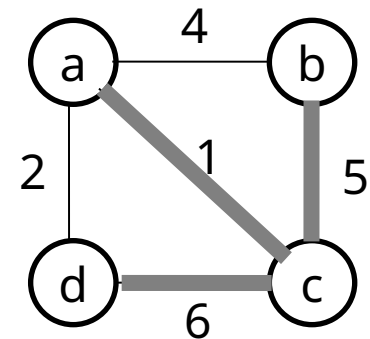
# Exercise

- True or **False?**

- If all the weights of a connected, weighted graph are distinct, then distinct spanning trees of  $G$  have distinct weights



Cost 12



Cost 12

# Minimum Spanning Trees

---

- Kruskal's algorithm
  - Start with each vertex being its own component
  - Repeatedly merge two components into one by choosing the **light edge** (minimum cost) that connects them
  - During the algorithm the MST is a forest of trees
- Prim's algorithm
  - The edges added to the MST always form a single tree
  - Repeatedly add light edges that connect vertices from outside the MST to vertices in the MST

# Variants of Shortest Paths

---

- **Single-source shortest path**
  - $G = (V, E) \Rightarrow$  find a shortest path from a given source vertex  $s$  to each vertex  $v \in V$
  - Bellman-Ford algorithm
  - Single-source shortest paths in acyclic graphs
  - Dijkstra's algorithm
- Know when the algorithms are applicable and how they work



# Relaxation

- **Relaxing** an edge  $(u, v)$  = testing whether we can improve the shortest path to  $v$  found so far by going through  $u$

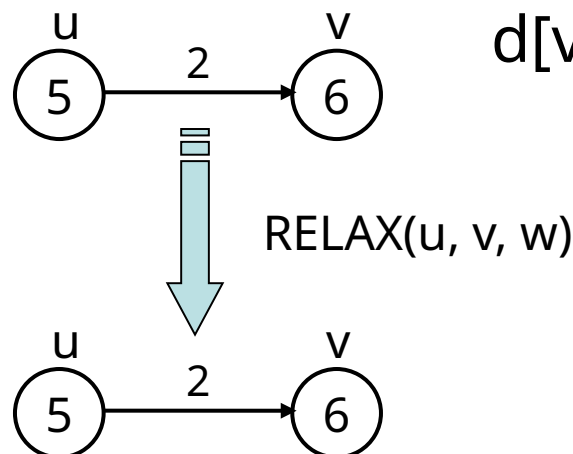
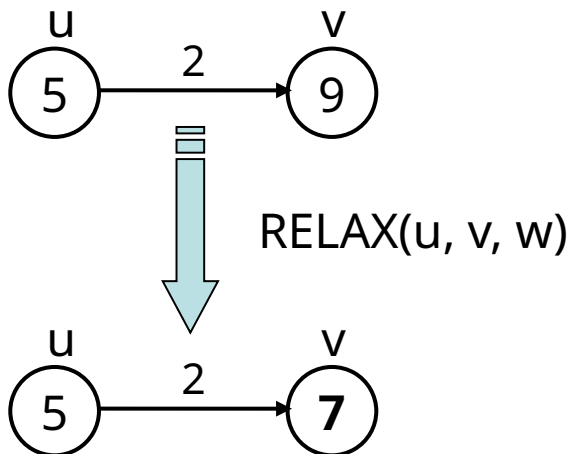
If  $d[v] > d[u] + w(u, v)$

we can improve the shortest path to  $v$

$\Rightarrow$  update  $d[v]$  and  $\pi[v]$

After relaxation:

$$d[v] \leq d[u] + w(u, v)$$



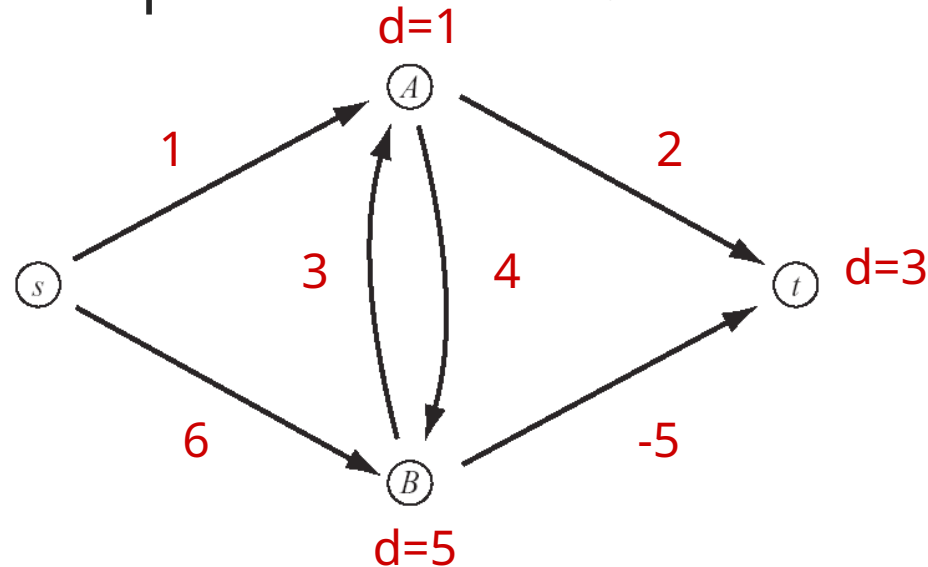
# Single Source Shortest Paths

---

- Bellman-Ford Algorithm
  - Allows negative edge weights
  - TRUE if no negative-weight cycles are reachable from the source  $s$  and FALSE otherwise
  - Traverse all the edges  $|V - 1|$  times, every time performing a relaxation step of each edge
- Single-Source Shortest Paths in DAGs
  - Topologically sort the vertices of the graph
  - Relax the edges according to the order given by the topological sort
- Dijkstra's Algorithm
  - No negative-weight edges
  - Repeatedly select a vertex with the minimum shortest-path estimate  $d[v]$  – uses a queue, in which keys are  $d[v]$

# Sample Problem

(a) Write down lengths for the edges of the following graph, so that Dijkstra's algorithm would not find the correct shortest path from  $s$  to  $t$ .



Dijkstra's algorithm will visit the vertices in the order  $s - A - t$   
 $d[t] = 3$  and  $d[B] = 5$   
However, the path  $s-A-B-t$  is the shortest.

# Sample Problem

---

(b) Which of the shortest path algorithms described in class would be most appropriate for finding paths in the graph of part (a) with the weights you gave? Explain your answer.

- Bellman-Ford, because it can handle graphs with negative edge weights and cycles.
- We can't use the DAG algorithm because this graph is not a DAG.

# True/False Questions

---

(a) TrueFalse

- Merge sort, Quicksort and Insertion sort are comparison-based sorting algorithms.

(b) True False

- In a red-black tree, if a node is black than both its children are red.

(c) TrueFalse

- A reverse-sorted array (i.e., decreasing order) is always a max-heap.

# True/False Questions

---

(d) True ☒ False

- In a directed graph with positive edge weights, the edge with minimum weight belongs to the shortest paths tree for any source vertex.

(e) True ☒ False

- Given any weighted directed graph with all distinct edge weights and any specified source vertex, the shortest paths tree is unique.

# True/False Questions

---

(f) True False

- After running DFS in a graph  $G = (V, E)$ , a vertex  $v (\in V)$  is a proper descendant of another vertex  $u (\in V)$   
 $\iff d[v] < d[u] < f[u] < f[v]$

(g) True False

- The problem of determining an optimal order for multiplying a chain of matrices can be solved by a greedy algorithm, since it displays the optimal substructure and overlapping subproblems properties.

# True/False Questions

---

(h) True False

- Kruskal's algorithm for finding a minimum spanning tree of a weighted, undirected graph is an example of a dynamic programming algorithm.



# True/False Questions

---

(i) TRUE FALSE

The depths of nodes in a red-black tree can be efficiently maintained as fields in the nodes of the tree.

- No, because the depth of a node depends on the depth of its parent
- When the depth of a node changes, the depths of all nodes below it in the tree must be updated
- Updating the root node causes  $n - 1$  other nodes to be updated

# Sample Questions

---

What do we mean by the running time of an algorithm?

- The running time of an algorithm is the number of (primitive) operations it performs before coming to a halt. The running time is expressed as a function of the input size  $n$ .

List at least three methods for solving recurrences.

- Masters method, iteration, substitution

# Sample Questions

---

Describe the characteristic feature of randomized algorithms. What is the main advantage of using such algorithms?

- The behavior is determined in part by values produced by a random-number generator: the algorithm generates its own randomness.
- Advantage: No particular input can consistently elicit worst case behavior.

# Sample Questions

---

What does dynamic programming have in common with divide-and-conquer, and what is the principal difference between the two techniques?

- **Common:** Both divide the initial problem into subproblems.
- **Difference:** Dynamic programming is used when the subproblems are not independent.

# Exercise

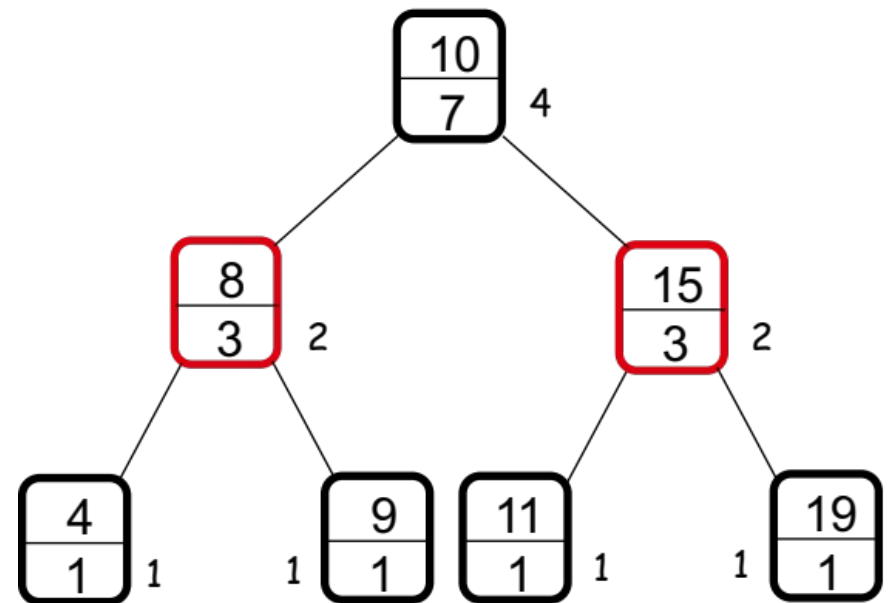
- In an OS-tree, the size field can be used to compute the **rank'** of a node  $x$ , **in the subtree for which  $x$  is the root**. If we want to store this rank in each of the nodes, show how can we maintain this information during insertion and deletion.

## Insertion

- add 1 to  $\text{rank}'[x]$  if  $z$  is inserted within  $x$ 's left subtree
- leave  $\text{rank}'[x]$  unchanged if  $z$  is inserted within  $x$ 's right subtree

## Deletion

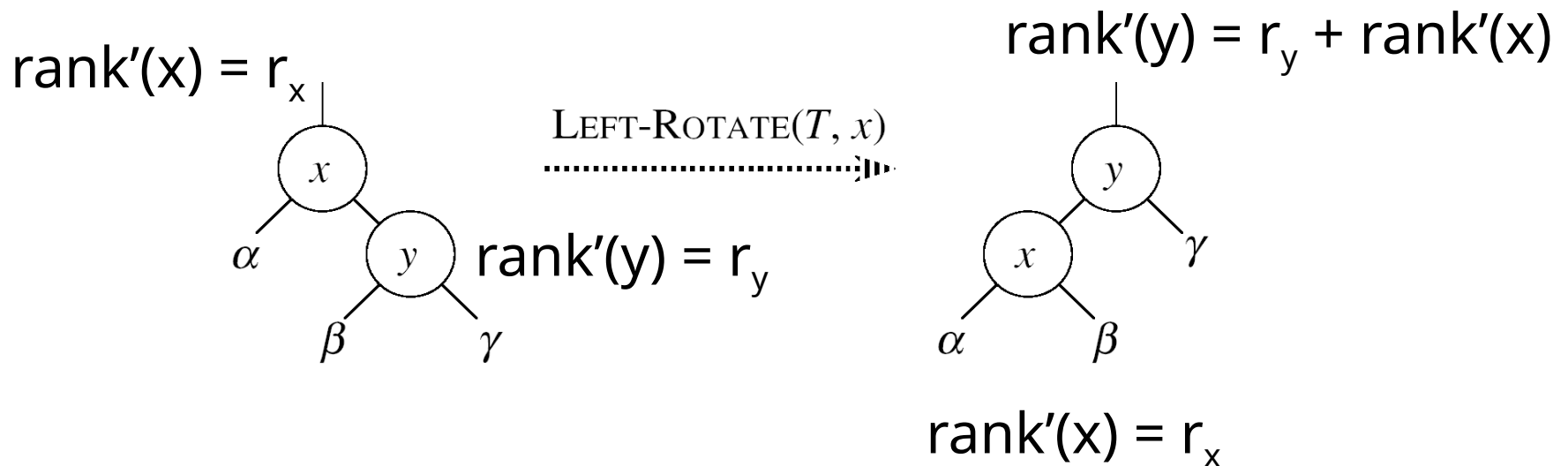
- subtract 1 from  $\text{rank}'[x]$  whenever the deleted node  $y$  had been in  $x$ 's left subtree.



$$\text{rank}'[x] = \text{size}[\text{left}] + 1$$

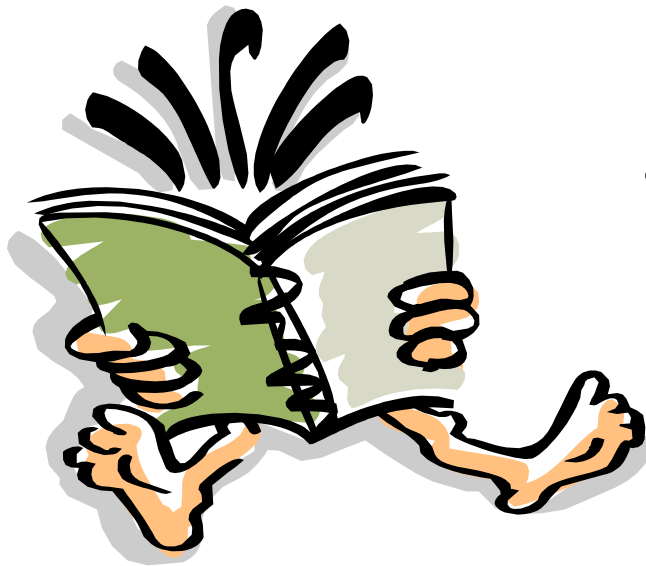
# Exercise (cont.)

- We also need to handle the rotations that occur during insertion and deletion



# Readings

---



- All topics covered during the semester