CS-3510: Design and Analysis of Algorithms

Exam 2: Review

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College of Computing Georgia Institute of Technology Summer 2022

Exam 2

- Date: Thursday, July 07, 2022
- Time: 03:30 pm 05:00 pm
- Location: Klaus 2443
- Closed book; No calculator
- One page sheet of notes
 - Letter size
 - Both sides
 - Typed or hand-written



No class on Tuesday (07/05)!



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Exam 2

- Contents:
 - Greedy algorithms
 - Graph algorithms
 - Definition and representation
 - Graph traversal (BFS, DFS)
 - Graph traversal applications
 - Minimum spanning tree
 - Shortest path in weighted graph (→ Final exam)



- Build the solution step-by-step
- At each step, make a decision that is locally optimal
- <u>Never look back</u> and hope for the best!
- Do NOT always yield optimal solutions, but for many problems they do



Exam 2: Greedy Choice Property

- Greedy choice = locally optimal choice
- Greedy-choice property: we can assemble a globally optimal solution by making locally optimal choices.
- In other words, when we are considering which choice to make, we make the choice that looks best in the current problem, without considering results from subproblems.

(The main difference with dynamic programming)

- Make whatever choice seems best at the moment and then solve the subproblem that remains.
- Makes its first choice before solving any subproblems.



Divide-and-Conquer

Dynamic Programming

Greedy Approach



Optimal substructure But only one subproblem



- Seems "easier" than dynamic programming?
- Two major "questions/problems":
 - What is the best/correct greedy choice to make?
 - How can we prove that the greedy algorithm yields an optimal solution?
- When is using the greedy approach a good idea?
 - Greedy can be optimal when the problem shows an <u>especially nice optimal</u> <u>substructure.</u>



8

problem

subproblem

Subsub

problem

• Examples

- Interval scheduling (activity selection)
- Interval partitioning
- Schedule to minimize lateness
- . . .

• Applications in Graph (next week)

- Kruskal's algorithm (minimum spanning tree)
- Prim's algorithm (minimum spanning tree)
- Dijkstra's algorithm (shortest path)

Type of Questions in Exam-2:

- Short answers
- Definition
- True/False questions



Exam 2: Graph

- Graph definition and representation
 - Adjacency matrix
 - Adjacency list
- Graph traversal
 - Breadth first search (BFS)
 - Shortest path (<u>unweighted</u> graphs)
 - Testing bipartiteness
 - Tree traversal (level-order)
 - Connected components
 - Depth first search (DFS)
 - Topological sorting
 - Tree traversal (in-order, pre-order, post-order)
 - Connected components

- Graph problems/algorithms
 - Minimum spanning tree (MST)
 - Kruskal (greedy)
 - Prim (greedy)
 - Shortest path (directed weighted graphs)
 - Dijkstra (greedy)
 - Bellman-Ford (dynamic programming)
 - Floyd-Warshall (dynamic programming)
 - Flow network
 - Max-flow min-cut theorem
 - Ford-Fulkerson algorithm



Graph Properties and Terminology Review

• Notation. G = (V, E)

- $V = \text{nodes} \text{ (or vertices). } \{0, 1, 2, ..., n-1\}$
- E = edges (or arcs) between pairs of nodes. { $e_1, e_2, \dots e_m$ } where $e_i = (v_i, v_j)$
- Captures pairwise relationship between objects.



Graph Representation: Summary

• Two common ways to represent graphs

- Adjacency matrix
- Adjacency list
- Adjacency matrix
 - Space: n² elements for n vertices
 - Easy to check if a link exists between two vertices
- Adjacency list
 - More common representation: most large real-world graphs are sparse
 - Space: Number of edges [2*(number of edges) if undirected] + number of vertices, i.e., (m+n) or (2m+n)
 - Linked list implementation is typically used



Graph Definitions and Terminology: Summary

- Paths and connectivity
- Connected graph, connected component
- Cycle
- DAG
- Bipartiteness
- Trees

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- True/False questions
- Graph representation



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Graph Traversal

- Traversal = Exploring = Searching
- A graph needs to be traversed in order to determine some properties

•	Breadth-first search (BFS)
	• Shortest path (unweighted graphs)

- Testing bipartiteness
- Tree traversal (level-order)
- Connected components
- Depth-first search (DFS)
 - Topological sorting
 - Tree traversal (in-order, pre-order, post-order)
 - Connected components

	Implementation	Data Structure		
BFS	Iterative	Queue (FIFO)		
DFS	Recursive	(not explicitly required \rightarrow execution stack)		
	Iterative	Stack (LIFO)		



- An efficient graph traversal procedure
- BFS starts from a source vertex "s"
- At each vertex u, all neighbors, i.e., vertices v adjacent to u are visited before moving on to vertices adjacent to some v
- <u>Iterative</u> implementation.
- Needs queue data structure

• Traversal = Exploring = Searching (visiting vertices one-by-one)



- BFS runs in O(|V| + |E|) time
- The worst case is when the graph is connected.
 - Each vertex is added to the queue and removed from it exactly once
 - Each adjacency list is used exactly once



- An efficient graph traversal procedure
- BFS starts from a source vertex "s"
- At each vertex u, all neighbors, i.e., vertices v adjacent to u are visited before moving on to vertices adjacent to some v



BFS(G, s)

for each vertex $u \in G.V - \{s\}$ u.color = WHITE2 3 $u.d = \infty$ $u.\pi = \text{NIL}$ parent s.color = GRAY $6 \ s.d = 0$ $s.\pi = \text{NIL}$ $Q = \emptyset$ ENQUEUE(Q, s)while $Q \neq \emptyset$ 10 u = DEQUEUE(Q)11 12 for each $v \in G.Adj[u]$ if v. color == WHITE 13 v.color = GRAY14 15 v.d = u.d + 116 $v.\pi = u$ 17 ENQUEUE(Q, ν) 18 u.color = BLACKblack := visited & all unvisited neighbors added to the queue

s} white := unvisited node distance from source parent

gray := visited node



Graph Traversal: BFS the "shortest distance" from the source!

Source: "s"

- An efficient graph traversal procedure
- BFS starts from a source vertex "s"
- At each vertex u, all neighbors, i.e., vertices v adjacent to u are visited before moving on to vertices adjacent to some v
- Queue = $\{A, B, C, F, D, E, G\}$
- Visited = $\{A, B, C, F, D, E, G\}$



d = 2

Nothing left in the queue \rightarrow All nodes are visited \rightarrow Halt

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- DFS follows a single path as far (deep) as possible and then backtracks to the last alternative path.
- No explicit storage of vertices is required (BFS needs a queue)
- However, calls for each vertex build up on the <u>execution stack</u> (<u>recursive</u> implementation)
- An <u>iterative</u> implementation is possible using an explicit <u>stack</u> data structure.
- Traversal = Exploring = Searching (visiting vertices one-by-one)



• DFS follows a single path as far (deep) as possible and then backtracks to the last alternative path



DFS(G) 1 for each vertex $u \in G.V$ 2 u.color = WHITE3 $u.\pi = NIL$ 4 time = 0 5 for each vertex $u \in G.V$

- if *u*.*color* == WHITE
 - DFS-VISIT(G, u)



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• DFS follows a single path as far (deep) as possible and then backtracks to the last alternative path

- Stack = $\{x, x, x, x, x, x, x, x\}$
- Visited = $\{A, B, C, D, E, G, F\}$
- No more path to explore \rightarrow backtrack
- No more element in the stack \rightarrow Halt





- DFS also runs in O(|V| + |E|) time
- DFS is called exactly once per vertex
- Each adjacency list is used exactly once

	Implementation	Data Structure	Running Time
BFS	Iterative	Queue (FIFO)	O(V + E)
DFS	Recursive	(not explicitly required \rightarrow execution stack)	O(V + E)
	<u>Iterative</u>	Stack (LIFO)	



BFS and DFS

• Both are graph traversal algorithms

DFS
Recursive: (execution stack), Iterative: Stack(LIFO) <u>Time</u> :O(IVI + IEI), <u>Space</u> : O(IVI)
The predecessor subgraph of a depth-first search forms a depth-first forest comprising several depth-first trees.
DFS timestamps each node with two numbers;
 d (discovery time) and f (finishing time). > The timestamps have parenthesis structure.



Type of Questions in Exam-2:

- Short answers /Definition/ True/False
- Running BFS/DFS on a given graph (show your steps)
 BFS/DFS trees, discovery/finishing times, ...

Breadth first search (BFS)

Depth first search (DFS)

Connectivity problem (Connected components)

Shortest path (unweighted graphs) Testing bipartiteness

Tree traversal

• level-order

Topological sorting Strongly connected components

Tree traversal

• In-order, Pre-order, post-order









Graph Traversal: Connected Component

• Ex1: Given a set of flight plans, can we travel from Atlanta (ATL) to London (LHR)?

JFK

ATL

source

- Flights:(JFK, ATL)
 - (ATL, LAX)
 - (LAX, SFO)
 - (JFK, SFO)
 - (SFO, JFK)
 - (JFK, LHR)

Define the corresponding graph Run BFS or DFS from the source node, i.e., the node associated with ATL During the traversal check if the destination (LHR) is a neighbor of the current node

Demo code time!

LH

destination



SFO

LA X

Graph Traversal: Connected Component

Ex2 [Grid problems]: Given an m-by-n 2D binary matrix in which 0 represent water and 1 represent land, design an algorithm computing the number islands. An island includes one or more horizontally or vertically cells surrounded by water.
 We know the nodes (= grid

cells) and we know the neighbors (the relationship), so we can skip the graph definition part!
Each cell = graph node Neighbors of grid[i][j]:



Demo code!



grid[i-1][j]

grid[i+1][j]

grid[i][j-1]

grid[i][j+1]

BFS: Shortest paths

• BFS intuition. Explore outward from s in all possible directions, adding nodes one "layer" at a time.

BFS algorithm.

- $L_0 = \{ s \}.$
- $L_1 =$ all neighbors of L_0 .
- L₂ = all nodes that do not belong to L₀ or L₁, and that have an edge to a node in L₁.

 $-L_2$ - $\cdot \cdot \cdot$

- L_{i+1} = all nodes that do not belong to an earlier layer, and that have an edge to a node in L_i .
- Theorem. For each *i*, *L_i* consists of all nodes at distance exactly *i* from *s*. There is a path from *s* to *t* if and only if *t* appears in some layer.



BFS: Shortest paths

• Property. Let *T* be a BFS tree of G = (V, E), and let (u, v) be an edge of *G*. Then, the levels of *u* and *v* differ by at most 1.



Lemma 1

3

BFS: Shortest paths

- Theorem: Correctness of BFS; Shortest Paths
 Let G = (V, E) be a directed/undirected graph, and BFS is run on G
 from a given source s ∈ V. Then, during the execution,
 - BFS discovers every vertex $v \in V$ that is reachable from the source s, and
 - Upon termination, $d = \delta(s, v)$ for all $v \in V$, where d is the distance computed by BFS.
 - Moreover, for any vertex $v \neq s$ that is reachable from s, one of the shortest paths from s to v is a shortest path from s to $\pi(v)$ followed by edge ($\pi(v), v$).



BFS: Testing Bipartiteness

• Def. A bipartite graph is an <u>undirected</u> graph G = (V, E) in which V can be partitioned into two sets V_1 and V_2 such that $(u, v) \in E$ implies either $u \in V_1$ and $v \in V_2$ or $u \in V_2$ and $v \in V_1$. That is, all edges go between the two sets V_1 and V_2 .

bipartite graph = 2-colorable graph

• Def. An undirected graph G = (V, E) is bipartite if the nodes can be colored blue or red such that every edge has one blue and one red end.

 V_{2}

BFS: Testing Bipartiteness (KT 3.4)

- Lemma. If a graph G is bipartite, it cannot contain an odd-length cycle.
- Proof. Not possible to 2-color the odd-length cycle, let alone G.



BFS: Testing Bipartiteness

- Lemma. Let *G* be a connected graph, and let *L*₀, ..., *L_k* be the layers produced by BFS starting at node *s*. Exactly one of the following holds:
 - 1. No edge of G joins two nodes of the same layer, and G is bipartite.
 - 2. An edge of *G* joins two nodes of the same layer, and *G* contains an odd-length cycle (and hence is not bipartite).







BFS: Testing Bipartiteness

- We can modify the BFS algorithm to color each neighbor with the opposite color when it explores a node.
- If a neighbor has already been colored (i.e., visited), and has the same color, then <u>return false</u>.
- If the BFS can traverse the entire graph and color all nodes, then <u>return</u> <u>true.</u>



• Def. A <u>directed acyclic graphs (DAG)</u> is a directed graph that contains no directed cycles.

Topological ordering = Topological sort (Top-Sort)

a topological ordering

• Def. A <u>topological order</u> of a directed graph G = (V, E) is an ordering of its nodes as $v_1, v_2, ..., v_n$ so that for every edge (v_i, v_j) we have i < j.





a DAG

- Def. A directed acyclic graphs (DAG) is a directed graph that contains no directed cycles.
- Def. A <u>topological order</u> of a directed graph G = (V, E) is an ordering of its nodes as $v_1, v_2, ..., v_n$ so that for every edge (v_i, v_j) we have i < j.
- Topological Ordering → Precedence Constraints
 - Precedence constraints: edge (v_i, v_j) means task v_i must occur before v_j .
- Applications
 - Course prerequisite graph: course v_i must be taken before v_j
 - Compilation: module v_i must be compiled before v_j
 - Pipeline of computing jobs: output of job v_i needed to determine input of job v_j



- If G has a topological order, then G is a DAG.
 - Q. Does every DAG have a topological ordering?
 - Q. If so, how do we compute one?
- If G is a DAG, then G has a topological ordering.
 - If G is a DAG, then G has a node with no entering edges.

G is a DAG \Leftrightarrow G has a topological ordering

• Algorithm finds a topological order (topological sort) in O(m + n) time.



• Algorithm finds a topological order in O(m + n) time

- TOPOLOGICAL-SORT
 - Call DFS to compute finishing times for each vertex v.
 - As each vertex is finished, insert it onto the front of a linked list
 - Return the linked list of vertices
- Pf. (CLRS, Theorem 22.12)
- Note topological ordering can also be obtained using "Kahn's algorithm", which is BFS approach starting from a node with no entering edge, in O(m + n) time.



- Problem: Decomposing a directed graph into its strongly connected components
- A classic application of DFS.
 - The <u>strongly connected components</u> of a <u>directed</u> graph are the equivalence classes of vertices under the "<u>are mutually reachable</u>" relation.
 - Given directed graph G=(V, E) an SCC is a maximal set of vertices $C \subseteq V$ such that for every pair of vertices u and v in C, we have both uvv and vvu; that is, vertices u and v are reachable from each other.
 - A directed graph is <u>strongly connected</u> if it has only <u>one</u> strongly connected component.



- Linear-time ($\Theta(|V| + |E|)$) algorithm to compute the strongly connected components of a directed graph G = (V, E) using two depth-first searches, one on G and one on G^{T} .
- G^T = (V, E^T), where E^T = {(u, v)|(v, u) ∈ E} In other words, same graph except all edges are reversed.
 Adjacency list representation: G^T can be obtained in O(|V| + |E|)
- **Observation:** <u>G and G^{T} have the same SCC's</u>. (*u* and v are reachable from each other in G if and only if reachable from each other in G^{T} .)



STRONGLY-CONNECTED-COMPONENTS (G)

- 1. DFS(G) to compute f(u)
- 2. Compute G^{T}
- 3. Call DFS(G^{T}) in the order of decreasing f(u) (topology ordering of G)
- 4. Each tree in the depth-first forest formed in line 3 is a strongly connected component

• Lemma

Let *C* and *C'* be distinct SCCs in G = (V, E). Suppose there is an edge $(u, v) \in E$ such that $u \in C$ and $v \in C'$. Then f(C) > f(C').

• <u>Corollary-1</u> Suppose there is an edge $(u,v) \in E^T$, where $u \in C$ and $v \in C'$. Then f(C) < f(C').

C

U

• <u>Corollary-2</u> Suppose f(C) > f(C'). Then, there cannot be an edge from C' to C in G^T .



• When we start the second DFS on G^T:

- We begin with SCC C such that f(C) is maximum.
- So, the second DFS starts from some $x \in C$, which visits all C vertices.
- Corollary-2 says that since f(C) > f(C') for all $C' \neq C$, there are no edges from C to C' in G^{T} . Therefore, the second DFS only visits vertices in C, i.e., the depth-first tree rooted at x contains *exactly* the vertices of C.
- <u>The next root chosen in the second DFS</u> is in SCC C' such that f(C') is maximum over all SCCs other than C. DFS visits all vertices in C', but the only edges out of C' go to C, which we have already visited. Therefore, the only tree edges will be to vertices in C'.
- <u>We can continue the process</u>. Each time we choose a root based on the topological order, where we have only edges to the current SCC nodes (and the earlier ones but they are already visited), and there is no edge to the next SCC (Corollary-2); therefore, the DFS only visits the current SCC nodes.



С

u



level-order

• In-order, Pre-order, post-order



Exam 2: Graph

- Graph definition and representation
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- Graph traversal
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 - Shortest path (<u>unweighted</u> graphs)
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Type of Questions in Exam-2:

- Short answers /Definition/ True/False
- Designing (explaining) an algorithm for a graph-related problem
 - Detecting that the problem is a graph-related problem (Explain how we can formulate the given problem as a graph problem, and how the given information can be represented as a graph, i.e., adjacency list, adjacency matrix, etc.
 - Which one of the discussed algorithms (DFS, BFS, testing bipartiteness, SCC, topology ordering can be employed to solve the given problem and justify (explain) the correctness of your approach.
 - Discuss the overall time complexity. (The running time takes to create the corresponding graph and the running time takes to solve the problem)
 - Tree traversal (level-order)
 - Connected components
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Minimum Spanning Tree

- Weighted graphs
 - Each edge has an associated weight, cost, or distance.
 - Edge $(u, v) \rightarrow w(u, v)$
- Spanning tree
 - Given graph G = (V, E), a tree $T = (V, E_T)$ such that $E_T \subseteq E$ is a spanning tree of G.
 - Tree T spans the graph G



Minimum Spanning Tree (MST)

- Spanning tree
 - Given graph G = (V, E), a tree $T = (V, E_T)$ such that $E_T \subseteq E$ is a spanning tree of G.
 - Tree T spans the graph G
- Minimum spanning tree = <u>Minimum-weight</u> spanning tree
- Spanning tree T for G such that the sum $w(T) = \sum w(u, v)$ is minimized

 $(u,v)\in T$

- Approach: "Greedy choice"
- Algorithms:
 - Kruskal
 - Prim



Growing a Minimum Spanning Tree

- This greedy strategy is captured by the following generic method, which grows the minimum spanning tree one edge at a time.
- The generic method manages <u>a set of edges A</u>, maintaining the following loop invariant:
 - Prior to each iteration, A is a subset of some minimum spanning tree.
- At each step, we determine an edge (*u*, *v*) that we can add to A without violating this invariant *A* ∪ {(*u*, *v*)} is also a subset of an MST
- An edge is safe edge if adding it to A will not violate the invariant.



Some Definitions

• Cut

• A cut (S, V - S) of an undirected graph G = (V, E) is a partition of V.



- With this definition, we say
 - An edge $(u, v) \in E$ crosses the cut (S, V S) if one of this endpoints is in *S*, and the other in V S
 - A cut <u>respects</u> a set A of edges if no edge in A crosses the cut.
 - An edge is <u>a light edge</u> crossing a cut if its weight is the <u>minimum</u> of any edge crossing the cut.



Generic-MST

• Theorem:

Let G = (V, E) be a connected, <u>undirected</u> graph with a real-valued <u>weight</u> function w defined on E. Let A be a subset of E that is included in some minimum spanning tree for G, let (S, V - S) be any cut of G that respects A, and let (u, v) be a light edge crossing this cut. Then edge (u, v) is safe for A. GENERIC-MST(G, w)

$$A = \emptyset$$

2 while A does not form a spanning tree

find an edge
$$(u, v)$$
 that is safe for A

$$A = A \cup \{(u, v)\}$$

5 return A

Generic-MST

• Notes

- The set A is always <u>acyclic</u>.
- At any point $G_A = (V, A)$ is a forest
- At first when $A = \phi$, we have |V| trees 4 A = 1 in the forest G_A , each a tree of one vertices 5 **return** A

GENERIC-MST(G, w)

$$A = \emptyset$$

3

- while A does not form a spanning tree
 - find an edge (u, v) that is safe for A

$$A = A \cup \{(u, v)\}$$

- At each iteration, the number of trees is reduced by one.
- While loop (line 2-4) runs for |V|-1 times to find the edges required to form the minimum spanning <u>tree</u>.
- The method terminates when we have one tree (clearly, with |V|-1 edges).



Generic-MST

- Let G = (V, E) be a connected, <u>undirected</u> graph with a real-valued <u>weight</u> function w defined on E. Let A be a subset of E that is included in some minimum spanning tree for G,
- [Theorem:] let (S, V S) be any cut of G that respects A, and let (u, v) be a light edge crossing this cut. Then edge (u, v) is safe for A.
- [Corollary:] let $C = (V_C, E_C)$ be a connected component (tree) in the forest $G_A = (V, A)$. If (u, v) is a light edge connecting C to some other component in G_A , Then edge (u, v) is safe for A.
 - Pf. Cut $(V_C, V V_C)$ respects A, and (u, v) is a light edge for this cut \rightarrow safe

MST Algorithms

- Kruskal's algorithm
 - The set A is a forest whose vertices are all those of the given graph.
 - The safe edge added to A is always a least-weight edge in the graph that connects two distinct components. (so it is not creating a loop)
- Prim's algorithm
 - The set A forms a single tree.
 - The safe edge added to A is always a least-weight edge connecting the tree to a vertex not in the tree.



MST: Summary

- Spanning tree
 - Given graph G = (V, E), a tree $T = (V, E_T)$ such that $E_T \subseteq E$ is a spanning tree of G.
 - Tree T spans the graph G
- Minimum spanning tree
 - Spanning tree T for G such that the sum $w(T) = \sum_{(u,v) \in T} w(u,v)$ is minimized

Algorithm	Paradigm	Data Structure Used	Running Time
Kruskal	Greedy	Disjoint-Set (Union-Find)	O(E log V)
Prim	Greedy	Priority Queue (Binary Min-Heap)	O(E log V)



Type of Questions in Exam-2:

- Short answers /Definition/ True/False
- Running Kruskal's/Prim's algorithms on a given graph (show your steps)
- Proving some properties of minimum spanning trees.
 - Similar to HW4/Q3
 - Proof by contradiction (assume the given statement is not correct, then show this assumption will cause some contradictions → Thus, the given statement is true.)
 - Spanning tree T for G such that the sum $w(T) = \sum_{(u,v)\in T} w(u,v)$ is minimized

Algorithm	Paradigm	Data Structure Used	Running Time
Kruskal	Greedy	Disjoint-Set (Union-Find)	O(E log V)
Prim	Greedy	Priority Queue (Binary Min-Heap)	O(E log V)



Exam 2: Practice Problems

Course website

#	Assignment	Release Date	Deadline
1	hw1:	05/20	05/27
	[pdf I tex I solution]		
2	hw2:	05/27	06/03
	[pdf I tex I solution]		
3	hw3:	06/03	06/17
	[pdf I tex I solution]		
4	hw4:	06/21	07/02
	[pdf tex solution]		
5	hw5:	07/08	07/16
	[pdf tex solution]		

home policies lectures assignments resources

You can use this LaTeX template file to prepare your solutions on the cloud-based LaTeX editor OverLeaf.

#	Exam	Date (mm/dd)	Time (EST)	Location	
1	Exam 1: Complexity, Devide-and-Conquer, Dynam Programming [practice pdf solution]	06/09 ic Thursday	03:30 pm	Klaus 2443	
2	Exam 2: Greedy Algorithms, Graph Algorithms [practice pdf solution]	07/07 Thursday	03:30 pm	Klaus 2443	
3	Final Exam: Inclusive (including all discussed topics) [practice solution]	07/28 Thursday	03:00 pm	Klaus 2443	



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