ECE 374 B: Algorithms and Models of Computation, Summer 2024 Midterm 2 – July 23, 2024

- You will have 90 minutes (1.5 hours) to solve all the problems. Most have multiple parts. Don't spend too much time on questions you don't understand and focus on answering as much as you can! Make sure to check both sides of all the pages and make sure you answered everything.
- No resources are allowed for use during the exam except a multi-page cheatsheet and scratch paper on the back of the exam. Do not tear out the cheatsheet or the scratch paper! It messes with the auto-scanner.
- You should write your answers *completely* in the space given for the question. We will not grade parts of any answer written outside of the designated space.
- Please *use a dark-colored pen* unless you are *absolutely* sure your pencil writing is forceful enough to be legible when scanned. We will take off points if we have difficulty reading the uploaded document.
- Incorrect algorithms will receive a score of 0, but slower than necessary but correct
 algorithms will always receive some points, even brute force ones. Thus, you should
 prioritize the correctness of your submitted algorithms over speed; you will receive more
 points that way. On the other hand, submit the fastest algorithms that you know are
 correct; faster algorithms will receive more points.
- Any recursive backtracking algorithm or dynamic programming algorithm given without an *English* description of the recursive function (i.e., a description of the output of the function *in terms of their inputs*) will receive a score of 0.
- Any greedy algorithm or a modification of a standard graph algorithm given without a proof of correctness will receive a score of 0.
- For problems with a graph given as input, you may assume the graph is simple (i.e., it has no self-loops or parallel edges).
- Only algorithms referenced in the cheat sheet may be referred to as a "black box". You may not simply refer to a prior lab/homework for the solution and must give the full answer.
- Unless explicitly mentioned, a runtime analysis is required for each given algorithm.
- Don't cheat. If we catch you, you will get an F in the course.
- · Good luck!

Name:
VetID:
Date:

1 Short answer (2 questions) - 15 points

Answer the following questions. You may **briefly** (no more than 2 sentences) justify your answers, but a complete proof is not required.

(a) Give a *tight* asymptotic upper-bound for the following recurrences :

(i)

$$A(n) = 2A(n-1) + n/2$$
 $A(0) = A(1) = A(2) = A(3) = 1$

(ii)

$$B(n) = 2B(n/2) + n$$
 $B(0) = B(1) = 1$

(b) Give the asymptotic running time that describes the following program assuming n > 0. (Hint: think about how the work grows with as the input increases, if you're thinking there's a typo, there isn't):

```
function foo(int n)
   if n > 100
      return 1
   else
      return foo(n+1) + foo(n+2) + foo(n+3)
```

Running time:

2 Short answer II (2 questions) - 10 points

Answer the following questions. You *may* **briefly** (no more than 2 sentences) justify your answers, but a complete proof is not required.

(a) Recall in lecture/discussion we discussed the median of median (linear time selection) algorithm. The algorithm we discussed breaks a array into lists of size five. What if we break the array into lists of size 9. What is the recurrence and running time for the this modified version of linear time selection

(b) Give an example of a problem that has a recursive solution but that can not be efficiently memoized with dynamic programming.

3 Recursion - 15 points

Given a decimal number as input, we need to write a program to convert the given decimal number into an equivalent binary number.

Input: 7
Output: 111
Input: 10
Output: 1010

4 Dynamic programming - 15 points

A sequence is bitonic if it monotonically increases and then monotonically decreases. For example the sequences $\langle 1, 4, 6, 8, 3, -2 \rangle$ and $\langle 1, 2, 3, 4 \rangle$ are bitonic, but $\langle 1, 3, 12, 4, 2, 10 \rangle$ is not bitonic.

Given a sequence of n distinct integers (\mathbb{A}), give a efficient algorithm to find the largest bitonic subsequence in \mathbb{A} .

(You cannot simply reference a prior lab/homework. You must give the actual solution.)

Recurrence and short English description(in terms of the parameters):

Memoization data structure and evaluation order:

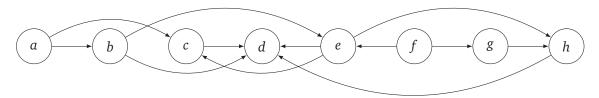
Return value:

Time Complexity:

5 Short answer IV (4 questions) - 20 points

Answer the following questions. You *may* **briefly** (no more than 2 sentences) justify your answers, but a complete proof is not required. For the following graph problems, use the notation G = (V, E), n = |V| and m = |E|

(a) Consider the following graph:



What is the order of vertices visited using the vanilla Depth first search algorithm starting from vertex a. When there is a coice for whihe vertex to visit, choose the vertex according to alphabetical order:

(b) You are given a **connected**, **undirected** graph and after running DFS you realize there are no backward, forward or cross edges! how many paths are there between any two vertices $u, v \in V$.

(c) Give an example of a graph with atleast 5 nodes, that has the same BFS and DFS vertex traversals.

(d) What type of graph with n nodes has the few possible topological sorts?

6 Short answer V (3 questions) - 15 points

Answer the following questions. You *may* **briefly** (no more than 2 sentences) justify your answers, but a complete proof is not required. For the following graph problems, use the notation G = (V, E), n = |V| and m = |E|

(a) I have a directed graph with all positive edge weights. Give an efficient algorithm to find the all shortest paths between any two vertices $i, j \in V$.

(b) Given a directed graph G = (V, E), how do we efficiently find all the vertices that can reach u.

(c) You are given a directed graph G = (V, E) but instead of the edges having weights, now the vertices have an associated cost and edges can no cost. Describe an efficient algorithm to find the path with the smallest cost (minimize the vertex costs you have to pay).

7 Graph algorithm - 10 points

You are given three weighted, directed acyclic graphs ($F = (V, E_F)$, $G = (V, E_G)$ and $H = (V, E_H)$) that have the same vertices but different edges. Every edge is assigned an integer value that could be positive or negative.

You are given two nodes s and t. The length of a path in each graph is defined as $\ell_F(s,t)$ for F, $\ell_G(s,t)$ for G and $\ell_H(s,t)$ for H. You need to find the path that exists in both G and H with the minimum combined length (in other words, find minimum value of $\ell_F(s,t) + \ell_G(s,t) + \ell_H(s,t)$). (Hint: topological sort is your friend)

(You *cannot* simply reference a prior lab/homework. You must give the actual solution.) **Note:** These details don't necessarily matter but might be useful to some. Both graphs are given as adjacency matrices where 0 indicates no edge and 1 indicates an edge. The weights are given by a weight matrix $w_F[i,j]$, $w_G[i,j]$ or $w_H[i,j]$ for F, G and H respectively. Remember pseudo-code is one of the worst ways to describe your algorithm(s); be clear and succinct.

This page is for additional scratch work!

ECE 374 B Algorithms: Cheatsheet

1 Recursion

Simple recursion

- · Reduction: solve one problem using the solution to another.
- · Recursion: a special case of reduction reduce problem to a smaller instance of itself (self-reduction).

- **Definitions** Problem instance of size n is reduced to *one or more* instances of size n-1 or less.
 - For termination, problem instances of small size are solved by some other method as base cases

Arguably the most famous example of recursion. The goal is to move n disks one at a time from the first peg to the last peg

Hanoi (n, src, dest, tmp): if (n > 0) then **Hanoi** (n-1, src, tmp, dest)Move disk n from src to dest **Hanoi** (n-1, tmp, dest, src)

Tower of Hano

Divide and conquer

Divide and conquer is an algorithm paradigm involving the decomposition of a problem into the same subproblem, solving them separately and combining their results to get a solution for the original problem

	Algorithm	Runtime	Space
Sorting algo-	Mergesort	$O(n \log n)$	$O(n \log n)$ O(n) (if optimized)
rithms	Quicksort	$O(n^2)$	O(n)
		$O(n \log n)$ if using MoM	· /

algorithm

We can divide and conquer multiplication like so:

$$bc = 10^n b_L c_L + 10^{n/2} (b_L c_R + b_R c_L) + b_R c_R.$$

We can rewrite the equation as:

$$\begin{split} bc &= b(x)c(x) = (b_L x + b_R)(c_L x + c_R) = (b_L c_L)x^2 \\ &+ ((b_L + b_R)(c_L + c_R) - b_L c_L - b_R c_R)x \\ &+ b_R c_R, \end{split}$$

Its running time is $O(n^{\log_2 3}) = O(n^{1.585})$

Recurrences

Suppose you have a recurrence of the form T(n) = rT(n/c) + f(n).

The master theorem gives a good asymptotic estimate of the recurrence. If the work at each level is:

$$\begin{array}{ll} \text{Decreasing: } rf(n/c) = \kappa f(n) \text{ where } \kappa < 1 & T(n) = O(f(n)) \\ \text{Equal: } & rf(n/c) = f(n) & T(n) = O(f(n) \cdot \log_c n) \\ \text{Increasing: } & rf(n/c) = Kf(n) \text{ where } K > 1 & T(n) = O(n^{\log_c r}) \end{array}$$

Some useful identities:

- Sum of integers: $\sum_{k=1}^{n} k = \frac{n(n+1)}{2}$
- Geometric series closed-form formula: $\sum_{k=0}^{n} ar^k = a \frac{1-r^{n+1}}{1-r}$
- Logarithmic identities: $\log(ab) = \log a + \log b, \log(a/b) = \log a \log b, a^{\log_C b} = b^{\log_C a}$ $(a, b, c > 1), \log_a b = \log_c b / \log_c a.$

Backtracking

Backtracking is the algorithm paradigm involving guessing the solution to a single step in some multi-step process and recursing backwards if it doesn't lead to a solution. For instance, consider the longest increasing subsequence (LIS) problem. You can either check all possible subsequences:

```
algLISNaive(A[1..n]):
  maxmax = 0
  for each subsequence B of A do
     if B is increasing and |B| > \max then
       max = |B|
  return max
```

On the other hand, we don't need to generate every subsequence; we only need to generate the subsequences that are increasing:

```
LIS_smaller(A[1..n], x):
   if n=0 then return 0
    \max = \mathbf{LIS\_smaller}(A[1..n-1], x)
   \text{if } A[n] < x \text{ then }
      \max = \max \left\{ \max, 1 + \text{LIS\_smaller}(A[1..(n-1)], A[n]) \right\}
```

Linear time selection

The median of medians (MoM) algorithms give a element that is larger than $\frac{3}{10}$'s and smaller than $\frac{7}{10}$'s of the array elements. This is used in the linear time selection algorithm to find element of rank k.

```
Median-of-medians (A. i):
   sublists = [A[j:j+5] for j \leftarrow 0, 5, \dots, \text{len}(A)]
   medians = [sorted (sublist)[len (sublist)/2]
             \textbf{for} \ \text{sublist} \in \text{sublists}]
   // Base case if len (A) \leq 5 return sorted (a)[i]
    // Find median of medians
   if len (medians) \leq 5
      pivot = sorted (medians)[len (medians)/2]
       pivot = Median-of-medians (medians, len/2)
    // Partitioning step
   low = [j \text{ for } j \in A \text{ if } j < pivot]
   high = [j \text{ for } j \in A \text{ if } j > pivot]
   k = len (low)
   if i < k
      return Median-of-medians (low, i)
      return Median-of-medians (low, i-k-1)
   else
   return pivot
```

Dynamic programming

Dynamic programming (DP) is the algorithm paradigm involving the computation of a recursive backtracking algorithm iteratively to avoid the recomputation of any particular subproblem.

Longest increasing subsequence

The longest increasing subsequence problem asks for the length of a longest increasing subsequence in a unordered sequence, where the sequence is assumed to be given as an array. The recurrence can be written as:

$$\mathit{LIS}(i,j) = \begin{cases} 0 & \text{if } i = 0 \\ \mathit{LIS}(i-1,j) & \text{if } A[i] \geq A[j] \\ \max \left\{ \begin{array}{ll} \mathit{LIS}(i-1,j) & \text{else} \\ 1 + \mathit{LIS}(i-1,i) \end{array} \right. \end{cases}$$

$\begin{aligned} & \textbf{LIS-Iterative}(A[1..n]): \\ & A[n+1] = \infty \\ & \textbf{for } j \leftarrow 0 \textbf{ to } n \\ & \textbf{ if } A[i] \leq A[j] \textbf{ then } LIS[0][j] = 1 \end{aligned}$ $& \textbf{for } i \leftarrow 1 \textbf{ to } n-1 \textbf{ do} \\ & \textbf{ for } j \leftarrow i \textbf{ to } n-1 \textbf{ do} \\ & \textbf{ if } A[i] \geq A[j] \\ & LIS[i,j] = LIS[i-1,j] \\ & \textbf{ else} \\ & LIS[i,j] = \max \left\{LIS[i-1,j], \\ & 1 + LIS[i-1,i] \right\} \end{aligned}$ & return LIS[n,n+1]

Edit distance

The edit distance problem asks how many edits we need to make to a sequence for it to become another one. The recurrence is given as:

$$\mathrm{Opt}(i,j) = \min \begin{cases} \alpha_{x_i y_j} + \mathrm{Opt}(i-1,j-1), \\ \delta + \mathrm{Opt}(i-1,j), \\ \delta + \mathrm{Opt}(i,j-1) \end{cases}$$

Base cases:
$$\mathrm{Opt}(i,0) = \delta \cdot i$$
 and $\mathrm{Opt}(0,j) = \delta \cdot j$

$$EDIST(A[1..m], B[1..n])$$
 for $i \leftarrow 1$ to m do $M[i, 0] = i\delta$ for $j \leftarrow 1$ to n do $M[0, j] = j\delta$
$$\text{for } i = 1 \text{ to } m \text{ do}$$
 for $j = 1$ to n do
$$\text{for } j = 1 \text{ to } n \text{ do}$$

$$M[i][j] = \min \begin{cases} COST[A[i]][B[j]] \\ + M[i-1][j-1], \\ \delta + M[i-1][j], \\ \delta + M[i][j-1] \end{cases}$$

2 Graph algorithms

Graph basics

A graph is defined by a tuple G=(V,E) and we typically define n=|V| and m=|E|. We define (u,v) as the edge from u to v. Graphs can be represented as **adjacency lists**, or **adjacency matrices** though the former is more commonly used.

- path: sequence of distinct vertices v_1, v_2, \dots, v_k such that $v_i v_{i+1} \in E$ for $1 \le i \le k-1$. The length of the path is k-1 (the number of edges in the path). Note: a single vertex u is a path of length 0.
- cycle: sequence of distinct vertices v_1, v_2, \ldots, v_k such that $(v_i, v_{i+1}) \in E$ for $1 \le i \le k-1$ and $(v_k, v_1) \in E$. A single vertex is not a cycle according to this definition.

Caveat: Sometimes people use the term cycle to also allow vertices to be repeated; we will use the term tour.

- A vertex u is connected to v if there is a path from u to v.
- The $\emph{connected component}$ of u, $\emph{con}(u)$, is the set of all vertices $\emph{connected}$ to u.
- A vertex u can reach v if there is a path from u to v. Alternatively v can be reached from u. Let $\operatorname{rch}(u)$ be the set of all vertices reachable from u.

Directed acyclic graphs

Directed acyclic graphs (dags) have an intrinsic ordering of the vertices that enables dynamic programming algorithms to be used on them. A topological ordering of a dag G=(V,E) is an ordering \prec on V such that

if $(u, v) \in E$ then $u \prec v$.

```
Kahn(G(V, E).u): toposort—empty list for v \in V: \operatorname{in}(v) \leftarrow |\{u \mid u \rightarrow v \in E\}| while v \in V that has \operatorname{in}(v) = 0: Add v to end of toposort Remove v from V for v in u \rightarrow v \in E: \operatorname{in}(v) \leftarrow \operatorname{in}(v) - 1 return toposort
```

Running time: O(n+m)

- · A dag may have multiple topological sorts.
- A topological sort can be computed by DFS, in particular by listing the vertices in decreasing post-visit order.

DFS and BFS

```
Pseudocode: Explore (DFS/BFS)
```

```
\begin{aligned} & \textbf{Explore}(G,u): \\ & \textbf{for } i \leftarrow 1 \text{ to } n: \\ & \text{Visited}[i] \leftarrow \text{False} \\ & \text{Add } u \text{ to ToExplore and to } S \\ & \text{Visited}[u] \leftarrow \text{True} \\ & \text{Make tree } T \text{ with root as } u \\ & \textbf{while B is non-empty do} \\ & \text{Remove node } x \text{ from B} \\ & \textbf{for each edge } (x,y) \text{ in } Adj(x) \textbf{ do} \\ & \text{if Visited}[y] \leftarrow \text{True} \\ & \text{Add } y \text{ to B, } S, T \text{ (with } x \text{ as parent)} \end{aligned}
```

Note:

- · If B is a queue, Explore becomes BFS.
- If B is a stack, *Explore* becomes DFS.

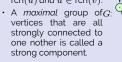
Pre and post numbering aids in analyzing the graph structure. By looking at the numbering we can tell if a edge (u,v) is a:

Pre/post numberina

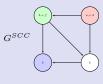
- $\cdot \ \textit{Forward edge:} \ \mathsf{pre}(u) < \mathsf{pre}(v) < \mathsf{post}(v) < \mathsf{post}(u)$
- Backward edge: pre(v) < pre(u) < post(u) < post(v)
- Cross edge: pre(u) < post(u) < pre(v) < post(v)

Strongly connected components

• Given G, u is strongly connected to v if $v \in \operatorname{rch}(u)$ and $u \in \operatorname{rch}(v)$.







Pseudocode: Metagraph - linear time

```
 \begin{aligned} & \mathsf{Metagraph}(G(V,E)): \\ & \mathsf{Compute}\ \mathsf{rev}(G)\ \mathsf{by}\ \mathsf{brute}\ \mathsf{force} \\ & \mathsf{ordering}\ \leftarrow \mathsf{reverse}\ \mathsf{postordering}\ \mathsf{of}\ V\ \mathsf{in}\ \mathsf{rev}(G) \\ & \mathsf{by}\ \mathsf{DFS}(\mathsf{rev}(G),s)\ \mathsf{for}\ \mathsf{any}\ \mathsf{vertex}\ s \\ & \mathsf{Mark}\ \mathsf{all}\ \mathsf{nodes}\ \mathsf{as}\ \mathsf{unvisited} \\ & \mathsf{for}\ \mathsf{each}\ u\ \mathsf{in}\ \mathsf{ordering}\ \mathsf{do} \\ & \mathsf{if}\ u\ \mathsf{is}\ \mathsf{not}\ \mathsf{visited}\ \mathsf{and}\ u \in V\ \mathsf{then} \\ & S_u \leftarrow \mathsf{nodes}\ \mathsf{reachable}\ \mathsf{by}\ u\ \mathsf{by}\ \mathsf{DFS}(G,u) \\ & \mathsf{Output}\ S_u\ \mathsf{as}\ \mathsf{as}\ \mathsf{strong}\ \mathsf{connected}\ \mathsf{component} \\ & G(V,E) \leftarrow G - S_u \end{aligned}
```

Shortest paths

Dijkstra's algorithm:

Find minimum distance from vertex \boldsymbol{s} to **all** other vertices in graphs without negative weight edges.

```
Pseudocode: Dijkstra  \begin{aligned} & \text{for } v \in V \text{ do} \\ & d(v) \leftarrow \infty \\ & X \leftarrow \varnothing \\ & d(s,s) \leftarrow 0 \\ & \text{for } i \leftarrow 1 \text{ to } n \text{ do} \\ & v \leftarrow \arg\min_{u \in V - X} d(u) \\ & X = X \cup \{v\} \\ & \text{for } u \text{ in Adj}(v) \text{ do} \\ & d(u) \leftarrow \min \left\{ (d(u), \ d(v) + \ell(v,u)) \right\} \end{aligned}
```

Running time: $O(m+n\log n)$ (if using a Fibonacci heap as the priority queue)

Bellman-Ford algorithm:

Find minimum distance from vertex s to **all** other vertices in graphs without negative cycles. It is a DP algorithm with the following recurrence:

$$d(v,k) = \begin{cases} 0 & \text{if } v = s \text{ and } k = 0 \\ \infty & \text{if } v \neq s \text{ and } k = 0 \\ \min \begin{cases} \min_{uv \in E} \left\{ d(u,k-1) + \ell(u,v) \right\} \\ d(v,k-1) \end{cases} & \text{else} \end{cases}$$

Base cases: d(s,0)=0 and $d(v,0)=\infty$ for all $v\neq s$.

```
\begin{aligned} & \textbf{for } \mathsf{each} \ v \in V \ \textbf{do} \\ & d(v) \leftarrow \infty \\ & d(s) \leftarrow 0 \end{aligned} \begin{aligned} & for \ k \leftarrow 1 \ \mathsf{to} \ n - 1 \ \textbf{do} \\ & for \ \mathsf{each} \ v \in V \ \textbf{do} \\ & for \ \mathsf{each} \ v \in V \ \textbf{do} \\ & for \ \mathsf{each} \ \mathsf{edge} \ (u,v) \in \mathsf{in}(v) \ \textbf{do} \\ & d(v) \leftarrow \min\{d(v), d(u) + \ell(u,v)\} \end{aligned} \end{aligned} \mathsf{return} \ d
```

Running time: O(nm)

Floyd-Warshall algorithm:

Find minimum distance from *every* vertex to *every* vertex in a graph *without* negative cycles. It is a DP algorithm with the following recurrence:

$$d(i,j,k) = \begin{cases} 0 & \text{if } i=j \\ \infty & \text{if } (i,j) \not\in E \text{ and } k=0 \\ \min \begin{cases} d(i,j,k-1) & \text{else} \end{cases} \end{cases}$$

Then d(i, j, n - 1) will give the shortest-path distance from i to j.

```
\begin{aligned} & \mathsf{Metagraph}(G(V,E)): \\ & \text{ for } i \in V \text{ do} \\ & d(i,j,0) \leftarrow \ell(i,j) \\ & (*\ell(i,j) \leftarrow \infty \text{ if } (i,j) \not\in E, \text{ } 0 \text{ if } i=j \text{ } *) \end{aligned} \begin{aligned} & \mathsf{for } k \leftarrow 0 \text{ to } n-1 \text{ do} \\ & \mathsf{for } i \in V \text{ do} \\ & \mathsf{for } j \in V \text{ do} \end{aligned} & d(i,j,k) \leftarrow \min \begin{cases} d(i,j,k-1), \\ d(i,k,k-1) + d(k,j,k-1) \end{cases} & \mathsf{for } v \in V \text{ do} \\ & \mathsf{if } d(i,i,n-1) < 0 \text{ then} \\ & \mathsf{return } \exists \text{ negative cycle in } G^* \end{aligned} & \mathsf{return } d(\cdot,\cdot,n-1) \end{aligned}
```

Running time: $\Theta(n^3)$