CPSC 320 2017W2: Quiz 3 Blanks

1 Ol' McDonald is safer, he hi he hi ho

1.1 Ol' McDonald is safer, he hi he hi ho (Group)

The manager in a McDonald's comes to you with the following problem: she is in charge of a group of n workers. Each worker works one shift each day of the week (i.e., a particular worker works from the same start time to the same finish time each day, though two different workers may work different times). There can be multiple shifts happening at once. Assume that no shift starts before midnight or ends after midnight and that shifts overlap even if they just "meet" at their start or end times.

The manager is trying to choose a subset of these n workers to form a safety committee that she can meet with once a week. She considers such a committee to be complete if, for every worker X not on the committee, X's shift overlaps at least partially the shift of some worker who is on the committee. In this way, each worker's obedience to safety protocols can be observed by at least one person who is serving on the committee.

Example: Suppose that n = 3 and the shifts are

- Worker A: 0:00 to 10:00
- Worker B: 7:00 to 19:00
- Worker C: 12:00 to 23:59

then the smallest complete safety committee would consist of just worker B since the second shift overlaps both the first and the third.

1. Consider the following algorithm to produce a complete safety committee containing as few workers as possible.

Each worker's shift is an interval. Suppose that I_x is the interval with the earliest finishing time (we will define time 0 as being midnight). We find all intervals that contain I_x 's finishing time, and choose the interval I_y with the latest finishing time among those as part of the safety committee. We then delete all intervals that overlap I_y (including I_y), and repeat the operation until no intervals are left.

Give an example that shows that this algorithm is **not** optimal.

2. Describe briefly in words an efficient algorithm that takes the schedule of n shifts and produces a complete safety committee containing as few workers as possible.

1.2 Ol' McDonald is safer, he hi he hi ho (Individual)

As in the group stage: The manager in a McDonald's comes to you with the following problem: she is in charge of a group of n workers. Each worker works one shift each day of the week (i.e., a particular worker works from the same start time to the same finish time each day, though two different workers may work different times). There can be multiple shifts happening at once. Assume that no shift starts before midnight and ends after midnight and that shifts overlap even if they just "meet" at their start or end times.

The manager is trying to choose a subset of these n workers to form a safety committee that she can meet with once a week. She considers such a committee to be complete if, for every worker X not on the committee, X's shift overlaps at least partially the shift of some worker who is on the committee. In this way, each worker's obedience to safety protocols can be observed by at least one person who is serving on the committee.

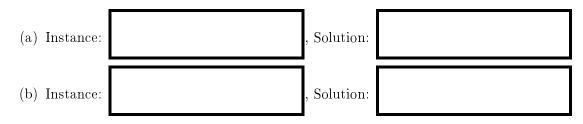
Example: Suppose that n = 3 and the shifts are

- Worker A: 0:00 to 10:00
- Worker B: 7:00 to 19:00
- Worker C: 12:00 to 23:59

then the smallest complete safety committee would consist of just worker B since the second shift overlaps both the first and the third.

Moving on to new questions:

1. Describe two trivial instances for this problem and their solutions.



2. One correct and efficient algorithm to solve this problem sorts the endpoints of the shifts in increasing order and then iterates through them making greedy choices (left unspecified here) of whether to include people in the safety committee. We will call a shift *s* covered if the worker for shift *s* is in the safety committee, or isn't but *s* overlaps a shift whose worker **is** in the safety committee. A shift that is not covered will be called *uncovered*.

Which data structure(s) will this algorithm need to maintain as it is iterating through the endpoints of the shifts? Assume that p is the shift endpoint currently being processed. Fill in the box next to **all** statements that apply.

A list of all covered shifts.

A list of all uncovered shifts.

 \Box The shifts S_1, \ldots, S_j whose workers were added to the safety committee so far.

- \square A list of the covered shifts whose starting time is after S_j 's finishing time, but no later than p.
- \square A list of the uncovered shifts whose starting time is after S_j 's finishing time, but no later than p.
- 3. Which of the following is the best **lower bound** on the running time of the algorithm mentioned in the previous question?

 $\begin{array}{c} \bigcirc \ \Omega(\log n) \\ \bigcirc \ \Omega(n) \\ \bigcirc \ \Omega(n \log n) \\ \bigcirc \ \Omega(n \log n) \\ \bigcirc \ \Omega(n^2) \\ \bigcirc \ \Omega(n^2 \log n) \\ \bigcirc \ \Omega(n^2 \log n) \\ \bigcirc \ \text{None of these is a lower bound.} \end{array}$

4. Let S_1, \ldots, S_k be the set of shifts returned by the greedy algorithm. Assume without loss of generality that S_1, \ldots, S_k is sorted by finishing times. Also, let T_1, \ldots, T_m be the shifts in a smallest safety committee, sorted by increasing finishing time. We will denote the starting and finishing times of shift x by s(x) and f(x) respectively.

Which of the following statements would it be useful to prove as part of a proof that the greedy algorithm is optimal? Fill in the box next to **all** statements that apply.

 $\begin{array}{c} \bigsqcup \ \mathbf{k} < \mathbf{m} \\ \ \bigsqcup \ \mathbf{k} \leq m \\ \ \bigsqcup \ \mathbf{k} \geq m \\ \ \bigsqcup \ \mathbf{For} \ j = 1, \dots, k, \ s(S_j) \geq s(T_j). \\ \ \bigsqcup \ \mathbf{For} \ j = 1, \dots, k, \ f(S_j) \geq f(T_j). \\ \ \bigsqcup \ \mathbf{For} \ j = 1, \dots, m, \ s(S_j) \geq s(T_j). \\ \ \bigsqcup \ \mathbf{For} \ j = 1, \dots, m, \ f(S_j) \geq f(T_j). \end{array}$

2 **Recurrences resolve runtimes**

2.1 Recurrences resolve runtimes (Group)

A recurrence relation is a function definition that expresses the value of the function for an argument n in terms of its values for arguments smaller than n. Here is a well known example of the recurrence relation for Fibonacci numbers:

$$F(n) = \begin{cases} F(n-1) + F(n-2) & \text{if } n \ge 2\\ 1 & \text{if } n = 1\\ 0 & \text{if } n = 0 \end{cases}$$

Knowing that F(0) = 0 and F(1) = 1, we can then compute

- F(2) = F(1) + F(0) = 1 + 0 = 1
- F(3) = F(2) + F(1) = 1 + 1 = 2
- F(4) = F(3) + F(2) = 2 + 1 = 3
- F(5) = F(4) + F(3) = 3 + 2 = 5

etc.

When we are given a recursive algorithm, one (and, frequently, the only) way to determine its worst case running time is by writing a recurrence relation for it. In this week's reading quiz, you learn how to solve the recurrence relations that arise from a specific class of algorithms called *divide and conquer* algorithms. This tutorial quiz gives you practice obtaining these recurrence relations. Take **mergesort** as an example, with annotations about its runtime on the right:

Suppose that the sublist A[left ... right] of A that we are sorting contains n elements (that is, right – left +1 = n). The **if** guides us to build two cases for the recurrence. In the case when n = 1, nothing happens and so the running time is constant. Otherwise, the function performs two recursive calls and a call to **merge**. The time taken by each recursive call is described by the recurrence relation itself. The first recursive call is on $\lceil n/2 \rceil$ elements, and the second on $\lfloor n/2 \rfloor$ elements for $T(\lceil n/2 \rceil)$ and $T(\lfloor n/2 \rfloor)$ time. Finally we know **merge** runs in $\Theta(n)$ time. We thus get the recurrence

$$T(n) = \begin{cases} T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + \Theta(n) & \text{if } n \ge 2\\ \Theta(1) & \text{if } n \le 1 \end{cases}$$

Write recurrence relation describing the worst case running time of the following functions:

```
1. define hanoi(n, from, to, using):
    if n > 0:
        hanoi(n-1, from, using, to)
        print ("Move disc from peg " + from + " to peg " + to)
        hanoi(n-1, using, to, from)
```

2. Assume that the call ElmerJ(A, first, n, x) runs in $\Theta(n \log n)$ time.

```
define BugsB(A, first, n):
    if n < 3:
        return A[first] - 1
    x = BinarySearch(A, first, n, BugsB(A, first + floor(n/3), floor(n/3)))
    ElmerJ(A, first, n, x)
    return BugsB(A, first, floor(n/6)) * BugsB(A, first + floor(n/2), floor(n/4))</pre>
```

3. In the following example, assume that A is a matrix (a 2-dimensional array), that Get(A, i, j, k, l) returns the square submatrix of A containing rows i through j and columns k through l, that Put(C, i, j, k, l, X) stores X into the part of C containing containing rows i through j and columns k through l. Both operations run in time proportional to the number of elements in the submatrix. You may also assume that Add(X,Y) returns the sum of matrices X and Y, and that Sub(X,Y) returns the difference of matrices X and Y. Add and Sub run in time proportional to the number of elements in the matrices X and Y.

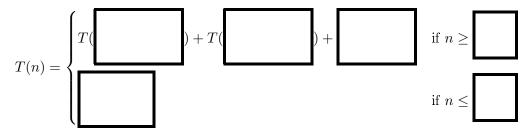
```
// Assumption: n is a power of 2.
define strassen(A, B, n):
   C = new matrix(n, n)
   if n = 1:
       C[0, 0] = A[0, 0] * B[0, 0]
   else:
       A11 = Get(A, O,
                         n/2-1, 0, n/2-1); A12 = Get(A, 0, n/2-1, n/2, n-1)
       A21 = Get(A, n/2, n-1, 0, n/2-1); A22 = Get(A, n/2, n-1, )
                                                                     n/2, n-1)
       B11 = Get(B, 0,
                         n/2-1, 0, n/2-1); B12 = Get(B, 0, n/2-1, n/2, n-1)
       B21 = Get(B, n/2, n-1,
                                0, n/2-1); B22 = Get(B, n/2, n-1),
                                                                     n/2, n-1)
       P1 = strassen(Add(A11, A22), Add(B11, B22), n/2)
       P2 = strassen(Add(A21, A22), B11, n/2)
       P3 = strassen(A11, Sub(B12, B22), n/2)
       P4 = strassen(A22, Sub(B21, B11), n/2)
       P5 = strassen(Add(A11, A12), B22, n/2)
       P6 = strassen(Sub(A21, A11), Add(B11, B12), n/2)
       P7 = strassen(Sub(A12, A22), Add(B21, B22), n/2)
       Put(C, 0,
                                 n/2-1, Add(Add(P1, P7), Sub(P4, P5)))
                   n/2-1, 0,
       Put(C, 0,
                   n/2-1, n/2,
                                 n-1,
                                        Add(P3, P5))
       Put(C, n/2, n-1,
                                 n/2-1, Add(P2, P4))
                          0,
       Put(C, n/2, n-1,
                        n/2, n-1, Add(Add(P1, P6), Sub(P3, P2)))
  endif
  return C
```

2.2 Recurrences resolve runtimes (Individual 1)

Fill in the blanks in the recurrence relations describing the worst case running time of each of the following functions. You may ignore floors and ceiling when you write the terms of the recurrence relations.

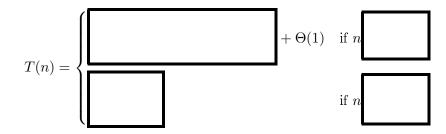
1. Assume that the call merge(A, left, mid, right) runs in $\Theta(\text{right} - \text{left})$ time.

```
define skewedsort(A, left, right):
    if left + 1 < right:
        mid = floor ((2*left + right)/3)
        skewedsort(A, left, mid)
        skewedsort(A, mid+1, right)
        merge(A, left, mid, right)
    else if left < right:
        if A[left] > A[right]:
            swap(A[left], A[right])
```



2. This one is a bit tricky, so be careful.

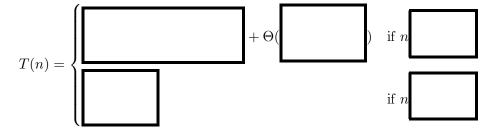
```
define pointless(A, n)
    // A is an array with n elements
    if n > 1:
        return pointless(A, n-1) * pointless(A, n-2)
    else:
        return A[n-1]
```



Don't miss the question on the next page!

3. Assume that the call ALittleBitOfThis(A, size) runs in $\Theta(1)$ time, and the call ALittleBitOfThat(A, size, i) (correction: size should have been n) runs in $\Theta(\log n)$ time,

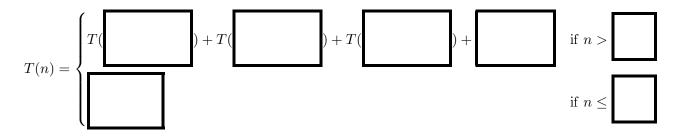
```
define recurrence(A, size, n)
    // A is an array with size elements, and 0 <= n < size.
    if n = 0:
        ALittleBitOfThis(A, size)
else:
        i = n
        while i > 0:
            i = floor(i/2)
            ALittleBitOfThat(A, n, i)
            recurrence(A, size, n-1)
            ALittleBitOfThat(A, n, i)
```



2.3 Recurrences resolve runtimes (Individual 2)

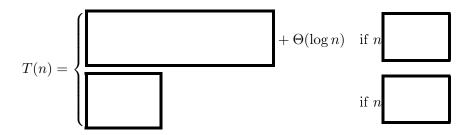
Fill in the blanks in the recurrence relations describing the worst case running time of each of the following functions. You may ignore floors and ceiling when you write the terms of the recurrence relations.

1. Assume that a call to LinearTimeAlgorithm(A, first, n1, second, n2) runs in O(n1+n2) time.



2. This one is a bit tricky, so be careful.

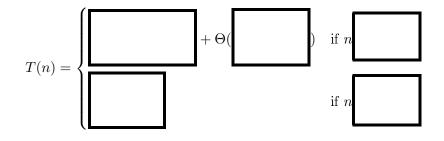
```
define pointless(A, n, item)
    // A is an array with n elements; item is an array element
    if n > 0
        x = binarysearch(A, n, item)
        pos = pointless(A, n-1, pointless(A, n-2, x))
    else:
        return 1
```



Don't miss the question on the next page!

3. Assume that the call ALittleBitOfThis(A, size) runs in $\Theta(1)$ time, and the call ALittleBitOfThat(A, n, i) runs in $\Theta(\sqrt{n})$ time,

```
define recurrence(A, size, n)
   // A is an array with size elements, and 0 <= n < size.
   if n <= 1:
        ALittleBitOfThis(A, size)
   else:
        for i = n-1 downto 0:
            ALittleBitOfThat(A, size, i)
            recurrence(A, size, n-2)
            ALittleBitOfThat(A, size, i)</pre>
```



3 Playing the Blame Game

A distributed computing system composed of n nodes is responsible for ensuring its own integrity against attempts to subvert the network. To accomplish this, nodes in the system can assess each others' integrity, which they always do in pairs. A node in such a pair with its integrity intact will correctly assess the node it is paired with to report either "intact" or "subverted". However, a node that has been subverted may freely report "intact" or "subverted" regardless of the other node's status.

The goal is for an outside authority to determine which nodes are intact and which are subverted. If n/2 or more nodes have been subverted, then the authority cannot necessarily determine which nodes are intact using any strategy based on this kind of pairing. However, if more than n/2 nodes are intact, it is possible to confidently determine which are which.

Throughout this problem, we assume that **more** than n/2 of the nodes are intact. Further, we let one "round" of pairings be any number of pairings overall as long as each node participates in at most one pairing. (I.e., a round is a matching that may not be perfect.)

1. Imagine that nodes a and b have been paired. No matter what report we receive, both nodes could have been subverted because subverted nodes may respond arbitrarily to a pairing.

Fill in the circle next to **all other** possible situations corresponding to a given report from the nodes:

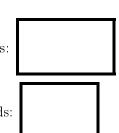
a reports b is	b reports a is	Could be:		
intact	intact	\Box both intact	$\Box a$ intact, b subverted	$\Box a$ subverted, b intact
intact	subverted	both intact	$\Box a$ intact, b subverted	$\Box a$ subverted, b intact
subverted	intact	both intact	$\Box a$ intact, b subverted	$\Box a$ subverted, b intact
subverted	subverted	both intact	$\Box a$ intact, b subverted	$\Box a$ subverted, b intact

(Reminder: both subverted is **always** a possibility.)

2. Imagine that we've found **one** node that is definitely intact and that, having done so, we now only make a pairing when at least one node in the pair is known to be intact. Give good asymptotic lowerand upper-bounds on the number of rounds required to determine for every other node whether it is intact or subverted.

Asymptotic lower-bound on rounds:

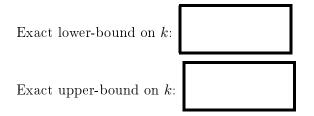
Asymptotic upper-bound on rounds:



(We can now have as a goal finding one intact node.)

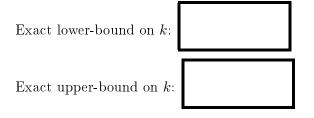
3. Now, imagine that we are searching for one definitely-intact node among a set of nodes of which **more** than half are intact. We find a way to discard from the search k > 0 nodes with the guarantee that at least half of the discarded nodes are subverted (i.e., at least as many discards are subverted as intact).

Give exact lower- and upper-bounds on k in terms of n such that more than half of the new search space is guaranteed to be intact.



4. Again imagine that we are searching for one definitely-intact node among a set of nodes of which **more** than half are intact. This time, we find a way to discard from the search k > 0 nodes that may be any mixture of intact and subverted nodes such that there are k nodes **still in the search space** in exactly the same mixture of intact and subverted nodes. (There are no other nodes discarded, but there may be others kept in the search space.)

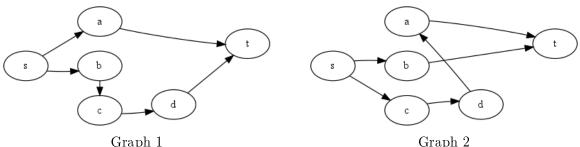
Give exact lower- and upper-bounds on k in terms of n such that more than half of the new search space is guaranteed to be intact. (Note that we do care about $\lfloor \text{floors} \rfloor$ and $\lceil \text{ceilings} \rceil$ in your answer, though substantial partial credit is available without them.) Consider only the case where $n \equiv 1 \mod 4$, i.e., n-1 is exactly divisible by 4.



4 Mixed Nets

You're working on the routing for an anonymization service called a "mixnet" in which a network of computers pass messages through a sequence of handoffs from one source computer to another target computer.

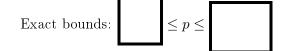
To represent this, you have a weakly-connected, directed acyclic graph (DAG) G = (V, E) composed of designated source and target vertices $s, t \in V$ and a set of p > 0 simple paths (along which p messages pass) each of which starts at s, ends at t, and includes at least one vertex in between s and t. The paths are also vertex disjoint besides s and t (i.e., no two paths share any other vertex). There are no other vertices or edges in the graph. For example, here are two different graphs both over the same set of vertices and both with p = 2:



1. Let n = |V|. Give exact lower- and upper-bounds on p in terms of n.

Exact bounds: $\leq p \leq$

2. Let m = |E|. Give exact lower- and upper-bounds on p in terms of m.



- 3. In a valid instance of this problem, the p paths are vertex disjoint besides the shared start at s and end at t. Are they therefore also *edge-disjoint* (i.e., no two paths include the same edge)?
 - ALWAYS
 SOMETIMES
 NEVER
- 4. In a valid instance of this problem, is there **any** path from s to t?
 - \bigcirc ALWAYS

 \bigcirc Sometimes

 \bigcirc NEVER

On the next page, you'll work on this variation to the problem:

Your mixnet actually involves a single set of computers (with a designated start and target computer) and two entirely separate sets of paths among those computers, as with the two sample graphs on the previous page. At some point as each message passes along its path among the first set of paths, it switches to using one of the second set of paths instead (never switching back).

Specifically, you have an overall graph made up of two subgraphs like those specified in the previous part, where one subgraph's vertices is an exact copy of the other's, i.e., $G_1 = (V_1, E_1)$ and $G_2 = (V_2, E_2)$, where a vertex $v_1 \in V_1$ if and only if there is a vertex $v_2 \in V_2$. (s_1 and t_1 are the start and target vertices in G_1 and their corresponding vertices s_2 and t_2 are the start and target in G_2 .) Each subgraph is based on its own set of p paths, but p is the same for both. There is **also** a directed edge (v_1, v_2) for each vertex $v_1 \in V_1$ except s_1 and t_1 leading from G_1 to G_2 . (There is no edge from s_1 to s_2 or t_1 to t_2 .)

So, for the example on the previous page, the overall graph would include both Graph 1 (with each node subscripted like a_1) and Graph 2 (with each node subscripted like a_2) plus 4 more edges: $(a_1, a_2), (b_1, b_2), (c_1, c_2), (d_1, d_2)$.

Your goal is to find p vertex-disjoint paths in the overall graph that start at s_1 in G_1 and end at t_2 in G_2 . Thus, no two paths visit the exact same vertex (besides s_1 and t_2). Additionally, you want to ensure that no two paths even visit the same vertex in *different* subgraphs (i.e., no path contains v_1 if any other path contains v_2).

- 5. In a valid instance of this problem, is there **any** path from s_1 to t_2 ?
 - ALWAYS○ SOMETIMES○ NEVER
- 6. In a valid instance of this problem, is there **any** path from t_1 to t_2 ?
 - Always
 Sometimes
 Never
- 7. Fill in the box next to **all** the edges used in a solution to the problem based on the examples on the previous page:
 - $\Box (a_1, a_2)$ $\Box (b_1, b_2)$ $\Box (c_1, c_2)$ $\Box (d_1, d_2)$
- 8. Fill in the box next to **all** the edges in a solution to the problem based on the examples on the previous page that has the right number of well-formed paths but violates one of the other constraints on solutions:
 - $\Box (a_1, a_2)$ $\Box (b_1, b_2)$ $\Box (c_1, c_2)$ $\Box (d_1, d_2)$
- 9. Given a valid instance of this problem, does it necessarily have a valid solution (i.e., a solution with p vertex-disjoint paths where no two paths even visit the same vertex in *different* subgraphs)?
 - \bigcirc ALWAYS
 - \bigcirc Sometimes
 - \bigcirc NEVER

5 Cover Charge

In the "minimum edge cover" problem, the input is a simple, undirected, connected graph G = (V, E) with $|V| \ge 2$, and the output is the smallest possible set E' such that $E' \subseteq E$ and for all vertices $v \in V$, there is an edge $\{v, u\}$ (which is the same as $\{u, v\}$) in E'. That is, every vertex in the graph is the endpoint of some edge in E'.

Consider the following greedy algorithm for this problem:

```
Sort the vertices in increasing order by degree
Mark all vertices as uncovered
Let the cover E' be empty.
While there are vertices remaining, pick the next one v:
    If v is uncovered:
        If there is any neighbour u of v that is uncovered:
            Find the uncovered neighbour u of v with lowest degree
            Add {u, v} to E' and mark u and v as covered
        Else:
            Pick an arbitrary edge {u, v}, add it to E', and mark v as covered
```

- 1. Does this greedy algorithm produce an edge cover (whether or not it is minimal)?
 - ALWAYS
 SOMETIMES
 NEVER
- 2. Does this greedy algorithm produce a minimal edge cover?
 - ALWAYS
 - \bigcirc Sometimes
 - \bigcirc NEVER
- 3. Assuming that the graph is represented as an adjacency list and that we can determine the degree of a vertex in constant time, give a good big-O bound on the runtime of this greedy algorithm in terms of n = |V| and m = |E|.

Big-O Bound:

A maximum matching in a graph G = (V, E) is the largest set E'' such that $E'' \subseteq E$ and there are no three vertices $v_1, v_2, v_3 \in V$ such that $\{v_1, v_2\}$ and $\{v_1, v_3\}$ are in E''. That is: E'' "marries off" as many vertices as possible without having any one vertex "married" to two or more vertices.

In this part, assume you have a graph G = (V, E) and a maximum matching for the graph E''.

- 4. Under what conditions is E'' itself a minimal edge cover? Fill in the circle next to the **best** answer.
 - $\bigcirc E''$ is a perfect matching
 - $\bigcirc E''$ is a stable matching

$$\bigcirc \left\{ |E''| = \frac{|E|}{2} \right\}$$

 \bigcirc none of these guarantees E'' is a minimal edge cover

- 5. Let E'' be a maximum matching in G = (V, E), $v \in V$ be a vertex that is **not** covered by E'', and $\{u, v\} \in E$ (i.e., there's an edge from u to v). Is u covered by E''?
 - \bigcirc ALWAYS
 - \bigcirc sometimes
 - \bigcirc NEVER