# CSC 236 HW 4

Check in period: 7/20 - 7/25

## About

Please see the guide to hw, and guide to check ins for information and tips for the HW and the check ins!

These homeworks are on lectures and tutorials 1-8 inclusive.

## 1 Task Scheduling Part 2

This question extends some ideas from HW3. It might be a good idea to review that problem! I have made an optional companion notebook for this problem which you can access here. It contains the prerequisite relationships for the CSC courses.

Let  $\prec$  be a valid prerequisites relation<sup>1</sup> over a finite non-empty set of tasks Tasks. For any two tasks  $a, b \in \text{Tasks}$ ,  $a \prec b$  if and only if a is a prerequisite for b.

In HW3 we showed, by induction, that there exists an ordering of all the tasks in Tasks that respects ≺. In this problem we'll prove the same thing **algorithmically**. I.e. to show that the ordering exists, we'll demonstrate an algorithm that finds the ordering!

In this problem, we'll view the prerequisite relation  $\prec$  over Tasks as a graph. Specifically, the vertices will be the tasks and there will be a directed edge (a,b) if and only if a is a direct prerequisite of b. I.e.  $a \prec b$  and there does not exists a different task c such that  $a \prec c$ , and  $c \prec b$ . Formally, define  $G_{\text{Tasks},\prec} = (\text{Tasks}, E)$  where  $E = \{(a,b) : (a \prec b) \land \neg \exists c. (a \prec c \land c \prec b)\}$ 

Consider the following implementations of isMinimal and findMinimal:

#### **Algorithm 1:** isMinimal(a, G = (V, E))

```
1 for b \in V do

2 | if (b, a) \in E then

3 | return False

4 | end

5 end

6 return True
```

### **Algorithm 2:** findMinimal(G = (V, E))

```
\begin{array}{c|c} \mathbf{1} & \mathbf{for} \ \underline{a \in V} \ \mathbf{do} \\ \mathbf{2} & \mathbf{if} \ \underline{\mathbf{isMinimal}}(a,V,E) \ \mathbf{then} \\ \mathbf{3} & \mathbf{return} \ \underline{a} \\ \mathbf{4} & \mathbf{end} \\ \mathbf{5} & \mathbf{end} \end{array}
```

We propose Algorithm 3, FindOrdering1, as an algorithm to find an ordering of Tasks that respects  $\prec$ .

<sup>&</sup>lt;sup>1</sup>Defined in HW3

#### **Algorithm 3:** FindOrdering1(G = (V, E))

```
1 ordering = [];
2 n = len(V);
3 while len(ordering) < n do
4 | t = findMinimal(V, E);
5 | ordering.append(t);
6 | V.remove(t);
7 end
8 return ordering
```

**Question 1a.** Prove that FindOrdering1 is correct. I.e. Let Tasks be any non-empty finite set of tasks and  $\prec$  be any valid prerequisites relation over Tasks. Prove that FindOrdering1 on input  $G_{\text{Tasks},\prec}$  returns an ordering of all the tasks in Tasks that respects  $\prec$ .

You may assume findMinimal and isMinimal are correct.

Define the following loop invariant.

- P(i). At the end of the *i*th iteration, the following hold:
  - a.) ordering<sub>i</sub> =  $[o_1, ..., o_i]$  such that for any j, k with j < k,  $o_k \not\prec o_j$ . I.e. the ordering is so far valid and has length i.
  - b.) For all t in ordering, and  $t' \in V_i$ ,  $t' \not\prec t$ . Nothing in  $V_i$  is a prerequisite for anything in ordering,
  - c.)  $V_i$  and ordering, form a partition of the tasks. I.e., every task is in exactly one of V or ordering.

**Initialization.** The loop invariant holds at the start of the for loop since ordering<sub>0</sub> = [], and  $V_0$  contains all the tasks.

**Maintenance.** Let  $i \in \mathbb{N}$ , and assume P(i), we'll show P(i+1).

By P(i).a, P(i).c, and the fact that the loop condition passed, we have that  $V_i$  is non-empty and finite. By HW3, there is a minimal element, and by the assumption that findMinimal, is correct, t is a minimal element in  $V_i$ . It is appended to ordering and removed from V. We'll now show the loop invariant holds.

We have ordering<sub>i+1</sub> = ordering<sub>i</sub> + [t],  $V_{i+1} = V \setminus \{t\}$ .

- a.) P(i + 1).a holds because P(i).b implies that t is not a prerequisites of anything in ordering.
- b.) Since t is minimal in  $V_i$ , and P(i).b, P(i+1).b.

c.) P(i+1).c holds from the fact that we simply moved one task from V to ordering Thus, the loop invariant holds at the end of iteration i.

**Termination.** Part a.) of the Loop Invariant implies that the loop will terminate after the nth iteration, at which point, P(n).a, and P(n).c imply that ordering contains an ordering of all the tasks in Tasks that respects  $\prec$ .

Here's the story so far.

- · HW3: There exists an ordering
- · Previous problem: We can **find** an ordering

The next chapter has to do with efficiency - how efficiently can we find the ordering?

Let  $T_1(n, m)$  be the runtime of FindOrdering1 on a graph G = (V, E) where |V| = n, and |E| = m. Assume V and E are given as sets, and that set operations are constant time.

**Question 1b.** Prove that the worst case runtime of FindOrdering1 is  $O(n^3)$ 

isMinimal takes time O(n) in the worst case since it iterates over all the vertices. FindMinimal takes time  $O(n^2)$  in the worst case since there may be just one minimal element, and if we get unlucky with the iteration order and it was the last element, we needed to run isMinimal (n-1) times.

Then FindOrdering1 runs FindMinimal n times. Thus, in the worst case FindOrdering takes time  $O(n^3)$ .

**Question 1c.** Find a non-empty finite set of tasks, Tasks, and a valid prerequisite relation  $\prec$  over Tasks. Such that, FindOrdering1 run on  $G_{\text{Tasks},\prec}$  takes time  $\Omega(n^3)$ . You can assume that the order of iteration over the sets is bad (works in your favor to prove the lower bound).

Let Tasks = [n], and  $\prec$  be <. Note that there is only ever 1 minimal element. Let  $|V_i| = n - i + 1 = k_i$ , in the worst case, FindMinimal tests the minimal element last, and so we do at least  $k_i(k_i - 1) + k_i = \Omega(k_i^2)$  work. In the first n/2 iterations, we do  $\Omega(n^2) + \Omega((n - 1)^2) + ... + \Omega((n/2)^2)$  work, which is at least  $n/2 \cdot \Omega((n/2)^2) = \Omega(n^3)$ , as required.

Algorithm 4 is another proposed algorithm for the same task.

Note that numPrereqs and adjacent can be computed by iterating over all edges once and, and minimal can be computed by iterating over all the vertices. In the algorithm we use Python dictionary and list comprehension syntax.

#### **Algorithm 4:** FindOrdering2(G = (V, E))

```
1 ordering = [];
2 numPrereqs = \{b : |\{(a, b) \in E\}|\};
adjacent = \{a : [b : (a, b) \in E]\};
4 minimal = \{a : numPrereqs[a] == 0\};
5 for i=0,...,n-1 do
      a = minimal.pop();
                                      // retrieves and removes some element
      ordering.append(a);
7
      for b in adjacent[a] do
8
         numPrereqs[b] -= 1;
9
         if numPrereqs[b] == 0 then
10
             minimal.add(b)
11
         end
12
      end
13
14 end
15 return ordering
```

**Question 1d.** Prove that this algorithm is correct. You may assume the inner for loop does what you want it to do - you could prove the inner for loop works as well but that would make the proof extra long.

We'll show the following loop invariant

P(i): At the end of the ith iteration, Let  $G_i = (V_i, E_i)$  be the graph with G with the vertices already in ordering $_i$  removed. I.e.  $V_i = V \setminus \text{ordering}_i$ , and  $E_i = \{(u, v) : u, v \in V'\}$ 

- a.) numPrereqs<sub>i</sub> is a mapping such that for all  $v \in V_i$ , numPrereqs[v] is the number of tasks  $u \in V_i$  such that  $(u, v) \in E_i$ .
- b.) minimal<sub>i</sub> contains all the vertices  $v \in V_i$  with numPrereqs<sub>i</sub>[v] = 0.
- c.) ordering, is a list of i elements that respects  $\prec$ .
- d.) For any  $v \in \text{ordering}_i$ ,  $u \in V_i$ ,  $u \not\prec v$ . Nothing in  $V_i$  is a prereq for anything in ordering.
- e.) Every task v is either in ordering, minimal, or has numPrereqs $_i(v) > 0$ .

**Initialization.** numPrereqs<sub>0</sub>, minimal<sub>0</sub> are assumed to be correct, thus a., b., and e. hold. ordering<sub>0</sub> = [] so c. and d. also hold.

**Maintenance.** Let  $i \in \mathbb{N}$ , and assume P(i), we'll show P(i+1).

By HW3 and P(i).b, minimal<sub>i</sub> is non-empty so we get a task a.

We have ordering<sub>i+1</sub> = ordering<sub>i+1</sub> = P(i). By P(i).d, P(i).d, P(i).d is not a prereq for anything in ordering<sub>i</sub> and thus loop invariant P(i).

Since a was in minimal, P(i).b implies numPrereqs(a) = 0, thus, P(i+1).d holds.

Next, we decrement numPrereqs[b] for every  $b \in V'$  for which a was a prerequisite, and update minimal if any vertex now has 0 prerequisites. This step ensures a.) and b.), and e.) hold.

Thus, P(k+1).

**Termination.** After iteration n, the loop invariant P(n).c, and P(n).e implies that ordering is a list of n elements that respects  $\prec$ , which is what we wanted.

**Question 1e.** Prove that the worst case runtime of FindOrdering2 is O(n + m).

numPrereqs, and adjacent are computed by iterating over the edges once, which take O(m) time, minimal is then computed by iterating through all the vertices which takes O(n) time.

Let k be the sum of all the values in numPrereqs. Note that before the first for loop, k=m since each edge contributes one prerequisite. Every time we reach line 9, k is decremented by 1. Since, numPrereqs[v] is always at least 0 for each  $v, k \geq 0$ , so we reach lines 9 through 12 at most m times. Lines 6 and 7 are executed n times in the loop.

In total, the algorithm takes time at most O(n + m).

Since m is always at most  $n^2$ , FindOrdering2 is strictly better than FindOrdering1 in the worst case!

## 2 Majority

Let l be any list of length n. The majority element of a list is an element that occurs > n/2 times. Note that not all lists have majority elements.

The following are two algorithms that return the majority element in a list if one exists. If one does not exists, it returns None implicitly.

Recall that it requires  $\Theta(\log(n))$  bits of space to represent the natural number n.

checkMajority(l, m) returns True iff m occurs in l > n/2 times. The implementation keeps a counter and iterates over the list. In the worst case, the runtime is O(n) and the algorithm requires  $O(\log(n))$  space (to store the counter, which has value at most n/2+1.).

#### **Algorithm 5:** MajorityNaive

```
Input: a list l

1 n = len(l);

2 counts = defaultdict(int);

3 for \underline{x} in l do

4 | counts[x] += 1;

5 | if \underline{counts[x] > n/2} then

6 | \underline{return x};

7 | end

8 end
```

#### **Algorithm 6:** Majority2

```
Input: a list l
_1 m = None;
i = 0;
\mathbf{3} for \mathbf{x} in l do
      if i = 0 then
          m = x; i = 1;
 5
       else if m = x then
 6
          i += 1;
 7
       else
          i -= 1;
       end
10
11 end
12 if checkMajority(l, m) then
       return m
14 end
```

**Question 2a.** Let n be the length of the list.

Show that MajorityNaive requires at least  $\Omega(n)$  space in the worst case.

Hint. Create an example list that makes the variable counts very big!

Consider the list l = [1, 2, 3..., n]. Since the keys to the count dictionary are the unique elements in the list and each of the elements is distinct, by the end of the for loop, count is a dictionary with n keys and thus requires  $\Omega(n)$  space.

**Question 2b.** Trace Majority2 on input l = [2, 1, 3, 1, 1]. Report the values of m and i at the state of every iteration.

$$m_0 = ext{None}, & i_0 = 0 \\ m_1 = 2, & i_1 = 1 \\ m_2 = 2, & i_2 = 0 \\ m_3 = 3, & i_3 = 1 \\ m_4 = 3, & i_4 = 0 \\ m_5 = 1, & i_5 = 1$$

The return value is  $m_5 = 1$ .

**Question 2c.** Let l be any list with a majority element. Prove that by the end of the for loop, the variable m contains the majority element. You should use a well-selected loop invariant.

Let  $\alpha$  be the majority element. Let b(k) be the number non- $\alpha$  elements in l[k:], and a(k) be the number of  $\alpha$ s in l[k:].

We'll show the following loop invariant.

P(j): At the end of the jth iteration i is non-negative and

- a.) Either  $m = \alpha$  and b(j) a(j) < i.
- b.) Or  $m \neq \alpha$  and a(j) b(j) > i.

**Initialization.**  $m_0 \neq \alpha \ i_0 = 0$ , and we have a(0) - b(0) > 0 since  $\alpha$  is the majority element.

**Maintenance.** Let  $j \in \mathbb{N}$  and suppose P(j). We'll show that P(j+1) holds.

There are several cases.

Case  $m_j = \alpha, x = \alpha, i_j = 0$ .

$$b(j+1) - a(j+1) = b(j) - (a(j) - 1)$$

$$< i_j + 1$$

$$= i_{j+1}$$

Case  $m_j = \alpha, x = \alpha, i_j > 0$ . The calculation is the same as above.

Case  $m_i = \alpha, x \neq \alpha, i_i > 0$ .

$$b(j+1) - a(j+1) = (b(j) - 1) - a_j$$

$$< i_j - 1$$

$$= i_{j+1}$$

Case  $m_j = \alpha, x \neq \alpha, i_j = 0$ . Then we claim P(j+1).b. We have  $m_{j+1} \neq \alpha$ , and

$$\begin{aligned} a(j+1) - b(j+1) &= a(j) - (b(j) - 1) \\ &= a(j) - b(j) + 1 \\ &> i_j + 1 \\ &= i_{j+1} \end{aligned} \qquad (b(j) - a(j) < i_j)$$

Case  $m_j \neq \alpha, x = \alpha, i_j = 0$ . We claim that P(j+1).a

$$b(j+1) - a(j+1) = b(j) - (a_j - 1)$$

$$< i_j + 1$$

$$< i_{j+1}$$

Case  $m_j \neq \alpha, x = \alpha, i_j > 0$ .

$$a(j+1) - b(j+1) = a(j) - 1 - b(j)$$
  
>  $i_j - 1$   
>  $i_{j+1}$ 

Case  $m_j \neq \alpha, x \neq \alpha, i_j = 0$ .

$$a(j+1) - b(j+1) = a(j) - (b(j) - 1)$$
  
>  $i_j + 1$   
>  $i_{j+1}$ 

Case  $m_j \neq \alpha, x \neq \alpha, i_j > 0$ . The calculation is the same as the above.

Thus, in every case, the loop invariant holds.

**Termination.** The loop invariant holds at the end of the nth iteration. Note that b.) from the loop invariant can not hold because a(n) - b(n) = 0 and  $i_n \ge 0$ . Thus, we a.) holds and we have  $m_n = \alpha$  as required.

**Question 2d.** Show that the worst-case runtime of Majority2 is O(n)

The for loop runs n times, and every operation in the for loop is constant time. checkMajority also runs time O(n) time. Thus, the overall runtime of the algorithm is O(n).

**Question 2e.** Show that the worst-case space usage of Majority2 is  $O(\log(n))$ 

In the for loop, we only ever keep track of a single element and a single counter which has value at most n (hence requiring  $O(\log(n))$  space to store). Furthermore, checkMajority requires  $O(\log(n))$  space, thus we never use more than  $O(\log(n))$  space.

## 3 Modular Exponentiation

In lecture we studied algorithms for multiplication. In this problem, we'll explore another common mathematical operation - exponentiation! To make things a little more interesting/manageable we'll study **modular exponentiation**.

Given  $b, e, p \in \mathbb{N}$ , the goal is to compute  $b^e \mod p$ . Here are several proposed algorithms.

For the each of the algorithms, the precondition is that  $b, e, p \in \mathbb{N}$ , with  $0 \le b < p$ . The postcondition is that the function returns  $b^e \mod p$ 

#### **Algorithm 7:** ExpNaive

```
Input: b, e, p \in \mathbb{N}

1 r = 1;

2 for i = 0...e - 1 do

3 r = r \cdot b

4 end

5 return r \mod p
```

#### **Algorithm 8:** ExpModFirst

```
Input: b, e, p \in \mathbb{N}

1 r = 1;

2 for i = 0...e - 1 do

3 | r = (r \cdot b) \mod p

4 end

5 return \underline{r}
```

#### Algorithm 9: ExpRec

```
Input: b, e, p \in \mathbb{N}

1 if e = 0 then

2 | return 1 \mod p;

3 else if e is even then

4 | return ExpRec(b^2 \mod p, e/2, p) \mod p;

5 else e is odd

6 | return \underline{b \cdot ExpRec(b^2 \mod p, (e-1)/2, p) \mod p};

7 end
```

The next algorithm (see next page) uses the binary expansion of e, i.e. Write  $e = \sum_{j=0}^{k-1} x_j 2^j$  for some  $x_0, ..., x_{k-1} \in \{0, 1\}$ 

#### Algorithm 10: ExpRepeatedSquaring

```
Input: b, e, p \in \mathbb{N}

1 x_0, ..., x_{k-1} is the binary expansion of e.;

2 r = 1;

3 a = b;

4 for i = 0...k - 1 do

5 | if x_i = 1 then

6 | r = (r \cdot a) \mod p;

7 end

8 | a = (a \cdot a) \mod p;

9 end

10 return r
```

You may assume the following lemma

```
Lemma. Let a,b,p\in\mathbb{N}. Then (a\mod p)\cdot (b\mod p)\mod p=(a\cdot b)\mod p
```

For any number  $a \in \mathbb{N}$ , assume the number of digits for a is  $\Theta(\log(a))$ .

Assume it takes time  $\Theta(n^2)$  to multiply/divide two numbers both with at most n digits. In particular, for any two numbers  $x, y \in \mathbb{N}$ , if  $x \leq p$  and  $y \leq p$ , then it takes time  $\Theta(\log^2(p))$  to compute  $x \cdot y$ , and  $x \mod y$ . Note  $\log^2(p)$  is shorthand for  $(\log(p))^2$ .

**Question 3a.** Show that worst case asymptotic runtime of ExpNaive in terms of e and p is at least  $\Omega(e^3 \log^2(p))$ .

Pick b=p-1. On the e/2+1th iteration, you're computing  $b^{e/2} \cdot b$  which takes time  $\Omega(\log(b^{e/2})^2) = \Omega(e^2\log^2(b))$ . Summing over the last e/2 iterations (which each take at least the time of iteration e/2+1), we get that the total runtime of the algorithm is at least  $e/2 \cdot \Omega(e^2\log^2(b)) = \Omega(e^3\log^2(b)) = \Omega(e^3\log^2(b)) = \Omega(e^3\log^2(b))$ .

**Question 3b.** Show that the worst case asymptotic runtime of ExpModFirst in terms of e, and p is at most  $O(e \log^2(p))$ .

Since  $b \le p$  and  $r \le p$ , the multiplication and modding at each iteration requires time at most  $O(\log^2(p))$ . There are e iterations, so the overall runtime is at most  $O(e \log^2(p))$ .

**Question 3c.** Prove ExpRec is correct.

Define P(e): For all  $b, p \in \mathbb{N}$  with  $0 \le b < p$ , ExpRec returns  $b^e \mod p$  on input b, e, p. By induction on e. **Base Case.** Note that when e = 0, we return 1, and  $b^0 = 1$ , so we are good.

**Inductive Step.** Let  $k \in \mathbb{N}$  with k > 0, and suppose P(0), ..., P(k-1). We'll show P(k). If k is even, then we return ExpRec called with the e parameter being k/2. Since k is even and k > 0,  $k \ge 2$ , and  $0 \le k/2 \le k-1$ . Thus, the IH applies. Therefore, we return

$$(b^2 \mod p)^{k/2} \mod p = b^{2k/2} \mod p = b^k \mod p$$

as expected. Note that we used the Lemma here.

The calculation is similar in the other case. If k is odd, then  $0 \le (k-1)/2 < k$ , so again, the IH applies. In this case, we return

$$b \cdot (b^2 \mod p)^{(k-1)/2} \mod p = b \cdot b^{2(k-1)/2} \mod p = b^k \mod p$$

as required.

**Question 3d.** Informally explain why the running time of ExpRec is  $O(\log(e) \log^2(p))$ .

The recurrence for the runtime is  $T(e) = T(e/2) + \log^2(p)$ , since T is being called on  $\lfloor e/2 \rfloor$ , and we do at most 2 multiplications of numbers at most p (which takes  $\log^2(p)$  time. Since the non-recursive work doesn't depend on e, the total work is just  $\log^2(p)$  times the height of the recursion tree which is  $\log(e)$ . Hence we have  $T(e) = O(\log(e)\log^2(p))$ .

**Question 3e.** Prove ExpRepeatedSquaring is correct by defining a loop invariant and proving initialization/maintenance/termination.

Hint.

$$b^e = b^{\sum_{j=0}^{k-1} x_j 2^j}$$

It might be useful to define the empty product to be equal to 1 or the empty sum to be equal to 0.

Define the empty sum to be equal to zero. That is if you have  $\sum_{i=a}^{b} ...$ , and b < a, then the sum is equal to 0.

Let P(i) be the following loop invariant.

At the end of the *i*th iteration,

$$\cdot \ r_i = b^{\sum_{j=0}^{i-1} x_j 2^j} \mod p$$

$$a_i = b^{2^i} \mod p$$

**Initialization.**  $r_0 = 1 = b^0$ , and  $a_i = b = b^1$ . Thus, the loop invariant holds before the start of the loop.

**Maintenance** Let  $m \in \mathbb{N}$  and suppose P(m). The variables get updated as follows.

If  $x_m = 0$ , then r is unchanged

$$r_{m+1} = r_m$$

$$= b^{\sum_{j=0}^{m-1} x_j 2^j} \mod p$$

$$= b^{\sum_{j=0}^m x_j 2^j} \mod p$$

If  $x_m = 1$ 

$$\begin{aligned} r_{m+1} &= r_m \cdot a_m \mod p \\ &= \left( b^{\sum_{j=0}^{m-1} x_j 2^j} \mod p \right) \cdot \left( b^{2^m} \mod p \right) \mod p \\ &= b^{\sum_{j=0}^m x_j 2^j} \mod p \end{aligned}$$

Finally,

$$a_{m+1} = a_m \cdot a_m \mod p$$

$$= (b^{2^m} \mod p) \cdot (b^{2^m} \mod p) \mod p$$

$$= b^{2^{m+1}} \mod p$$

as required.

**Termination.** At the end of the kth iteration the loop invariant implies that the return value  $r_k$  is equal to

$$b^{\sum_{j=0}^{k-1} x_j 2^j} \mod p = b^e \mod p$$

as required.

**Question 3f.** Find and prove the worst case asymptotic runtime of ExpRepeated-Squaring in terms of e and p. Just prove Big-O. Assume everything before the for loop takes constant time.

 $k = O(\log(e))$ . There are k iterations, and each iteration does two multiplications of numbers at most p. In total, this takes time  $O(\log(e)\log^2(p))$ .