

# COL351

## Assignment 3

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## 1 Question 1

Question 1

**Question.** *Question*

*Solution.*

**Claim 1.1.**

*Proof.*



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**Algorithm 1** DP solution for partitioning

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1:

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## 2 Question 2

### Question 2

**Question.** The total net force on particle  $j$ , by Coulomb's Law, is equal to

$$F_j = \sum_{i < j} \frac{Cq_i q_j}{(j-i)^2} - \sum_{i > j} \frac{Cq_i q_j}{(j-i)^2} \quad (1)$$

Design an algorithm that computes all the forces  $F_j$  in  $O(n \log n)$  time.

**Solution.** We will use polynomial multiplication to solve this question. Consider the polynomials:

$$\begin{aligned} A(x) &= (0, q_1, q_2, \dots, q_n) \\ B(x) &= \left( -\frac{1}{(n-1)^2}, -\frac{1}{(n-2)^2}, \dots, -\frac{1}{1^2}, 0, \frac{1}{1^2}, \dots, \frac{1}{(n-2)^2}, \frac{1}{(n-1)^2} \right) \end{aligned} \quad (2)$$

In the above representation, only the coefficients of  $A(x), B(x)$  are shown. The degrees of  $A(x)$  and  $B(x)$  are  $n$  and  $2n-2$  respectively. Now, in the product  $P(x) = A(x) \cdot B(x)$ , consider the coefficient of  $x^{j+n-1}$ . To visualise this, we will write the polynomials as:

$$\begin{aligned} A(x) &= q_n x^n + \dots + q_{j+1} x^{j+1} + q_j x^j + q_{j-1} x^{j-1} + \dots + q_1 x^1 + 0x^0 \\ B(x) &= \dots + -\frac{1}{(n-j)^2} x^{j-1} + \dots + -\frac{1}{1^2} x^{n-2} + 0x^{n-1} + \frac{1}{1^2} x^n + \dots + \frac{1}{(j-1)^2} x^{j+n-2} \\ &\quad + \dots \end{aligned} \quad (3)$$

Multiplication of corresponding terms gives terms with power of  $x$  as  $j+n-1$ , and thus formally, the coefficient of  $x^{j+n-1}$  can be written as:

$$P(x)[j+n-1] = \sum_{k=1}^{n-j} q_{j+k} \cdot -\frac{1}{k^2} + 0 + \sum_{k=1}^{j-1} q_{j-k} \cdot \frac{1}{k^2} \quad (4)$$

Where  $P(x)[p]$  denotes the coefficient of  $x^p$  in  $P(x)$ .

Equation 4 can be rewritten as:

$$\begin{aligned} P(x)[j+n-1] &= \sum_{i=j+k, k=1}^{n-j} q_i \cdot -\frac{1}{(j-1)^2} + \sum_{i=j-k, k=1}^{j-1} q_i \cdot \frac{1}{(j-1)^2} \\ &= -\sum_{i=j+1}^n \frac{q_i}{(j-1)^2} + \sum_{i=1}^{j-1} \frac{q_i}{(j-1)^2} \\ &= \sum_{i < j} \frac{q_i}{(j-1)^2} - \sum_{i > j} \frac{q_i}{(j-1)^2} \\ &= \frac{F_j}{Cq_j} \\ \implies F_j &= P(x)[j+n-1] \times Cq_j \end{aligned} \quad (5)$$

Therefore, we have derived an alternate method for computing  $F_j$ . Since this involves computing product of polynomials, we can perform the polynomial product in  $O(n \log n)$  since both  $A(x), B(x)$  are polynomials of degree  $O(n)$ . Once  $C(x)$  has been computed, we can then compute  $F_j$  in  $O(1)$  for each  $j$  by dividing the corresponding coefficient with  $Cq_j$ . The exact algorithm is given as:

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**Algorithm 2** Computing  $F_j$  for  $j \in \{1, 2, \dots, n\}$

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```

1: procedure COMPUTE FORCES( $(q, n)$ )
2:    $A \leftarrow q$ 
3:    $B \leftarrow \left[ -\frac{1}{(n-1)^2}, -\frac{1}{(n-2)^2}, \dots, -\frac{1}{1^2}, 0, \frac{1}{1^2}, \dots, \frac{1}{(n-2)^2}, \frac{1}{(n-1)^2} \right]$ 
4:    $P \leftarrow \text{multiply}(A, B) \triangleright$  multiply  $A(x)$  and  $B(x)$  using FFT “divide and conquer” algo
5:    $F \leftarrow C[n : 2n - 1] \triangleright$  taking subarray corresponding to coefficients of  $x^{j+n-1}$ 
6:   for  $i \in [1, n]$  do  $\triangleright$  1-indexed array
7:      $F[i] \leftarrow F[i] \times Cq[i]$ 
8:   end for
9:   return  $F$ 
10: end procedure

```

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**Proof of Correctness:** The proof of correctness of the *FFT Algorithm* has been discussed in the lectures. The correctness of lines 5 – 8 has been proved from Equation 5.

**Time Complexity:** All operations except the *FFT Algorithm* are  $O(n)$  operations. The complexity of *FFT Algorithm* has been shown to be  $O(d \log d)$  where  $d$  is the degree of the polynomial. Since the degrees of  $A(x), B(x)$  are  $O(n)$ , the *FFT Algorithm* can be computed in  $O(n \log n)$  time.

Therefore, all forces  $F_j$  can be computed in  $O(n \log n)$  time. This completes the design of the algorithm along with proof of correctness and time complexity.  $\square$

### 3 Question 3

3.1 3.1

### Question 3(a)

**Question.** Prove that the graph  $H = (V, E_H)$  can be computed from  $G$  in  $O(n^\omega)$  time, where  $\omega$  is the exponent of matrix-multiplication.

*Proof.* Enumerate the vertices  $V$  in  $G$  as  $\{1, 2, \dots, |V| = n\}$  and let  $A_G$  be the adjacency matrix of  $G$ . Consider the term  $A_G^2$ . From **Lemma 1** of Lecture 22, we know that  $A_G^2$  is positive only if there exists a walk of length *exactly* 2. Therefore, we have the following claim:

**Claim 3.1.** The adjacency list for graph  $H$  is given as  $A_G + A_G^2 \succ 0$ , where  $A_G$  is adjacency matrix of  $G$ .

*Proof.* From definition of  $H$ , we have that edges in graph  $H$  consists of all edges of graph  $G$  and end points of walks of length 2. Therefore,  $E_H$  has all edges of walks of length 1 and 2. In other words,  $(A_H)_{ij}$  is positive only if there exists a walk of length 1 or 2 between nodes  $i, j$ . This can be formally written as:

$$\begin{aligned} (A_H)_{ij} &= (A_G)_{ij} > 0 \vee (A_G^2)_{ij} > 0 \\ &= (A_G)_{ij} + (A_G^2)_{ij} > 0 \\ \implies A_H &= A_G + A_G^2 \succ 0 \end{aligned} \tag{6}$$

□

Therefore, the algorithm for computing  $A_H$  is:

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#### Algorithm 3 Computing $H$

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```

1: procedure COMPUTEH( $G$ )
2:    $A_G \leftarrow \text{adjacency}(G)$ 
3:    $A_H \leftarrow A_G + A_G^2$ 
4:    $n \leftarrow |V_G|$ 
5:   for  $i, j \in [1, n] \times [1, n]$  do
6:     if  $(A_H)_{ij} > 0$  then
7:        $(A_H)_{ij} \leftarrow 1$ 
8:     else
9:        $(A_H)_{ij} \leftarrow 0$ 
10:    end if
11:  end for
12:  return  $\text{graph}(A_H)$ 
13: end procedure

```

---

**Time Complexity** Computing  $A_G^2$  will take  $O(n^\omega)$  time. All other steps take  $O(n^2)$  time. We know that  $\omega \geq 2$ . Therefore, the overall time complexity of the algorithm will be  $O(n^\omega)$ . Therefore, we have proposed an algorithm which computes the graph  $H$  via its adjacency matrix in  $O(n^\omega)$  time. This completes the proof. □

### 3.2 3.2

#### Question 3(b)

**Question.** Argue that for any  $x, y \in V$ ,  $D_H(x, y) = \left\lceil \frac{D_G(x, y)}{2} \right\rceil$

*Solution.* We will prove the given statement by first showing that there exists a path of length  $\left\lceil \frac{D_G(x, y)}{2} \right\rceil$  for each  $x, y$  in  $H$ . We will then prove that we cannot have a shorter path length in  $H$ .

*Note:* For this and subsequent parts, we call edges which are directly in  $G$  as edges of *type 1* and the other edges as edges of *type 2*.

**Claim 3.2.** For each  $x, y \in V$ , there exists a path of length  $\left\lceil \frac{D_G(x, y)}{2} \right\rceil$  in graph  $H$ , corresponding to the shortest path in  $G$ .

*Proof.* Let the shortest path between  $x, y$  in  $G$  be given as:

$$\begin{aligned} P_G(x, y) &= \{x, a_1, a_2, \dots, a_k, y\} \\ \implies D_G(x, y) &= k + 1 \end{aligned} \tag{7}$$

We now have two cases, when  $k$  is odd and when  $k$  is even. For the case when  $k$  is odd, we have:

$$\begin{aligned} P_H(x, y) &= \{x, a_2, a_4, \dots, a_{k-1}, y\} \\ \implies \text{length}(P_H(x, y)) &= \frac{k-1}{2} + 1 \\ &= \frac{k+1}{2} \\ &= \left\lceil \frac{D_G(x, y)}{2} \right\rceil \end{aligned} \tag{8}$$

When  $k$  is even, we have:

$$\begin{aligned} P_H(x, y) &= \{x, a_2, a_4, \dots, a_k, y\} \text{ } ((a_k, y) \text{ is the only edge of type 1}) \\ \implies \text{length}(P_H(x, y)) &= \frac{k}{2} + 1 \\ &= \frac{(k+1) + 1}{2} \\ &= \left\lceil \frac{D_G(x, y)}{2} \right\rceil \end{aligned} \tag{9}$$

Therefore we have shown the correctness of the claim for both cases of  $k$ . □

We will now show that there cannot exist a path between  $x, y$  of shorter length in  $H$ .

**Claim 3.3.** The shortest distance between  $x, y$  is given exactly as  $\left\lceil \frac{D_G(x, y)}{2} \right\rceil$

*Proof.* We will prove the claim using contradiction. Assume that there exists a shorter path  $Q_H(x, y)$ :

$$\begin{aligned} Q_H(x, y) &= \{x, b_1, b_2, \dots, b_m, y\} \\ \implies \text{length}(Q_H(x, y)) &= m + 1 < \left\lceil \frac{D_G(x, y)}{2} \right\rceil, \text{ from assumption} \end{aligned} \quad (10)$$

Consider the edges in  $G$  corresponding to this path  $Q_H(x, y)$ :

$$\begin{aligned} Q_G(x, y) &= \{x, c_1, b_1, c_2, b_2, \dots, c_m, b_m, c_{m+1}, y\}, c_i \text{ may be the same as } b_i \\ \implies \text{length}(Q_G(x, y)) &\leq 2m + 2 < 2 \left\lceil \frac{D_G(x, y)}{2} \right\rceil \\ \implies \text{length}(Q_G(x, y)) &< \begin{cases} D_G(x, y) + 1, & D_G(x, y) \text{ is odd} \\ D_G(x, y), & D_G(x, y) \text{ is even} \end{cases} \end{aligned} \quad (11)$$

We know that  $D_G(x, y)$  is the shortest path in  $G$  between vertices  $x, y$ . Therefore, we have that such a path cannot exist if  $D_G(x, y)$  is even and in the case when  $D_G(x, y)$  is odd, we notice that the inequality in  $\text{length}(Q_G(x, y))$  has an even number  $(2m + 2)$  in the RHS. Therefore, the equality cannot hold in this case as well. Thus, we have arrived at a contradiction on the length of shortest  $x, y$  path in  $G$ . Therefore,  $\left\lceil \frac{D_G(x, y)}{2} \right\rceil$  is the shortest path in  $H$ .  $\square$

Thus, from Claim 3.2 and Claim 3.3 we have shown that  $D_H(x, y) = \left\lceil \frac{D_G(x, y)}{2} \right\rceil$ .

Hence, proved.  $\square$

### 3.3 3.3

#### Question 1

**Question.** *Question*

*Solution.*

**Claim 3.4.**

*Proof.*

$\square$

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**Algorithm 4** DP solution for partitioning

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1:

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$\square$



### 3.4 3.4

#### Question 1

**Question.** *Question*

*Proof.* We will first propose the algorithm and then prove its correctness and time complexity.

---

**Algorithm 5** Computing  $D_G$  from  $D_H$

---

```

1: procedure COMPUTEDG( $G, D_H$ )
2:    $M \leftarrow D_H \times \text{adjacency}(G)$ 
3:    $D_G \leftarrow \text{init}()$ 
4:   for  $x \in V$  do
5:     for  $y \in V$  do
6:       if  $M(x, y) \geq \deg(G, y) \cdot D_H(x, y)$  then
7:          $D_G(x, y) \leftarrow 2D_H(x, y)$ 
8:       else
9:          $D_G(x, y) \leftarrow 2D_H(x, y) - 1$ 
10:      end if
11:    end for
12:  end for
13:  return  $D_G$ 
14: end procedure

```

---

Algorithm 5 computes the matrix  $D_G$  using the idea proven in Question 3.4. Therefore, from the proof given in Question 3.4, we can compute  $D_G$ .

**Time Complexity** Line 2 in Algorithm 5 takes  $O(n^\omega)$  time. The nested for loop takes  $O(n^2)$  time since each iteration takes  $O(1)$  time. Therefore the total running time is  $O(n^\omega)$  ( $\omega > 2$ ).

Therefore, we have used the proof of Question 3.4 to arrive at an  $O(n^\omega)$  solution for computing  $D_G$ . This completes the proof.  $\square$

### 3.5 3.5

#### Question 1

**Question.** *Question*

*Solution.* We propose the following algorithm for computing all-pairs-distances:

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**Algorithm 6** Computing all-pairs-distances

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```
1: procedure ALLPAIRDISTANCES( $G$ )
2:    $A_G \leftarrow \text{adjacency}(G)$ 
3:    $H \leftarrow \text{COMPUTE}H(G)$ 
4:   if  $H = G$  then
5:      $D_G \leftarrow A_G$ 
6:      $D_G \leftarrow$  all off-diagonal zero entries are set to  $\infty$ 
7:     return  $D_G$ 
8:   end if
9:    $D_H \leftarrow \text{ALLPAIRDISTANCES}(H)$ 
10:   $D_G \leftarrow \text{COMPUTEDG}(G, D_H)$ 
11:  return  $D_G$ 
12: end procedure
```

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This is a recursive algorithm that we use to compute the all-pairs-shortest distances. To prove the same, we will prove the correctness of the algorithm using reverse induction on the depth of the recursive calls.

**Base Case** If  $H$  is the same as  $G$ , then each component in  $G$  is fully connected. Therefore, the distance matrix will be the same as the adjacency matrix and the off-diagonal entries that are 0 will be  $\infty$  since there is no path between such vertices.

**Inductive Step** We assume that it is true for depth  $i + 1$ , now consider the call at depth  $i$ . We have already shown the correctness of line 3, 10 in Question 3.1 and Question 3.4 respectively. Additionally from the inductive assumption, we know that  $D_H$  is indeed the distance of graph  $H$ . Therefore, our recursive algorithm is correct.

However, we still have to prove termination. To do the same, we notice that any two vertices that have a path between them have a path of length  $< n$ . Additionally, the distance halves at each step as proved in Question 3.1. Therefore, the algorithm terminates in  $O(\log n)$  calls.

**Time Complexity** As stated above, the number of calls to `AllPairDistances` is  $O(\log n)$ . Each call of the function takes  $O(n^\omega)$  time as shown in Question 3.1 and Question 3.4. Therefore, the total time complexity of the algorithm is  $O(n^\omega \log n)$ .

This completes the algorithm along with proof of correctness and time complexity.  $\square$

## 4 Question 4

### 4.1 4.1

Question 1	
<b>Question.</b> <i>Question</i>	
<hr/>	
<i>Solution.</i>	
<b>Claim 4.1.</b>	
<i>Proof.</i>	<input type="checkbox"/>
<hr/>	
<b>Algorithm 7</b> DP solution for partitioning	
1:	
<hr/>	
	<input type="checkbox"/>

### 4.2 4.2

Question 1	
<b>Question.</b> <i>Question</i>	
<hr/>	
<i>Solution.</i>	
<b>Claim 4.2.</b>	
<i>Proof.</i>	<input type="checkbox"/>
<hr/>	
<b>Algorithm 8</b> DP solution for partitioning	
1:	
<hr/>	
	<input type="checkbox"/>

### 4.3 4.3

Question 1	
<b>Question.</b> <i>Question</i>	
<hr/>	
<i>Solution.</i>	
<b>Claim 4.3.</b>	
<i>Proof.</i>	<input type="checkbox"/>

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**Algorithm 9** DP solution for partitioning

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