

COL351

Assignment 3

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1 Question 1

Question 1

Question. *Question*

Solution.

Claim 1.1.

Proof.



Algorithm 1 DP solution for partitioning

1:



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:

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

```

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

```

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 ω 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:

Algorithm 3 Computing A_H

```

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

Algorithm 4 DP solution for partitioning

1:

\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 3.3. Therefore, from the proof given in Question 3.3, 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 3.3 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.

Claim 3.5.

Proof.

\square

Algorithm 6 DP solution for partitioning

1:



4 Question 4

4.1 4.1

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

4.2 4.2

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

4.3 4.3

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

Algorithm 9 DP solution for partitioning

1:

