Efficient Algorithms and Intractable Problems

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1 Complexity Analysis

Definition 1.1 (Asymptotic Notations).

$$\begin{split} f &= O(g) \approx f(n) \leq c \cdot g(n) \\ f &= o(g) \approx f(n) < c \cdot g(n) \\ f &= \Omega(g) \approx f(n) \geq c \cdot g(n) \\ f &= \omega(g) \approx f(n) > c \cdot g(n) \\ f &= \Theta(g) \approx f(n) = c \cdot g(n) \end{split}$$

Theorem 1.1 (Master Theorem). If $T(n) = aT([n/b]) + O(n^d)$ for some constants a > 0, b > 1, and $d \ge 0$, then

$$T(n) = \begin{cases} \Theta(n^d) & a < b^d \\ \Theta(n^d \log n) & a = b^d \\ \Theta(n^{\log_b a}) & a > b^d \end{cases}$$

2 Polynomial Interpolation

Given a degree n polynomial $A(x) = a_0 + a_1x + a_2x^2 + \cdots + a_nx^n$, the relationship between its values and coefficients can be represented by

$$\begin{bmatrix} A(x_0) \\ A(x_1) \\ \vdots \\ A(x_{n-1}) \end{bmatrix} = \begin{bmatrix} 1 & x_0 & x_0^2 & \cdots & x_0^{n-1} \\ 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ \vdots & \vdots & & & \\ 1 & x_{n-1} & x_{n-1}^2 & \cdots & x_{n-1}^{n-1} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$
 (evaluation)

where the matrix M is a Vandermonde matrix.

2.1 Fast Fourier Transform (FFT)

Definition 2.1 (Discrete Fourier Transform Matrix). For polynomials of degree < n (n is even; polynomials can be 0-padded), the Discrete Fourier Transform can be represented by the matrix

$$M_n(\omega) = \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & \omega & \omega^2 & \cdots & \omega^{n-1} \\ & & \vdots & & \\ 1 & \omega^j & \omega^{2j} & \cdots & \omega^{(n-1)j} \\ & & \vdots & & \\ 1 & \omega^{n-1} & \omega^{2(n-1)} & \cdots & \omega^{(n-1)(n-1)} \end{bmatrix}$$

where $\omega = e^{2\pi i/n}$ is the *n*th root of unity.

Remark. $M_n(\omega)$ is an unitary matrix whose columns forms the Fourier Basis.

Lemma 2.1.
$$M_n^{-1}(\omega) = \frac{1}{n} \overline{M_n(\omega)} = \frac{1}{n} M_n(\omega^{-1})$$

Lemma 2.2.
$$A(x) = A_{even}(x^2) + xA_{odd}(x^2)$$

Remark. If A is evaluated at points $\pm \omega_0, \ldots, \pm \omega_{n/2-1}$, then $A_e(x^2)$ and $A_o(x^2)$ will only need to evaluate half the amount of points $(T(n) = 2T(n/2) + O(n) = O(n \log n))$.

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Algorithm 1: Fast Fourier transform
      Input: A coefficient vector, \vec{a} = \langle a_0, \dots, a_{n-1} \rangle and the nth root of unity, \omega.
      Output: M_n(\omega)\vec{a}
 1 Function FFT(\vec{a}, \omega):
 \mathbf{2}
            if \omega = 1 then
                   return \vec{a}
 3
            else
 4
                  \langle A_e(0), \dots, A_e(n/2-1) \rangle \leftarrow \text{FFT}(\langle a_0, a_2, \dots, a_{n-2} \rangle, \omega^2)
\langle A_o(0), \dots, A_o(n/2-1) \rangle \leftarrow \text{FFT}(\langle a_1, a_3, \dots, a_{n-1} \rangle, \omega^2)
 \mathbf{5}
 6
                   for j := 0 to n/2 - 1 do
 7
                    A(j) \leftarrow A_e(j) + \omega^j A_o(j)A(j+n/2) \leftarrow A_e(j) - \omega^j A_o(j)
  8
 9
            return \langle A(0), \ldots, A(n-1) \rangle
10
```

2.2 Polynomial Multiplication

Algorithm 2: Fast Polynomial Multiplication

Input: Coefficient vectors, a and b, and the nth root of unity, ω .

Output: The coefficient vector of A(x)B(x)

- $1 \hat{\vec{a}} \leftarrow M_n(\omega)\vec{a}$ (FFT)
- $\mathbf{z} \ \hat{\vec{b}} \leftarrow M_n(\omega) \vec{b}$
- **3 for** i = 0 *to* n 1 **do**
- 4 $\hat{c}_i \leftarrow \hat{a}_i \hat{b}_i$
- 5 return $\frac{1}{n}M_n(\omega^{-1})\hat{\vec{c}}$ (inverse matrix)

2.3 Cross-Correlation

Definition 2.2 (Cross-Correlation). $corr(\vec{x}, \vec{y})[k] = \sum x_i y_{i-k}$, which measures similarity.

Algorithm 3: Cross-Correlation

Input: Two signal vectors, \vec{x} and \vec{y} .

Output: $corr(\vec{x}, \vec{y})$

- $1 \ X(t) \leftarrow x_{m-1} + x_{m-2}t + \dots + x_0t^{m-1}$
- 2 $Y(t) \leftarrow y_0 + y_1 t + \dots + y_{n-1} t^{n-1}$
- 3 $Q(t) \leftarrow X(t)Y(t)$ (Fast Polynomial Multiplication)
- 4 return \vec{q}

3 Graphs

Definition 3.1 (Graph). A graph is a pair G = (V, E), typically represented by an adjacency matrix or an adjacency list.

Table 1: Graph representations.

	Space	Connectivity	getNeighbors(u)	DFS Runtime
Adjacency Matrix	$\Theta(V ^2)$	O(1)	$\Theta(V)$	$\Theta(V ^2)$
Adjacency List	$\Theta(V + E)$	$\Theta(degree(u))$	$\Theta(degree(u))$	$\Theta(V + E)$

3.1 Depth-First Search

Algorithm 4: Depth-first search **Input:** V, E of directed graph G. 1 Function DFS(V, E): $n \leftarrow |V|$ 2 $clk \leftarrow 1$ 3 visited \leftarrow boolean[n] 4 preorder, postorder = int[n]5 for $v \in V$ do 6 7 if !visited/v| then EXPLORE(v)8 9 Function EXPLORE(v): $visited[v] \leftarrow True$ 10 $preorder[v] \leftarrow clk++$ 11 for $(v, w) \in E$ do 12if !visited/w| then 13 EXPLORE(w)14 $postorder[v] \leftarrow clk++$ **15**

Remark. Preorder-postorder intervals are either nested or disjoint.

Definition 3.2 (Back Edge). A back edge is an non-tree edge that goes from a descendant vertex to an ancestor vertex.

Remark. postorder[u] \leq postorder[v] iff (u, v) is a back edge.

Remark. G contains a cycle iff it contains a back edge.

3.1.1 Applications of DFS

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Algorithm 5: Topological sort.

Input: A directed cyclic graph G.
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Output: An ordered list of V such that u_i comes before v_i for all $(u_i, v_i) \in E$ (i.e., ordered by decreasing dependency).

1 $post \leftarrow DFS$ -visited vertexes ordered by postorder visits

2 return reverse(post)

Definition 3.3 (Strongly Connected Component). A SCC is a maximal partition of a directed graph in which every vertex is reachable from every other vertex.

u is in sink SCC of graph $G \Leftrightarrow u$ is in source SCC of reverse graph $G \Leftrightarrow \Box$

u is in source SCC if highest postorder number.

3.2 Single-Source Shortest Paths (SSSPs)

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Algorithm 6: Single-Source Shortest Path

Input: A directed graph G and a start vertex S.

Output: Two arrays prev[|V|] (shortest-path predecessor) and dist[|V|] (shortest-path distance).

1 Function BFS (G, S):

| /* Must have uniform edge weights. O(|V| + |E|) runtime. */

2 Function Dijkstra (G, S):

| /* Must have positive edge weights. O(|V| \log |V| + |E|) runtime if implemented using fibonacci heap. */

3 Function Bellman-Ford (G, S):

| /* Can have aribitrary edge weights. */
```

3.3 Minimum Spanning Tree

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Algorithm 7: MST algorithms.

Input: A graph G and a starting vertex v.

Output: The minimum spanning tree T of G.

/* Use the cut property.

*/

Function Prim(G, v):

T \leftarrow \emptyset

H \leftarrow \text{new Fibonacci heap}
```