

## Outline

### Paradigm

1. Break up given instance into considerably smaller ones.
2. Recursively solve those.
3. Combine their solutions into one for the given instance.

### Examples of common pattern

- ▶ Sorting (Mergesort)
- ▶ Counting inversions
- ▶ Finding a closest pair of points in the plane

### Lower bound for sorting

## Sorting

### Problem specification

**Input:** array  $A[1 \dots n]$  of integers with  $n \geq 1$

**Output:**  $\text{Sort}(A)$ , i.e.,  $A$  sorted from smallest to largest

### Mergesort

```
procedure MERGE-SORT( $A$ )
  if  $n = 1$  then
    return  $A$ 
  else
     $m \leftarrow \lfloor n/2 \rfloor$ 
     $L \leftarrow A[1, \dots, m]$ 
     $R \leftarrow A[m + 1, \dots, n]$ 
    return MERGE(MERGE-SORT( $L$ ), MERGE-SORT( $R$ ))
```

## Sorting Lower Bound

### Theorem

Every comparison-based sorting algorithm takes  $\Omega(n \log n)$  comparisons on arrays of length  $n$ .

### Proof

- ▶ Every such algorithm for a given  $n$  can be modeled as a binary decision tree  $T$ .
- ▶ Depth  $d$  of  $T$  is the maximum number of comparisons that  $A$  makes on arrays of length  $n$ .
- ▶ Number  $\ell$  of leaves is at least  $n! \doteq 1 \cdot 2 \cdot \dots \cdot n$ .
- ▶  $\ell \leq 2^d$
- ▶  $d \geq \log(\ell) \geq \log(n!)$
- ▶  $(n/2)^{n/2} \leq n! \leq n^n$  so  $\log(n!) = \Theta(n \log n)$

## Counting Inversions

### Definition

An inversion in an array  $A[1 \dots n]$  is a pair  $(i, j) \in [n] \times [n]$  with  $i < j$  and  $A[i] > A[j]$ .

### Example

$A = [3, 5, 4, 7, 3, 1]$

### Bounds on $\text{Inv}(A)$

Between 0 (sorted) and  $\binom{n}{2}$  (reverse sorted).

### Problem specification

**Input:** array  $A[1 \dots n]$  of integers with  $n \geq 1$

**Output:**  $\text{Inv}(A) \doteq$  number of inversions in  $A$

## Count

### D&C approach

### Counting cross inversions

- ▶ Problem specification  
**Input:** sorted arrays  $L[1 \dots n]$  and  $R[1 \dots m]$  with  $n, m \geq 1$   
**Output:**  $\text{Inv}(LR)$
- ▶ Algorithm running in time  $O(n + m)$

Running time of Count:  $O(n(\log n)^2)$

## Improved Count

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**Input:**  $A[1 \dots n]$ , an array of length  $n \geq 1$

**Output:**  $(\text{Inv}(A), \text{Sort}(A))$

```
1: procedure COUNT-AND-SORT( $A$ )
2:   if  $n = 1$  then
3:     return  $(0, A)$ 
4:   else
5:      $m \leftarrow \lfloor n/2 \rfloor$ 
6:      $(c_L, L) \leftarrow \text{COUNT-AND-SORT}(A[1, \dots, m])$ 
7:      $(c_R, R) \leftarrow \text{COUNT-AND-SORT}(A[m + 1, \dots, n])$ 
8:      $c_{\text{cross}} \leftarrow \text{COUNT-CROSS}(L, R)$ 
9:      $c \leftarrow c_L + c_R + c_{\text{cross}}$ 
10:     $B \leftarrow \text{MERGE}(L, R)$ 
11:    return  $(c_L + c_R + c_{\text{cross}}, B)$ 
```

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Running time:  $O(n \log n)$

## Closest Pair of Points in the Plane

### Problem specification

Input:  $(x_i, y_i) \in \mathbb{R}^2$  for  $i \in [n]$

Output:  $\delta \doteq \min\{\delta_{i,j} \text{ for } i, j \in [n] \text{ with } i \neq j\}$  where  
 $\delta_{i,j} \doteq \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$

### D&C approach

### Closest crossing pair in the plane

## Closest Pair of Points in the Plane

### Pseudocode for recursive case

1. Find  $x^*$ ,  $L$ , and  $R$
2. Recursively compute  $\delta_L$  and  $\delta_R$
3.  $\delta^* \leftarrow \min(\delta_L, \delta_R)$
4.  $M \leftarrow \{i \in [n] \text{ s.t. } x_i \in (x^* - \delta^*, x^* + \delta^*)\}$
5. Sort  $M$  based on  $y$ -coordinate
6.  $\delta_M \leftarrow \min\{\delta_{M[i], M[j]} \text{ for } i < j < i + 12\}$
7. Return  $\min(\delta^*, \delta_M)$

### Correctness

### Running time

- ▶ Using local sorting:  $O(n \log n)$  locally and  $O(n(\log n)^2)$  overall
- ▶ Using presorting:  $O(n)$  locally and  $O(n \log n)$  overall

## Polynomials

### Coefficient representation

$$A(x) = A_0 + A_1x + A_2x^2 + \dots + A_{n-1}x^{n-1} = \sum_{i=0}^{n-1} A_i x^i$$

### Evaluation

- ▶  $A(x) = A_0 + x(A_1 + x(A_2 + \dots + x(A_{n-1}) \dots))$  [Horner]
- ▶  $O(n)$  arithmetic operations

### Sum

- ▶  $A(x) + B(x) = (\sum_{i=0}^{n-1} A_i x^i) + (\sum_{i=0}^{n-1} B_i x^i)$   
 $= \sum_{i=0}^{n-1} (A_i + B_i) x^i$
- ▶  $O(n)$  arithmetic operations

### Product

- ▶  $A(x) \cdot B(x) = (\sum_{i=0}^{n-1} A_i x^i) \times (\sum_{j=0}^{n-1} B_j x^j)$   
 $= \sum_{k=0}^{2(n-1)} C_k x^k$  where  $C_k = \sum_{i=0}^{n-1} A_i B_{k-i}$
- ▶  $O(n^2)$  arithmetic operations trivially

## Alternate Representation of Polynomials

### Point-value representation

- ▶ No two polynomials of degree at most  $n-1$  can agree on  $n$  or more points.
- ▶ For any fixed choice of  $n$  points  $x_0, x_1, \dots, x_{n-1}$ , represent  $A(x) = \sum_{i=0}^{n-1} a_i x^i$  as  $(y_0, y_1, \dots, y_{n-1})$  where  $y_i = A(x_i)$ .

### Sum

- ▶  $O(n)$  arithmetic operations

### Product

- ▶  $O(n)$  arithmetic operations

### Evaluation

- ▶  $A(x) = \sum_{i=0}^{n-1} y_i \cdot \prod_{j \neq i}^{n-1} \frac{x - x_j}{x_i - x_j}$  [Lagrange interpolation]
- ▶  $O(n^2)$  arithmetic operations trivially

## Converting Between Representations

### Relationship

$$\begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ \vdots \\ y_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & x_0 & x_0^2 & \dots & x_0^{n-1} \\ 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n-1} & x_{n-1}^2 & \dots & x_{n-1}^{n-1} \end{bmatrix} \begin{bmatrix} A_0 \\ A_1 \\ A_2 \\ \vdots \\ A_{n-1} \end{bmatrix}$$

### Conversions

- ▶ From coefficient to evaluations:  $O(n^2)$
- ▶ From evaluations to coefficients:  $O(n^3)$
- ▶ Can improve both to  $O(n \log n)$  by picking evaluations points cleverly: Fast Fourier Transform (FFT)
- ▶ Implies that polynomial multiplication in the coefficient representation can be done in time  $O(n \log n)$ .

## Point Set for Efficient Simultaneous Polynomial Evaluation

### Splitting the coefficient array

- ▶  $A(x) = \sum_{i=0}^{n-1} A_i x^i$   
 $= \sum_{i \text{ even}} A_i x^i + \sum_{i \text{ odd}} A_i x^i$   
 $= A_{\text{even}}(x^2) + x \cdot A_{\text{odd}}(x^2)$
- ▶  $A(-x) = A_{\text{even}}(x^2) - x \cdot A_{\text{odd}}(x^2)$

### Divide & Conquer approach

- ▶ Pick  $x_0, \dots, x_{n-1}$  such that  $x_{n/2+i} = -x_i$  for  $i \in [n/2] - 1$ .
- ▶ Evaluation of  $A(x)$  at  $x_0, \dots, x_{n-1}$  reduces to evaluation of  $A_{\text{even}}(x)$  and  $A_{\text{odd}}(x)$  at  $x_0^2, \dots, x_{n/2-1}^2$ , with  $O(n)$  local work.
- ▶ Recursive application requires use of complex numbers. Every complex number  $z \neq 0$  has two distinct square roots:  $\pm\sqrt{z}$ .
- ▶  $x_0, \dots, x_{n-1}$  need to be distinct  $n$ -th roots of some  $c \in \mathbb{C}$  where  $n$  a power of 2. Can pick  $c = 1$ .

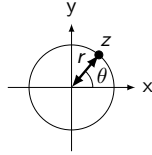
## Complex Numbers

### Definition

$$\mathbb{C} = \mathbb{R}(\sqrt{-1}) = \{x + y\sqrt{-1} : x, y \in \mathbb{R}\}$$

### Representation of $z = x + y\sqrt{-1}$

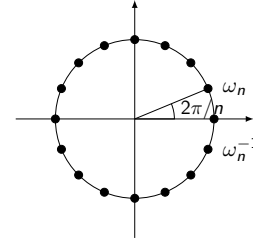
- ▶ Cartesian:  $(x, y)$
- ▶ Polar:  $(r, \theta)$  where  $z = r(\cos(\theta) + \sin(\theta)\sqrt{-1})$



### Arithmetic operations

- ▶ Sum:  $(x_1 + y_1\sqrt{-1}) + (x_2 + y_2\sqrt{-1}) = (x_1 + x_2) + (y_1 + y_2)\sqrt{-1}$
- ▶ Product:  $(x_1 + y_1\sqrt{-1}) \cdot (x_2 + y_2\sqrt{-1}) = (x_1x_2 - y_1y_2) + (x_1y_2 + x_2y_1)\sqrt{-1}$   
 $r = r_1 \cdot r_2$  and  $\theta = \theta_1 + \theta_2$

## Roots of Unity



### Primitive $n$ -th root of unity

- ▶ Definition:  $\omega \in \mathbb{C}$  such that  $\langle \omega \rangle \doteq \{\omega^k : k \in \mathbb{N}\}$  consists of all  $n$ -th roots of unity.
- ▶ Standard:  $\omega_n$
- ▶ If  $\omega$  is primitive  $n$ -th root of unity for even  $n$ , then  $\omega^2$  is primitive  $n/2$ -th root of unity.

## Discrete Fourier Transform

### Definition

Input:  $n \in \mathbb{N}; A_0, \dots, A_{n-1} \in \mathbb{C}$

Output:  $(y_0, \dots, y_{n-1})$  where  $y_i = A(\omega_n^i)$  for  $A(x) = A_0 + a_1x + \dots + A_{n-1}x^{n-1}$

### Algorithm for $n$ a power of 2

- ▶ D&C approach runs in time  $O(n \log n)$  assuming arithmetic over  $\mathbb{C}$  as elementary operations.
- ▶ Known as Fast Fourier Transform (FFT).
- ▶  $\omega_n$  used for specificity in definition of DFT.
- ▶ FFT works for arbitrary primitive  $n$ -th root.
- ▶ Extends to finite fields.

## Fast Fourier Transform

Input:  $n$ : power of 2;  $\omega$ : primitive  $n$ -th root of 1;  $A_0, \dots, A_{n-1} \in \mathbb{C}$

Output:  $(y_0, \dots, y_{n-1})$  where  $y_i = A(\omega^i)$  for  $A(x) = A_0 + A_1x + \dots + A_{n-1}x^{n-1}$

### Pseudocode

```

procedure FFT( $n, \omega, A_0, \dots, A_{n-1}$ )
  if  $n = 1$  then return ( $A_0$ )
  ( $e_0, \dots, e_{n/2-1}$ )  $\leftarrow$  FFT( $n/2, \omega^2, A_0, A_2, \dots, A_{n-2}$ )
  ( $f_0, \dots, f_{n/2-1}$ )  $\leftarrow$  FFT( $n/2, \omega^2, A_1, A_3, \dots, A_{n-1}$ )
   $x \leftarrow 1$ 
  for  $k = 0$  to  $n/2 - 1$  do
     $y_k \leftarrow e_k + x \cdot f_k$ 
     $y_{n/2+k} \leftarrow e_k - x \cdot f_k$ 
     $x \leftarrow x \cdot \omega$ 
  return ( $y_0, \dots, y_{n-1}$ )
    
```

Note: Can be implemented iteratively in-line.

## Inverting the Discrete Fourier Transform

$$\begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ \vdots \\ y_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega_n & \omega_n^2 & \dots & \omega_n^{n-1} \\ 1 & \omega_n^2 & \omega_n^4 & \dots & \omega_n^{2(n-1)} \\ & & & \ddots & \\ 1 & \omega_n^{n-1} & \omega_n^{(n-1)2} & \dots & \omega_n^{(n-1)(n-1)} \end{bmatrix} \begin{bmatrix} A_0 \\ A_1 \\ A_2 \\ \vdots \\ A_{n-1} \end{bmatrix}$$

- ▶ DFT:  $\vec{y} = F(\omega_n) \cdot \vec{A}$
- ▶  $F(\omega_n^{-1}) \cdot F(\omega_n) = n \cdot I$   
Proof:  $(F(\omega_n^{-1}) \cdot F(\omega_n))_{ij} = \sum_{k=0}^{n-1} F(\omega_n^{-1})_{ik} F(\omega_n)_{kj}$   
 $= \sum_{k=0}^{n-1} \omega_n^{(j-i)k}$   
 $= \begin{cases} n & i = j \\ 0 & i \neq j \end{cases}$
- ▶ Inverse DFT:  $\vec{A} = F(\omega_n^{-1}) \cdot \vec{y} / n$

## Applications of Fast Fourier Transform

### Polynomial multiplication in coefficient representation

- ▶ Algorithm on input  $A(x)$  and  $B(x)$  of degree at most  $d$ 
  1.  $\vec{A} \leftarrow \text{FFT}(n, \omega_n, A)$  and  $\vec{B} \leftarrow \text{FFT}(n, \omega_n, B)$  where  $n$  is smallest power of 2 at least  $2d + 1$
  2.  $\vec{C}_i \leftarrow \vec{A}_i \cdot \vec{B}_i$  for  $i = 0, \dots, n-1$
  3. Return  $C = \text{FFT}(n, \omega_n^{-1}, \vec{C}) / n$
- ▶ Running time:  $O(n \log n)$  assuming arithmetic on complex numbers as elementary operations

### Integer multiplication

- ▶ Integer  $a$  equals value of polynomial  $A(x)$  at  $x = 2$ .
- ▶ Compute  $C(x) = A(x) \cdot B(x)$  and output  $c = C(2)$ .
- ▶ Runs in time  $O(n \log n)$  assuming complex arithmetic, and time  $O(n \cdot \log n \cdot \log \log n)$  using bit operations only.
- ▶ Best known:  $O(n \log n)$  bit operations.

## Integer Multiplication

### Problem

**Input:** nonnegative integers  $a$  and  $b$  in binary notation

**Output:** product  $a \times b$  in binary notation

### Grade school algorithm

$$\begin{array}{r} \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \times \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \hline \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \phantom{1} \phantom{0} \\ \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \\ \hline \phantom{\times} \phantom{+} \phantom{1} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{0} \phantom{1} \phantom{0} \phantom{1} \phantom{0} \end{array}$$

## Integer Multiplication

### Improved D&C approach

- ▶  $a \times b = (a_L \times b_L) \cdot 2^n + (a_L \times b_R + a_R \times b_L) \cdot 2^{n/2} + (a_R \times b_R)$
- ▶  $a \times b = (a_L \times b_L) \cdot 2^n + (a_R \times b_R) + ((a_L + a_R) \times (b_L + b_R) - (a_L \times b_L) - (a_R \times b_R)) \cdot 2^{n/2}$

### Running time

- ▶ Recursion tree:  $O\left(\left(\sum_{i=0}^{d-1} \left(\frac{3}{2}\right)^i\right) \cdot n\right)$  where  $d = \log_2(n)$
- ▶ Geometric sum with ratio  $r$ :  $\sum_{i=0}^{d-1} r^i = \frac{r^d - 1}{r - 1} = \Theta(r^d)$
- ▶ For  $r = \frac{3}{2}$ :  $r^d = \left(\frac{3}{2}\right)^d = \frac{3^d}{2^d}$
- ▶ For  $d = \log_2(n)$ 
  - ▶  $2^d = n$
  - ▶  $3^d = 3^{\log_2(n)} = (2^{\log_2(3)})^{\log_2(n)} = (2^{\log_2(3)})^{\log_2(n)} = n^{\log_2(3)} \doteq n^q$
- ▶ Conclusion:  $O\left(\frac{n^q}{n} \cdot n\right) = O(n^q)$  where  $q = \log_2(3) \approx 1.585$

## Splitting

- ▶ **Input:**  $(A, p)$ 
  - ▶ array  $A[1, \dots, n]$  of integers
  - ▶ integer  $p$
- ▶ **Output:**  $(L, R)$ 
  - ▶ array  $L[1, \dots, |L|]$  consisting of all entries of  $A$  less than  $p$
  - ▶ array  $R[1, \dots, |R|]$  consisting of all entries of  $A$  larger than  $p$
- ▶ **Algorithm:**

```
procedure SPLIT( $A, p$ )
   $L, R \leftarrow$  empty lists
  for  $i = 1$  to  $n$  do
    if  $A[i] < p$  then
      append  $A[i]$  to  $L$ 
    else if  $A[i] > p$  then
      append  $A[i]$  to  $R$ 
  return  $L$  and  $R$  as arrays
```

## Selection Schema

```
procedure SELECT( $A, k$ )
  if  $n = 1$  then return  $A[1]$ 
  pick a pivot  $p$  from  $A$ 
   $(L, R) \leftarrow$  SPLIT( $A, p$ )
  if  $k \leq |L|$  then
    return SELECT( $L, k$ )
  else if  $k > n - |R|$  then
    return SELECT( $R, k - (n - |R|)$ )
  else
    return  $p$ 
```

- ▶ Correctness
- ▶ Running time assuming linear-time median as pivot:  $O(n)$

## Approximate Median

### Definition

A  $\rho$ -approximate median of  $A$  is an element  $p$  of  $A$  such that  $\text{Split}(A, p)$  returns  $L$  and  $R$  with  $|L| \leq \rho \cdot |A|$  and  $|R| \leq \rho \cdot |A|$ .

### Construction

1. Break up  $A$  into consecutive segments of length  $w$ .
2. Find the median of each segment.
3.  $A' \leftarrow$  subarray consisting of the segment medians.
4. Return the median of  $A'$ .

### Key claim

The median of  $A'$  is a  $\rho$ -approximate median of  $A$  with  $\rho = \frac{3}{4}$ .

## Linear-Time Selection

### Algorithm

```
procedure FAST-SELECT( $A, k$ )
  if  $n = 1$  then return  $A[1]$ 
   $A' \leftarrow$  array of medians of the  $n' \doteq \lceil \frac{n}{w} \rceil$ 
    consecutive length- $w$  segments of  $A$ 
   $p \leftarrow$  FAST-SELECT( $A', \lceil n'/2 \rceil$ )
   $(L, R) \leftarrow$  SPLIT( $A, p$ )
  if  $k \leq |L|$  then
    return FAST-SELECT( $L, k$ )
  else if  $k > n - |R|$  then
    return FAST-SELECT( $R, k - (n - |R|)$ )
  else
    return  $p$ 
```

Running time:  $O(n)$  for  $w \geq 5$