

\* GCD of  $a, b$ ,  $\gcd(a, b)$

$$a > b \geq 1$$

$S = \{ax + by \mid x, y \in \mathbb{Z}\}$

$$d = \gcd(a, b)$$

Let  $s$  be the min element of  $S$

∴ By WOP:

$$s = ax_1 + by_1$$

$$d \mid a \text{ and } d \mid b \therefore d \mid s \Rightarrow d \leq s$$

$$a = sq + r$$

$$\therefore r = a - sq$$

$$= a(1 - qy_1) + b(-qy_1)$$

# WOP ( $\because r < s$ )  $\Rightarrow d \nmid r$

$$\therefore r = 0 \Rightarrow s \mid a \text{ and } s \mid b$$

$$\therefore s \mid d \therefore s \leq d$$

$$\therefore d = s$$

\* GCD recursion Theorem

$$\frac{\gcd(a, b)}{d} = \underline{\underline{\gcd(b, a \bmod b)}}_{d_1}$$

$$d \mid d_1 \text{ and } d_1 \mid d$$

$$\therefore d = d_1$$

Proof: Use  $a = bq + a \bmod b$

## \* Euclid's GCD Algorithm

```
int gcd (int a, int b)
```

{

```
    if (b == 0) return a;
```

```
    return gcd (b, a % b);
```

}

→ Correctness: GCD recursion theorem

→ Terminates: b always ↓

## \* Tail recursive

k: iterations

$k = f(a, b)$

$$r_1 = b \quad | \quad a \quad (q_1 \quad | \quad a > r_1 \quad | \quad \text{now } r_1 < b)$$

$$r_2 \mid r_1 \quad (q_2 \quad | \quad r_1 > r_2 \quad | \quad \text{now } r_2 < b)$$

$$r_3 \mid r_2 \quad (q_3 \quad | \quad r_2 > r_3 \quad | \quad \text{now } r_3 < b)$$

⋮

$$r_k \mid r_{k-1} \quad (q_k \quad | \quad r_{k-1} > r_k \quad | \quad \text{now } r_k = 0)$$

0

b

b

b

b

b

b

b

b

b

b

\* Induction on  $k$ :

T.P.T.  $a \geq f_{k+2}$ ,  $b \geq f_{k+1}$  (Induction hypo)

$$\begin{array}{cccc} 1 & 1 & 2 & 3 \\ f_1 & f_2 & f_3 & f_4 \end{array}$$

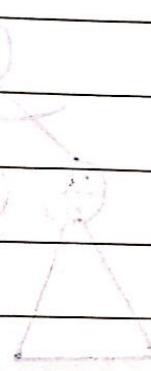
Wynon o hon dkt

\* Induction Base:  $k=1$

$$b \geq 1 \text{ & } a \geq 2 \therefore a > b$$

$$\therefore b \geq f_2 \text{ & } a \geq f_3$$

Hence Base ✓



\* Induction Step:

Let the no. of iterations be  $k-1$   $\therefore r_1 \geq f_{(k-1)+2}$

$$\begin{aligned} & \therefore b \geq f_{(k-1)+2} \\ \Rightarrow b & \geq f_{k+1} \quad \left. \begin{array}{l} \text{since } r_1 \geq f_{(k-1)+2} \\ \therefore b = r_1 \text{ then } r_1 \approx a \end{array} \right\} \end{aligned}$$

$$\text{But } b = r_1 \quad \therefore b + r_2 \geq f_{k+1} + f_k \geq f_{k+2} = (r_1 + r_2) \approx b$$

$$\therefore r_1 \geq f_{k+1} \Rightarrow b \geq f_{k+1}$$

$$\therefore a = b + r_2$$

$$\geq b + r_2 \geq f_{k+1} + f_k \geq f_{k+2} = (r_1 + r_2) \approx b$$

$$(r_1 + r_2) \approx b \Rightarrow b \geq f_{k+1}$$

$$\therefore b \geq f_{k+1} \geq \phi^{k-1}$$

$$\therefore k \text{ is } O(\log b)$$

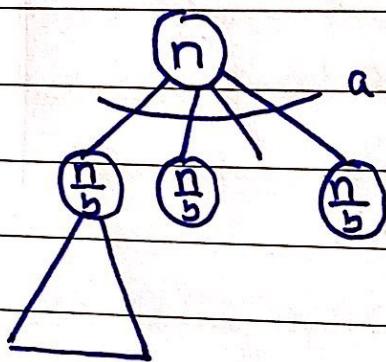
\* Weakly polynomial Time algorithm

\* Strongly polynomial Time algorithm

TC based on no. of inputs like sorting.

## \* Divide and Conquer

### \* Recurrence Tree



\* b and a maybe different what &  
can be done if b is not fixed

$$\begin{aligned} * T(n) &= aT(n/b) + f_1(n) + f_2(n) \quad n > c \\ &= f_3(n) \end{aligned}$$

Divide      Conquer

→ Same subproblem doesn't occur at any other node

$$\begin{aligned} * T(n) &= aT(n/b) + f(n) \quad n > c \\ &= O(1) \end{aligned}$$

→ Master Theorem:

$$\textcircled{1} \text{ If } f(n) = O(n^{\log_b a - \epsilon}) \rightarrow \gg 0$$

$$\text{then } T(n) = \Theta(n^{\log_b a})$$

$$\textcircled{2} \text{ If } f(n) = \Theta(n^{\log_b a})$$

$$\text{then } T(n) = \Theta(n^{\log_b a} \log n)$$

$$\textcircled{3} \text{ If } f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ & } af(n/b) \leq cf(n)$$

$$\text{then } T(n) = \Theta(f(n))$$

\*  $T(n) \leq qT(n/2) + c_1n$  for  $n > 2$  assuming  $q > 1$   
 $\leq c_2 n$  assuming  $q \leq 1$  for  $n \leq 2$  i.e.  $c_2 = 1$   
base case  $T(1) = 1$

① If  $q = 1$

③ follows :  $T(n) = \Theta(n)$

② If  $q = 2$

② follows :  $T(n) = \Theta(n \log n)$

③ If  $q > 2$

① follows :  $T(n) = \Theta(n^{\log_2 q})$

\* Correctness of Divide and Conquer

Induction Base: Leaf node

Induction Hypo: Merge two subtrees to give

correct result.

: base case

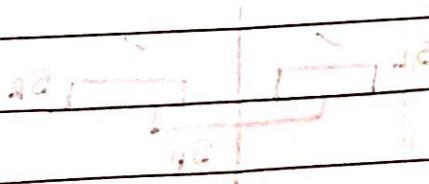
\* Divide and Conquer used to optimise polynomial time algorithm.

: base case

: base case

: base case

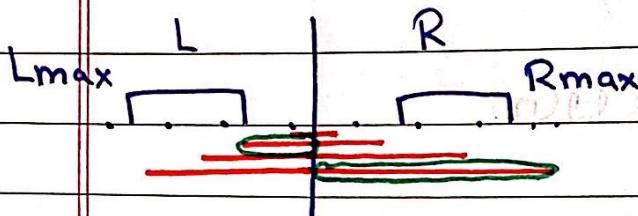
: base case



## ① Maximum Subarray

Naive:  $O(n^3)$ :  $n(n+1)/2$  subarrays

$\sum_{l,r}$  takes  $O(n)$



\* Max Sum across the splitter

$$T(n) = 2T(n/2) + O(1) + O(n) \text{ if } n > 1$$

$= O(1)$  Divide conquer

Prefix

$n=1$

$$\therefore T(n) = \Theta(n \log n)$$

\* Correctness:

Induction on no. of elements

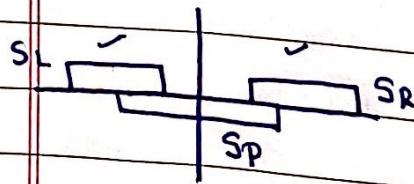
\* Induction Basis:

$n=1$ : Output element itself

\* Induction Hypo:

If no. of elements  $< n$  the algo outputs correctly

\* Induction Step: no. of elements =  $n$



Take max {S\_L, S\_p, S\_R}

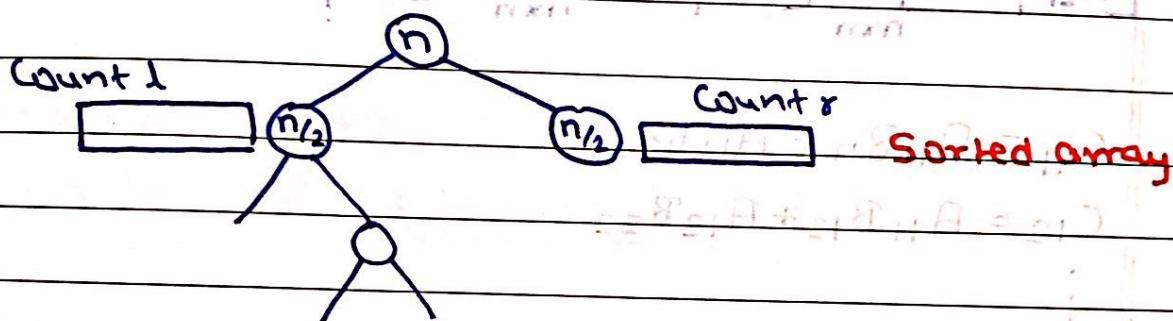
Subarray entirely in left or right or across the splitter

## ② Counting Inversions:

$a_1, a_2, \dots, a_i, \dots, a_j, \dots, a_n$  ( $\leq 10^6$  items)

$a_i > a_j \rightarrow 1$  inversion

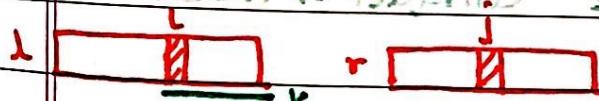
Naive:  $O(n^2)$



Inversion count = count<sub>L</sub> + count<sub>R</sub> + count<sub>m</sub>

inversion is required divide 1 to 10 =

↑  
inversions  
while  
merging



$l[i] > r[j]$  (# Sorted array)

then count<sub>m</sub> += k - l + 1 = 10

$$\rightarrow T(n) = 2T(n/2) + O(1) + O(n) \text{ if } n > 10 \\ = O(n) \text{ if } n = 1$$

$$\therefore T(n) = O(n \log n)$$

$$1000 \cdot 1000 + 1000 + 1000 \cdot 1000 = 1000 \cdot 1000 \cdot 2 = 2000000$$

$$1000 \cdot 1000 = 1000000$$

$$1000000 + 1000000 = 2000000$$

right approach - divide and conquer based

### ③ ★ Matrix Multiplication

Naive  $O(n^3)$

$$\begin{array}{c} A \\ \left[ \begin{array}{cc|cc} A_{11} & A_{12} & B_{11} & B_{12} \\ A_{21} & A_{22} & B_{21} & B_{22} \end{array} \right] \\ n \times n \end{array} \quad \begin{array}{c} B \\ \left[ \begin{array}{cc|cc} B_{11} & B_{12} & C_{11} & C_{12} \\ B_{21} & B_{22} & C_{21} & C_{22} \end{array} \right] \\ n \times n \end{array} = \begin{array}{c} C \\ \left[ \begin{array}{cc|cc} C_{11} & C_{12} & C_{11} & C_{12} \\ C_{21} & C_{22} & C_{21} & C_{22} \end{array} \right] \\ n \times n \end{array}$$

$$C_{11} = A_{11}B_{11} + A_{12}B_{21}$$

$$C_{12} = A_{11}B_{12} + A_{12}B_{22}$$

:

$$\therefore T(n) = 8T(n/2) + O(1) + O(n^2) \quad n > 1$$

$= O(n) \quad n=1$  Divide Conquer # Addition

### ★ Strassen's Matrix Multiplication

$$C_{12} = P_1 + P_2$$

$$C_{21} = P_3 + P_4$$

$$C_{11} = P_5 + P_6 - P_2$$

$$+ P_7 - P_5$$

$$C_{22} = P_5 + P_1 - P_3 - P_7$$

$$T(n) = 7T(n/2) + O(1) + O(n^2) \quad n > 1$$

$= O(n) \quad n=1$

$$\therefore T(n) = O(n^{\log_2 7}) = O(n^{2.81})$$

→ Read Coppersmith-Winograd's Algo.

## ⑥ Karatsuba's Algo:

Multiply two  $n$ -bit positive.

→ RAM model, bitwise logical op, 1. right and left shift

$$X \quad \boxed{x_1 \mid x_0} \quad Y \quad \boxed{y_1 \mid y_0}$$

$$\begin{aligned} XY &= (x_1 \cdot 2^{n/2} + x_0)(y_1 \cdot 2^{n/2} + y_0) \\ &= (\underline{x_1 y_1 \cdot 2^n} + \underline{(x_0 y_1 + x_1 y_0)} 2^{n/2} + x_0 y_0) \end{aligned} \quad \text{①}$$

$$\begin{aligned} T(n) &= 4T(n/2) + O(1) + O(n) \quad n > 1 \\ &= O(1) \quad \text{divide conquer} \quad n = 1 \end{aligned}$$

$$\therefore T(n) = O(n^2)$$

$$\textcircled{1} \equiv (x_1 + x_0)(y_1 + y_0) - x_1 y_1 - x_0 y_0$$

$$\begin{aligned} \therefore T(n) &= 3T(n/2) + O(1) + O(n) \quad n > 1 \\ &= O(1) \quad n = 1 \end{aligned}$$

$$\therefore T(n) = O(n^{\log_2 3}) = O(n^{1.59})$$

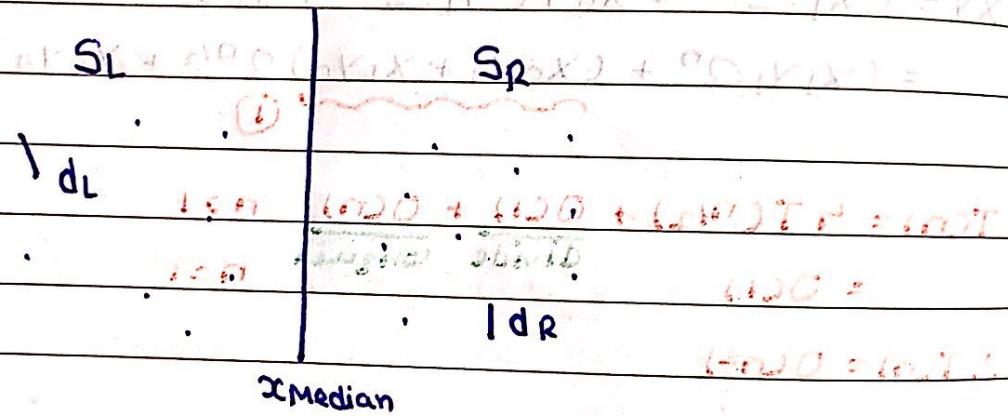
## (5) Closest Pair of Points (CPP)

No two have same x or same y

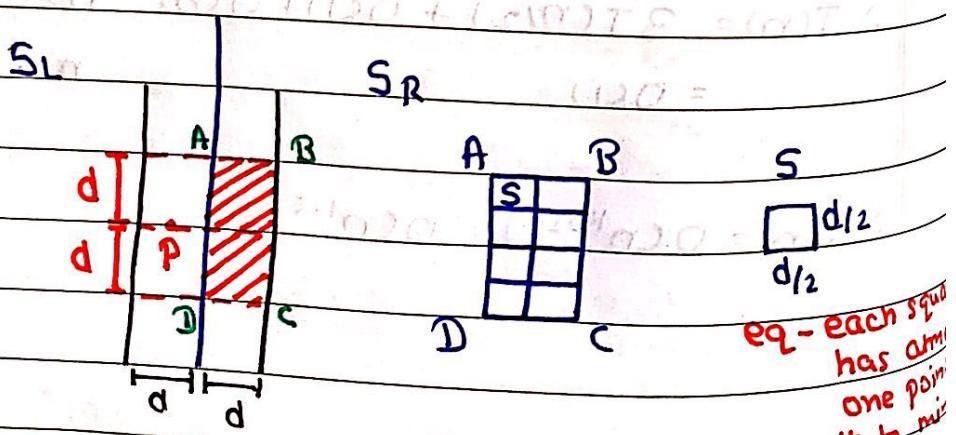
every 2 points have distinct distances

Naive:  $O(n^2)$

Find median x - co-ordinate & divide



$$d = \min(d_L, d_R)$$



eq - each square has only one point if to max of d

- \* We need to check only the points (pairs) with one  $\in S_L$  and other  $\in S_R$

\* Hence for each point  $P$  in the strip  $S_{y''}$  we need to check 4 points above and below it in  $S_y'$ .

$S_x$ : Sorted according to  $x$ -coordinate

$S_y$ : Sorted according to  $y$ -coordinate

Pre

compute

$S_L \mid S_R$

$S_y' \rightarrow$  all points

in the strip (right)

Sorted according to  
y coordinate

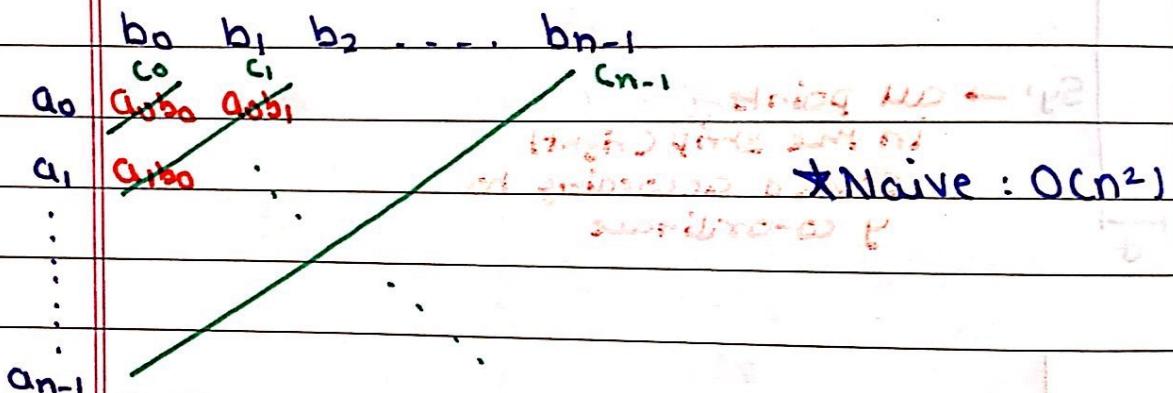
$P$

## ⑥ Fast Fourier Transform (FFT)

$$A(x) = \sum_{i=0}^{n-1} a_i x^i, \quad B(x) = \sum_{i=0}^{n-1} b_i x^i$$

$$C(x) = A(x) \cdot B(x)$$

$$= c_0 + c_1 x + \dots + c_{2n-2} x^{2n-2}$$



\* Polynomial can be represented as coefficient representation.

**Vector of coefficients**

\* Set of points representation

$$A(x) = \{(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1})\}$$

$$y_k = A(x_k)$$

\* Horner's Rule:

$$A(x) = a_0 + \dots + x(a_{n-3} + x(a_{n-2} + x a_{n-1}))$$

↓

Takes  $O(n)$

$$\begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^{n-1} \\ 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n-1} & x_{n-1}^2 & \dots & x_{n-1}^{n-1} \end{pmatrix} = X \quad \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{pmatrix} = A$$

$$\begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \end{pmatrix} = Y \quad * \quad XA = Y$$

$$A = X^{-1}Y$$

on the basis of point out last working is O(n^2)

\* Deg(C(x)) is 2n-2

(n^2) O(n^2)

Hence atleast 2n-1 points required to find its coefficients.

Complexity analysis  
Time complexity  
O(n^2)

→ Step 1:

$$x_0, x_1, \dots, x_{2n-2}$$

find A(x\_k) + B(x\_k) for each  $x_k$ :  $O(n^2)$

Complexity analysis  
Time complexity  
O(n^2)

→ Step 2:

$$C(x_k) = A(x_k) \cdot B(x_k) : O(n) \quad (\because 0 \leq k \leq 2n-2)$$

Complexity analysis  
Time complexity  
O(n^2)

→ Step 3:

Find c.e. of C(x)

$$C = X^{-1}Y : O(n^2)$$

Complexity analysis  
Time complexity  
O(n^2)

\*  $A(x) = a_0 + a_1x + \dots + a_{n-1}x^{n-1}$

$$A_{\text{even}}(x) = (a_0 + a_2x + a_4x^2 + \dots + a_{n-2}x^{\frac{n}{2}-1})$$

$$A_{\text{odd}}(x) = (a_1 + a_3x + a_5x^2 + \dots + a_{n-1}x^{\frac{n}{2}-1})$$

$$\therefore A(x) = A_{\text{even}}(x^2) + x A_{\text{odd}}(x^2)$$

\* Note Here we have to evaluate  $A(x)$  of deg  $n-1$  at  $2n$  points Let the time required to do so be  $T(n)$

→ Twiddle factors:

$$w_{j,r} = e^{2\pi j/r \cdot i} \\ = (\cos 2\pi j/r, \sin 2\pi j/r)$$

→ Let the required  $2n$  points be

$$w_{0,2n}, w_{1,2n}, \dots, w_{2n-1,2n}$$

\*  $A(w_{j,2n}) = A_{\text{even}}(w_{j,2n}^2) + w_{j,2n} A_{\text{odd}}(w_{j,2n}^2)$

$$w_{j+n,2n} = e^{2\pi(j+n)/2n \cdot i}$$

$$= e^{2\pi j/2n \cdot i} \cdot e^{in}$$

$$= -w_{j,2n}$$

$w_{j+n,2n}^2 = w_{j,2n}^2$

$$A(j) = A_{\text{even}}(j^2) + j A_{\text{odd}}(j^2)$$

$$A(-j) = A_{\text{even}}(j^2) - j A_{\text{odd}}(j^2)$$

$$A(w_{j,2n}) = A_{\text{even}}(w_{j,2n}^2) + w_{j,2n} A_{\text{odd}}(w_{j,2n}^2) \quad \dots (1)$$

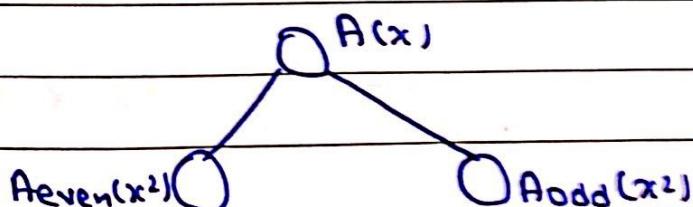
$$(\#*) \quad A(w_{j+n,2n}) = A_{\text{even}}(w_{j,2n}^2) - w_{j,2n} A_{\text{odd}}(w_{j,2n}^2) \quad \dots (2)$$

$$A(w_{j,2n}) \leftarrow \begin{array}{l} j=0,1,\dots,n-1 \text{ Apply } (1) \\ j=n,\dots,2n-1 \text{ Apply } (2) \end{array}$$

\* So To calculate  $A(x)$  for  $2n$  values

we effectively need to calculate  $A_{\text{even}}(x^2)$

&  $A_{\text{odd}}(x^2)$  for only  $n$  values



$$\therefore T(n) = 2T(n/2) + O(n) + O(n)$$

Divide Combine

$$\therefore T(n) = O(n \log n) \text{ for step 1}$$

\* DFT: Discrete Fourier Transform

Represent a polynomial using Twiddle factors