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5

Divide & Conquer

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Learning Outcomes

Unit 5: Divide & Conquer

- 1. Know the steps of the Divide & Conquer paradigm.
- **2.** Be able to solve simple Divide & Conquer recurrences.
- 3. Be able to design and analyze new algorithms using the Divide & Conquer paradigm.
- **4.** Know the performance characteristics of selection-by-rank algorithms.

Outline

5 Divide & Conquer

- 5.1 Divide & Conquer Recurrences
- 5.2 Order Statistics
- 5.3 Linear-Time Selection
- 5.4 Fast Multiplication
- 5.5 Majority
- 5.6 Closest Pair of Points in the Plane

Divide and conquer

Divide and conquer *idiom* (Latin: *divide et impera*) to make a group of people disagree and fight with one another so that they will not join together against one (Merriam-Webster Dictionary)

→ in politics & algorithms, many independent, small problems are better than one big one!

Divide-and-conquer algorithms:

- 1. Break problem into smaller, independent subproblems. (Divide!)
- **2.** Recursively solve all subproblems. (Conquer!)
- **3.** Assemble solution for original problem from solutions for subproblems.

Examples:

- Mergesort
- Quicksort
- ► Binary search
- ► (arguably) Tower of Hanoi

5.1 Divide & Conquer Recurrences

Back-of-the-envelope analysis

- before working out the details of a D&C idea, it is often useful to get a quick indication of the resulting performance
 - don't want to waste time on something that's not competitive in the end anyways!
- ▶ since D&C is naturally recursive, running time often not obvious instead: given by a recursive equation
- unfortunately, rigorous analysis often tricky
 - Remember mergesort?

$$C(n) = \begin{cases} 0 & n \le 1 \\ C(\lfloor n/2 \rfloor) + C(\lceil n/2 \rceil) + 2n & n \ge 2 \end{cases}$$

$$\Rightarrow C(n) = 2n \lfloor \lg(n) \rfloor + 2n - 4 \cdot 2^{\lfloor \lg(n) \rfloor} \quad \blacksquare$$

$$= \Theta(n \log n) \quad \textcircled{9}$$

▶ the following method works for many typical cases to give the right **order of growth**

The Master Method

- Assume a stereotypical D&C algorithm
 - ightharpoonup *a* recursive calls on (for some constant $a \ge 1$)
 - subproblems of size n/b (for some constant b > 1)
 - ▶ with non-recursive "conquer" effort f(n) (for some function $f: \mathbb{R} \to \mathbb{R}$)
 - base case effort d (some constant d > 0)

$$ightharpoonup \text{running time } T(n) \text{ satisfies} \qquad \boxed{ T(n) = \begin{cases} a \cdot T\left(\frac{n}{b}\right) + f(n) & n > 1 \\ d & n \leq 1 \end{cases} }$$

Theorem 5.1 (Master Theorem)

With $c := \log_h(a)$, we have for the above recurrence:

- (a) $T(n) = \Theta(n^c)$ if $f(n) = O(n^{c-\epsilon})$ for constant $\epsilon > 0$.
- **(b)** $T(n) = \Theta(n^c \log n)$ if $f(n) = \Theta(n^c)$.
- (c) $T(n) = \Theta(f(n))$ if $f(n) = \Omega(n^{c+\varepsilon})$ for constant $\varepsilon > 0$ and f satisfies the regularity condition $\exists n_0, \alpha < 1 \ \forall n \geq n_0 : a \cdot f\left(\frac{n}{h}\right) \leq \alpha f(n)$.

Master Theorem - Intuition & Proof Idea

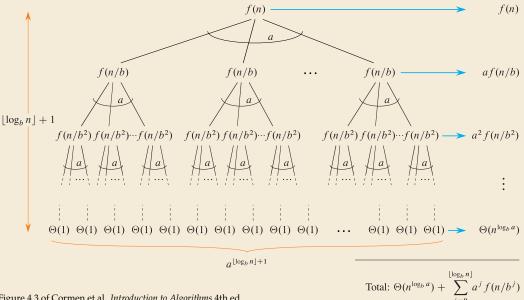


Figure 4.3 of Cormen et al. Introduction to Algorithms 4th ed.

When it's fine to ignore floors and ceilings

The polynomial-growth condition

▶ $f: \mathbb{R}_{>0} \to \mathbb{R}$ satisfies the *polynomial-growth condition* if

$$\exists n_0 \ \forall C \geq 1 \ \exists D > 1 \quad \forall n \geq n_0 \ \forall c \in [1,C] \ : \ \frac{1}{D} f(n) \leq f(cn) \leq D f(n)$$

- ▶ intuitively: increasing n by up to a factor C (and anywhere in between!) changes the function value by at most a factor D = D(C) (for sufficiently large n) zero allowed
- ► examples: $f(n) = \Theta(n^{\alpha} \log^{\beta}(n) \log \log^{\gamma}(n))$ for constants α , β , γ \rightarrow f satisfies the polynomial-growth condition

Lemma 5.2 (Polynomial-growth master method)

If the toll function f(n) satisfies the polynomial-growth condition, then the Θ -class of the solution of a D&C recurrence remains the same when ignoring floors and ceilings on subproblem sizes.

A Rigorous and Stronger Meta Theorem

Theorem 5.3 (Roura's Discrete Master Theorem)

Let T(n) be recursively defined as

$$T(n) = \begin{cases} b_n & 0 \le n < n_0, \\ f(n) + \sum_{d=1}^{D} a_d \cdot T\left(\frac{n}{b_d} + r_{n,d}\right) & n \ge n_0, \end{cases}$$

where $D \in \mathbb{N}$, $a_d > 0$, $b_d > 1$, for $d = 1, \ldots, D$ are constants, functions $r_{n,d}$ satisfy $|r_{n,d}| = O(1)$ as $n \to \infty$, and function f(n) satisfies $f(n) \sim B \cdot n^{\alpha} (\ln n)^{\gamma}$ for constants B > 0, α , γ . Set $H = 1 - \sum_{d=1}^{D} a_d (1/b_d)^{\alpha}$; then we have:

- (a) If H < 0, then $T(n) = O(n^{\tilde{\alpha}})$, for $\tilde{\alpha}$ the unique value of α that would make H = 0.
- **(b)** If H=0 and $\gamma>-1$, then $T(n)\sim f(n)\ln(n)/\tilde{H}$ with constant $\tilde{H}=(\gamma+1)\sum_{d=1}^D a_d\,b_d^{-\alpha}\ln(b_d)$.
- (c) If H = 0 and $\gamma = -1$, then $T(n) \sim f(n) \ln(n) \ln(\ln(n)) / \hat{H}$ with constant $\hat{H} = \sum_{d=1}^{D} a_d b_d^{-\alpha} \ln(b_d)$.
- (d) If H = 0 and $\gamma < -1$, then $T(n) = O(n^{\alpha})$.
- (e) If H > 0, then $T(n) \sim f(n)/H$.

5.2 Order Statistics

Selection by Rank

- Standard data summary of numerical data: (Data scientists, listen up!)
 - mean, standard deviation
 - ► min/max (range)
 - histograms
 - median, quartiles, other quantiles (a.k.a. order statistics)

easy to compute in $\Theta(n)$ time

? computable in $\Theta(n)$ time?

General form of problem: Selection by Rank

► **Given:** array A[0..n) of numbers and number $k \in [0..n)$.

but 0-based & /counting dups

- ▶ **Goal:** find element that would be in position k if A was sorted (kth smallest element).
- ▶ $k = \lfloor n/2 \rfloor$ \leadsto median; $k = \lfloor n/4 \rfloor$ \leadsto lower quartile k = 0 \leadsto minimum; $k = n \ell$ \leadsto ℓ th largest

Quickselect

- ► Key observation: Finding the element of rank *k* seems hard.

 But computing the rank of a given element is easy!
- \rightsquigarrow Pick any element A[b] and find its rank j.
 - ▶ j = k? \rightarrow Lucky Duck! Return chosen element and stop
 - ▶ j < k? \longrightarrow ... not done yet. But: The j + 1 elements smaller than $\leq A[b]$ can be excluded!
 - ▶ j > k? \rightarrow similarly exclude the n j elements $\geq A[b]$
- ▶ partition function from Quicksort:
 - returns the rank of pivot
 - separates elements into smaller/larger

```
procedure quickselect(A[l..r), k)

if r - \ell \le 1 then return A[l]

b := \text{choosePivot}(A[l..r))

j := \text{partition}(A[l..r), b)

if j == k

return A[j]

else if j < k

quickselect(A[j+1..n), k-j-1)

else \#/j > k

quickselect(A[0..j), k)
```

Quickselect – Iterative Code

Recursion can be replaced by loop (tail-recursion elimination)

```
procedure quickselect(A[l..r), k)
           if r - \ell \le 1 then return A[l]
2
           b := \text{choosePivot}(A[l..r))
3
           j := partition(A[l..r), b)
           if j == k
5
               return A[i]
           else if i < k
7
               quickselect(A[i+1..n), k-i-1)
           else //i > k
9
               quickselect(A[0..i), k)
10
```

```
procedure quickselectIterative(A[0..n), k)

l := 0; r := n

while r - l > 1

b := \text{choosePivot}(A[l..r))

j := \text{partition}(A[l..r), b)

if j \ge k then r := j - 1

return A[k]
```

- implementations should usually prefer iterative version
- analysis more intuitive with recursive version

Quickselect – Analysis

```
1 procedure quickselect(A[l..r), k)
2 if r - \ell \le 1 then return A[l]
3 b := \text{choosePivot}(A[l..r))
4 j := \text{partition}(A[l..r), b)
5 if j := k
6 return A[j]
7 else if j < k
8 quickselect(A[j+1..n), k-j-1)
9 else \#f > k
10 quickselect(A[0..j), k)
```

- ► cost = #cmps
- costs depend on n and k
- ▶ worst case: k = 0, but always j = n 2
 - \rightarrow each recursive call makes n one smaller at cost $\Theta(n)$
 - \rightarrow $T(n, k) = \Theta(n^2)$ worst case cost

average case:

- ightharpoonup let T(n,k) expected cost when we choose a pivot uniformly from A[0..n)
- \rightarrow formulate recurrence for T(n, k) similar to BST/Quicksort recurrence

$$T(n,k) = n + \frac{1}{n} \sum_{r=0}^{n-1} [r=k] \cdot 0 + [k < r] \cdot T(r,k) + [k > r] \cdot T(n-r-1,k-r-1)$$

Quickselect – Average Case Analysis

$$T(n,k) = n + \frac{1}{n} \sum_{r=0}^{n-1} [r=k] \cdot 0 + [k < r] \cdot T(r,k) + [k > r] \cdot T(n-r-1,k-r-1)$$

 $\blacktriangleright \operatorname{Set} \hat{T}(n) = \max_{k \in [0..n)} T(n, k)$

$$\rightsquigarrow \hat{T}(n) \le n + \frac{1}{n} \sum_{r=0}^{n-1} \max{\{\hat{T}(r), \hat{T}(n-r-1)\}}$$

▶ analyze hypothetical, worse algorithm: if $r \notin [\frac{1}{4}n, \frac{3}{4}n)$, discard pivot and repeat with new one!

$$ightharpoondown$$
 $\hat{T}(n) \leq \tilde{T}(n)$ defined by $\tilde{T}(n) \leq n + \frac{1}{2}\tilde{T}(n) + \frac{1}{2}\tilde{T}(\frac{3}{4}n)$
 $ightharpoondown$ $\tilde{T}(n) \leq 2n + \tilde{T}(\frac{3}{4}n)$

► Master Theorem Case 3: $\tilde{T}(n) = \Theta(n)$

Quickselect Discussion

- \bigcap $\Theta(n^2)$ worst case (like Quicksort)
- expected $cost \Theta(n)$ (best possible)
- no extra space needed
- adaptations possible to find several order statistics at once
- expected cost can be further improved by choosing pivot from a small sorted sample \rightarrow asymptotically optimal randomized cost: $n + \min\{k, n k\}$ comparisons in expectation achieved asymptotically by the *Floyd-Rivest algorithm*

5.3 Linear-Time Selection

Interlude – A recurring conversation

Cast of Characters:



Hi! I'm a computer science practitioner.

I love algorithms for the sometimes miraculous applications they enable. I care for **things** I can implement and **that actually work in practice**.



Hi! I'm a theoretical computer science researcher.

I find beauty in elegant and **definitive** answers to questions about complexity. I care for **eternal truths** and mathematically proven facts;

asymptotically optimal is what counts! (Constant factors are secondary.)

Quickselect Disagreements



For practical purposes, (randomized) Quickselect is perfect.

e.g. used in C++ STL std::nth_element



Yeah . . . maybe. But can we select by rank in O(n) deterministic **worst case** time?

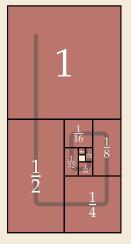
Better Pivots

It turns out, we can!

- All we need is better pivots!
 - ► If pivot was the exact median, we would at least halve #elements in each step
 - ▶ Then the total cost of all partitioning steps is $\leq 2n = \Theta(n)$.



But: finding medians is (basically) our original problem!





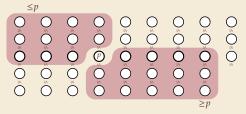
It totally suffices to find an element of rank αn for $\alpha \in (\varepsilon, 1 - \varepsilon)$ to get overall costs $\Theta(n)$!

The Median-of-Medians Algorithm

```
1 procedure choosePivotMoM(A[l..r))
       m := |n/5|
       for i := 0, ..., m-1
           sort(A[5i..5i + 4])
4
           // collect median of 5
5
            Swap A[i] and A[5i + 2]
       return quickselectMoM(A[0..m), \lfloor \frac{m-1}{2} \rfloor)
7
9 procedure quickselectMoM(A[1..r), k)
       if r - \ell \le 1 then return A[l]
10
       b := \text{choosePivotMoM}(A[l..r))
11
      j := partition(A[l..r), b)
12
       if i == k
13
            return A[i]
14
       else if i < k
15
            quickselectMoM(A[j+1..n), k-j-1) \rightsquigarrow C(n) \leq \Theta(n) + C(\frac{1}{5}n) + C(\frac{7}{10}n)
16
       else //i > k
17
            quickselectMoM(A[0..i), k)
18
```

Analysis:

- ► Note: 2 mutually recursive procedures → effectively 2 recursive calls!
- 1. recursive call inside choosePivotMoM on $m \leq \frac{n}{5}$ elements
- recursive call inside quickselectMoM



 \rightarrow partition excludes $\sim 3 \cdot \frac{m}{2} \sim \frac{3}{10}n$ elem.

$$\begin{array}{ll}
 & \longrightarrow C(n) \leq \Theta(n) + C(\frac{1}{5}n) + C(\frac{7}{10}n) \\
& \text{ansatz: overall} \\
& \text{cost linear} = \Theta(n) + C(\frac{9}{10}n) & \longrightarrow C(n) = \Theta(n)
\end{array}$$

5.4 Fast Multiplication

Integer Multiplication

- ▶ What's the cost of computing $x \cdot y$ for two integers x and y?
- → depends on how big the numbers are!
 - ▶ If x and y have O(w) bits, multiplication takes O(1) time on word-RAM
 - ▶ otherwise, need a dedicated algorithm!

Long multiplication (»Schulmethode«)

► Given
$$x = \sum_{i=0}^{n-1} x_i 2^i$$
 and $y = \sum_{i=0}^{n-1} y_i 2^i$, want $z = \sum_{i=0}^{2n-1} z_i 2^i$

```
1 for i := 0, ..., n-1

2 c := 0

3 for j := 0, ..., n-1

4 z_{i+j} := z_{i+j} + c + x_i \cdot y_j

5 c := \lfloor z_{i+j}/2 \rfloor

6 z_{i+j} := z_{i+j} \mod 2

7 end for

8 z_{i+n} := c

9 end for
```

- $ightharpoonup \Theta(n^2)$ bit operations
- ► could work with base 2^w instead of 2

$$\rightsquigarrow \Theta((n/w)^2)$$
 time

► here: count bit operations for simplicity can be generalized

Example:

easier in binary!
("shift and add")

1001010101 * 101101

110100011110001

Divide & Conquer Multiplication

- ▶ assume *n* is power of 2 (fill up with 0-bits otherwise)
- ▶ We can write
 - $x = a_1 2^{n/2} + a_2$ and
 - $y = b_1 2^{n/2} + b_2$
 - for a_1 , a_2 , b_1 , b_2 integers with n/2 bits

$$\rightarrow x \cdot y = (a_1 2^{n/2} + a_2) \cdot (b_1 2^{n/2} + b_2) = a_1 b_1 2^n + (a_1 b_2 + a_2 b_1) 2^{n/2} + a_2 b_2$$

- recursively compute 4 smaller products
- ightharpoonup combine with shifts and additions (O(n) bit operations)
- ▶ ... but is this any good?
 - $ightharpoonup T(n) = 4 \cdot T(n/2) + \Theta(n)$
 - \rightarrow Master Theorem Case 1: $T(n) = \Theta(n^2)$... just like the primary school method!?
 - but Master Theorem gives us a hint: cost is dominated by the leaves
 - → try to do more work in conquer step!

Karatsuba Multiplication

▶ how can we do "less divide and more conquer"?

Recall:
$$x \cdot y = |a_1b_12^n + (a_1b_2 + a_2b_1)2^{n/2} + a_2b_2|$$



-X- Let's do some algebra.

$$c := (a_1 + a_2) \cdot (b_1 + b_2)$$

= $a_1b_1 + (a_1b_2 + a_2b_1) + a_2b_2$

 $(a_1b_2 + a_2b_1) = c - a_1b_1 - a_2b_2$ this can be computed with 3 recursive multiplications $a_1 + a_2$ and $b_1 + b_2$ still have roughly n/2 bits

1 **procedure** karatsuba(x, y):

- // Assume x and y are $n = 2^k$ bit integers
- $a_1 := |x/2^{n/2}|$; $a_2 := x \mod 2^{n/2} // implemented by shifts$
- $b_1 := |y/2^{n/2}|$; $b_2 := y \mod 2^{n/2}$
- $c_1 := karatsuba(a_1, b_1)$
- $c_2 := karatsuba(a_2, b_2)$
- $c := karatsuba(a_1 + a_2, b_1 + b_2) c_1 c_2$
- **return** $c_1 2^n + c 2^{n/2} + c_2$ // shifts and additions

Analysis:

- nonrecursive cost: only additions and shifts
- ightharpoonup all numbers O(n) bits
- \rightarrow conquer cost $f(n) = \Theta(n)$

Recurrence:

- $ightharpoonup T(n) = 3T(n/2) + \Theta(n)$
- Master Theorem Case 1

$$\rightsquigarrow T(n) = \Theta(n^{\lg 3}) = O(n^{1.585})$$

much cheaper (for large n)!

Integer Multiplication

- until 1960, integer multiplication was conjectured to take $\Omega(n^2)$ bit operations
- → Karatsuba's algorithm was a big breakthrough
 - which he discovered as a student!
- ▶ idea can be generalized to breaking numbers into $k \ge 2$ parts (*Toom-Cook algorithm*)
- asymptotically *much* better algorithms are now known!
 - e. g., the *Schönhage-Strassen algorithm* with $O(n \log n \log \log n)$ bit operations (!)
 - ▶ these are based on the *Fast Fourier Transform* (FFT) algorithm
 - ▶ numbers = polynomials evaluated at base (e. g., z = 2)
 - → multiplication of numbers = convolution of polynomials
 - ▶ FFT makes computation of this convolution cheap by computing the polynomial via interpolation
 - ▶ Schönhage-Strassen adds careful finite-field algebra to make computations efficient

Matrix Multiplication

- ▶ The same trick can also be used for faster matrix multiplication
 - entry of A in row i and column k
- ▶ Recall: For $A, B \in \mathbb{R}^{n \times n}$ we define $C = A \cdot B$ via $c_{i,j} = \sum_{k=1}^{n} a_{i,k}^{j} b_{k,j}$
- \rightarrow Naive cost: n^2 sums with n terms each \rightarrow $\Theta(n^3)$ arithmetic operations
- ► Can use D&C as follows (assuming n is a power of 2 again)
 - ▶ Decompose $A = \begin{pmatrix} A_{1,1} & A_{1,2} \\ A_{2,1} & A_{2,2} \end{pmatrix}$, $B = \begin{pmatrix} B_{1,1} & B_{1,2} \\ B_{2,1} & B_{2,2} \end{pmatrix}$, $C = \begin{pmatrix} C_{1,1} & C_{1,2} \\ C_{2,1} & C_{2,2} \end{pmatrix}$

 - ▶ 8 recursive matrix multiplications on two $\frac{n}{2} \times \frac{n}{2}$ matrices + $\Theta(n^2)$ summations
 - # operations $T(n) = 8T(n/2) + \Theta(n^2)$
 - \longrightarrow Master Theorem Case 1: $T(n) = \Theta(n^3)$ © (but: still useful for better memory locality!)

Strassen Algorithm for Matrix Multiplication

- Observation (again): Can do more conquer for less divide!
- ► We recursively compute the following 7 products:

$$M_{1} := (A_{1,2} - A_{2,2}) \cdot (B_{2,1} + B_{2,2})$$

$$M_{2} := (A_{1,1} + A_{2,2}) \cdot (B_{1,1} + B_{2,2})$$

$$M_{3} := (A_{1,1} - A_{2,1}) \cdot (B_{1,1} + B_{1,2})$$

$$M_{4} := (A_{1,1} + A_{1,2}) \cdot B_{2,2}$$

$$M_{5} := A_{1,1} \cdot (B_{1,2} - B_{2,2})$$

$$M_{6} := A_{2,2} \cdot (B_{2,1} - B_{1,1})$$

$$M_{7} := (A_{2,1} + A_{2,2}) \cdot B_{1,1}$$

 \rightsquigarrow We then obtain the 4 parts of *C* as

$$C_{1,1} = M_1 + M_2 - M_4 + M_6$$

$$C_{1,2} = M_4 + M_5$$

$$C_{2,1} = M_6 + M_7$$

$$C_{2,2} = M_2 - M_3 + M_5 - M_7$$

(Proof: left as exercise 9)

Analysis:

- ► **conquer step:** larger but still *O*(1) # matrix add/subtract
- $\rightsquigarrow \Theta(n^2)$ operations for conquer
- \rightarrow total # arithmetic operations $T(n) = 7T(n/2) + \Theta(n^2)$
- \sim Master Theorem Case 1: $T(n) = \Theta(n^{\lg 7}) = O(n^{2.808})$

Open Problems

Multiplication is extremely fundamental, but its **computational complexity** is an **open problem** and subject of active research!

Integer multiplication:

- **conjectured** to require $\Omega(n \log n)$ bit operations (no proof known!)
- ▶ Harvey & van der Hoeven **2021**: $O(n \log n)$ algorithm possible!

Matrix multiplication (MM):

- more relevant than it might seem since complexity identical to
 - computing inverse matrices, determinants
 - ► Gaussian elimination (solving systems of linear equations)
 - recognition of context free languages
- \sim best exponent even has standard notation: smallest ω ∈ [2,3) so that MM takes $O(n^ω)$ operations
- ▶ Big open question: Is $\omega > 2$?
- ▶ best known bound: $\omega \le 2.371339$ (from 2024!)



5.5 Majority

Majority

- ▶ **Given:** Array A[0..n) of objects
- ► **Goal:** Check of there is an object x that occurs at $> \frac{n}{2}$ positions in A if so, return x
- ▶ Naive solution: check each A[i] whether it is a majority \longrightarrow $\Theta(n^2)$ time

Majority – Divide & Conquer

Can be solved faster using a simple Divide & Conquer approach:

- ► If *A* has a majority, that element must also be a majority of at least one half of *A*.
- → Can find majority (if it exists) of left half and right half recursively
- \rightsquigarrow Check these ≤ 2 candidates.
- ▶ Costs similar to mergesort $\Theta(n \log n)$

```
1 procedure majority(A[0..n))
        if n == 1 then return A[0] end if
        k := \lfloor \frac{n}{2} \rfloor
 3
        M_{\ell} := \text{majority}(A[0..k))
 4
        M_r := \text{majority}(A[k..n])
 5
        if M_{\ell} == M_r then return M_{\ell} end if
        m_{\ell} := 0; m_r := 0
        for i := 0, ..., n-1
             if A[i] == M_{\ell} then m_{\ell} = m_{\ell} + 1 end if
             if A[i] == M_r then m_r = m_r + 1 end if
10
        end for
11
        if m_{\ell} \geq k+1
12
             return Me
13
        else if m_r > k+1
14
             return M_r
15
        else
16
             return NO MAJORITY ELEMENT
17
```

Majority – Linear Time

We can actually do much better!

```
1 def MJRTY(A[0..n))

2  c := 0

3  for i := 1, ..., n-1

4  if c == 0

5  x := A[i]; c := 1

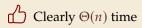
6  else

7  if A[i] == x then c := c+1 else c := c-1

8  return x
```



- ightharpoonup MJRTY(A[0..n)) returns *candidate* majority element
- either that candidate is the majority element or none exists(!)



5.6 Closest Pair of Points in the Plane

Closest Pair of Points in the Plane

- ► **Given:** Array P[0..n) of points in the plane (\mathbb{R}^2) each has x and y coordinates: P[i].x and P[i].y
- ► **Goal:** Find pair P[i], P[j] that is closest in (Euclidean) distance i. e., i and j that minimize $\sqrt{(P[i].x P[j].x)^2 + (P[i].y P[j].y)^2}$
- ▶ Naive solution: compute distance of each pair \longrightarrow $\Theta(n^2)$ time
 - ► cost here = # arithmetic operations
 - ▶ ignore numerical accuracy
 - → formally work on the real RAM
 - ▶ like word-RAM, but words contain **exact** real numbers
 - Support arithmetic operations and comparisons, but not bitwise operations or [·] and [·]

Closest Pair – Divide & Conquer

Closest Pair – Refined Conquer

Closest Pair - Code

```
1 procedure closestDist(P[0..n), byX[0..n), byY[0..n))
       // P contains n points with distinct x—coordinates
       //P[byX[0]].x \le P[byX[1]].x \le \cdots \le P[byX[n]].x
       //P[byY[0]].y \le P[byY[1]].y \le \cdots \le P[byY[n]].y
       if n == 2 return d_2(P[0], P[1])
       if n == 3 return min\{d_2(P[0], P[1]),
                         d_2(P[1], P[2]), d_2(P[0], P[2])
       // 1. Split by median x and recurse
       k := |n/2|; m := byX[m]
9
       byX_I := byX[0..m); byX_R := byX[m..n)
10
       P_L, P_R, byY_L, byY_R := \text{new empty array}
11
       for i := 0, ..., n-1
12
           if P[byY[i]].x < m
13
                P_L.append(P[byY[i]])
14
                byY_{I}.append(byY[i])
15
           else
16
                P_R.append(P[byY[i]])
17
                byY_R.append(byY[i])
18
            end if
19
       end for
20
       // ...
21
```

```
// ... closestDist continued
22
         \delta_L := \operatorname{closestDist}(P_L, byX_I, byY_I)
23
        \delta_R := \text{closestDist}(P_R, byX_R, byY_R)
24
         \delta := \min\{\delta_L, \delta_R\}
25
        // 2. Check pairs straddling x = m line
        // Find points close to m
27
        for i := 0, ..., n-1
28
              if |P[byY[i]].x - m| \leq \delta
29
                   C.append(byY[i])
30
              end if
31
        end for
32
        // Distance \leq \delta implies within 8 positions in C
33
         for i := 0, ..., C.size()
34
              for i := i + 1, ..., i + 7
35
                   \delta := \min\{\delta, d_2(P[C[i]], P[C[i]])\}
36
              end for
37
         end for
38
         return \delta
39
41 procedure d_2(P, Q)
         return \sqrt{(P.x - Q.x)^2 + (P.y - Q.y)^2}
```