

## Medians and Order Statistics

- Order statistic
  - The  $i_{th}$  order statistic of a set of  $n$  elements is the  $i_{th}$  smallest element
- Selection problem
  - Find the  $i_{th}$  smallest element
    - The  $i_{th}$  element if the array is sorted on non-decreasing order
- The median
  - $n$  is odd
    - The median is the  $(n+1)/2_{th}$  element
  - $n$  is even
    - Upper median:  $n/2 + 1$
    - Lower median:  $n/2$
  - For simplicity
    - We refer the median as the lower median: the  $\lfloor (n+1)/2 \rfloor_{th}$  smallest element

## Divide & Conquer for Selection

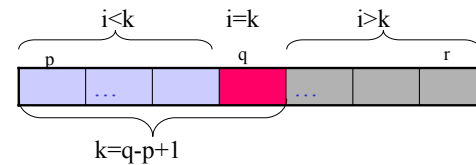
```

select(A[p..r], i) // return the i-th smallest element of A[p..r]
{
  if (p==r) // i must be 1 here
    return A[p];

  q = partition(A[p..r]); // same partition as used in quick sort
  k = q-p+1; // the k-th element now is the k-th smallest element

  if (i==k) return A[q];
  if (i<k) return select(A[p..q-1], i);
  if (i>k) return select(A[q+1..r], i-k);
}

```



## Cost

- Worst case
  - $\Theta(n^2)$
  - Can we make it  $\Theta(n)$ ?
- Average cost

$$T(n) \leq \frac{1}{n} \sum_{k=1}^n T(\max(k-1, n-k)) + O(n)$$

$$\max(k-1, n-k) = \begin{cases} k-1 & \text{if } k > \lfloor \frac{n}{2} \rfloor \\ n-k & \text{if } k \leq \lfloor \frac{n}{2} \rfloor \end{cases}$$

## Average cost

- Guess that  $T(n) \leq cn$  for some constant  $c$

$$\begin{aligned}
 T(n) &\leq \frac{1}{n} \sum_{k=1}^n T(\max(k-1, n-k)) + O(n) \\
 &\leq \frac{2}{n} \sum_{k=\lfloor \frac{n}{2} \rfloor}^{n-1} T(k) + O(n) \leq \frac{2c}{n} \sum_{k=\lfloor \frac{n}{2} \rfloor}^{n-1} k + O(n) \\
 &= \frac{2c}{n} \left( \sum_{k=1}^{n-1} k - \sum_{k=1}^{\lfloor \frac{n}{2} \rfloor - 1} k \right) + O(n) \\
 &= \frac{2c}{n} \left( \frac{1}{2} n(n-1) - \frac{1}{2} \left\lfloor \frac{n}{2} \right\rfloor \left( \left\lfloor \frac{n}{2} \right\rfloor - 1 \right) \right) + O(n) \\
 &\leq c(n-1) - \frac{c}{n} \frac{n-1}{2} \frac{n-3}{2} + O(n) \\
 &= \frac{3}{4} cn - \frac{3c}{4n} + O(n) \leq \frac{3}{4} cn + O(n) \leq cn
 \end{aligned}$$

### Relation between selection and median

- Any selection can be used to find the median
  - Select  $\lfloor (n+1)/2 \rfloor$ th smallest element
- Any median algorithm can be used for selection
  - Choose median as pivot
  - Partition the array  $A[p..r]$  into three sections:
    - $T[p..q-1]$  all elements less than the pivot
    - $T[q]$ : the element equal to the pivot
    - $T[q+1..r]$  all elements greater than the pivot
  - We are either done or continue selection on one section

### Selection using pseudomedian

```
select(A[p..r], i) {
    if (p==r) return A[p];

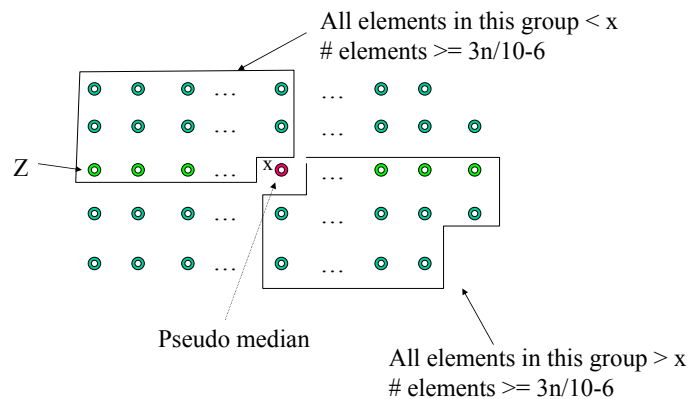
    x = pseudomedian(A[p..r]);
    q = partition(A[p..r], x);
    k = q-p+1;

    if (i==k)
        return A[q];
    if (i<k)
        return select(A[p..q-1], i);
    if (i>k)
        return select(A[q+1..r], i-k);
}
```

```
pseudomedian(T[1..n])
{
    if (n <= 5)
        return adhocmedian(A);
    z = ⌈n/5⌉;
    for (i=1; i<=z; i++)
        Z[i] = adhocmedian(A[5i-4..min(5i,n)]);
    return select(Z[1..z], ⌊(z+1)/2⌋);
}
```

We assume the elements are distinct.

### How does pseudomedian()?



At most  $7n/10+6$  elements are larger (smaller) than  $p$

### Theorem

- The selection algorithm used with pseudomedian finds the  $i$ -th smallest among  $n$  elements in  $\Theta(n)$  in the worst case. In particular, the median can be found in linear time in the worst case.

– There exists a const  $a$  such that

$$T(n) \leq an + T\left(\left\lceil \frac{n}{5} \right\rceil\right) + \max \{T(m) \mid m \leq \frac{7n}{10} + 6\}$$

– We like to show  $T(n) \in O(n)$

### Constructive induction

- We show that  $t(n) \leq cn$  for a constant  $c$  and all  $n \geq 1$

- Induction basis?
- Induction hypothesis?
- Induction step

$$\begin{aligned} T(n) &\leq T\left(\left\lceil \frac{n}{5} \right\rceil\right) + \max\{T(m) \mid m \leq \frac{7n}{10} + 6\} + an \\ &\leq \left(\frac{n}{5} + 1\right)c + \left(\frac{7n}{10} + 6\right)c + an \\ &= \frac{9c}{10}n + 7c + an \\ &= cn + \left(-\frac{cn}{10} + 7c + an\right) \end{aligned}$$

### Strassen's algorithm: matrix multiplication

- Classic matrix multiplication takes time  $\Theta(n^3)$

$$C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

- Strassen's algorithm makes it  $\Theta(n^{\lg 7}) \approx \Theta(n^{2.81})$

### How does it work?

- Given  $A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$  and  $B = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix}$

- Define
  - $m_1 = (a_{21} + a_{22} - a_{11})(b_{22} - a_{12} + b_{11})$
  - $m_2 = a_{11}b_{11}$
  - $m_3 = a_{12}b_{21}$
  - $m_4 = (a_{11} - a_{21})(b_{22} - b_{12})$
  - $m_5 = (a_{21} + a_{22})(b_{12} - b_{11})$
  - $m_6 = (a_{12} - a_{21} + a_{11} - a_{22})b_{22}$
  - $m_7 = a_{22}(b_{11} + b_{22} - b_{12} - b_{21})$

- Then  $C = \begin{pmatrix} m_2 + m_3 & m_1 + m_2 + m_5 + m_6 \\ m_1 + m_2 + m_4 - m_7 & m_1 + m_2 + m_4 + m_5 \end{pmatrix}$

### Cost analysis

- Let  $t(n)$  be the time needed to multiply two  $n \times n$  matrices.
  - $t(n) = 7t(n/2) + g(n)$  where  $g(n) \in \Theta(n^2)$
  - Applying the Master theorem, we get  $t(n) \in \Theta(n^{\lg 7})$