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Computer Theory Lab.



# 7.1 Description of quicksort

- Divide
- Conquer
- Combine



#### QUICKSORT(A,p,r)

- 1 if p < r
- 2 then  $q \leftarrow PARTITION(A, p, r)$
- 3 QUICKSORT(A,p,q)
- 4 QUICKSORT(A, q+1, r)

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# Partition(A, p, r)

```
1 x \leftarrow A[r]
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- $2 \quad i \leftarrow p-1$
- 3 for  $j \leftarrow p$  to r-1
- 4 **do if**  $A[j] \le x$
- 5 **then**  $i \leftarrow i + 1$
- 6 exchange  $A[i] \leftrightarrow A[j]$
- 7 exchange  $A[i+1] \leftrightarrow A[r]$
- 8 **return** i +1



At the beginning of each iteration of the loop of lines 3-6, for any array index k,

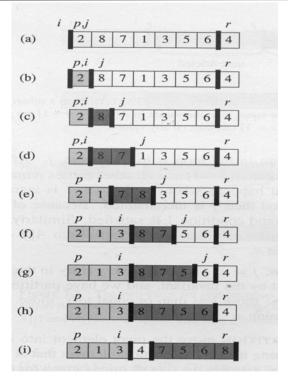
- 1. if  $p \le k \le i$ , then  $A[k] \le x$ .
- 2. if  $i + 1 \le k \le j 1$ , then A[k] > x.
- 3. if k = r, then A[k] = x.

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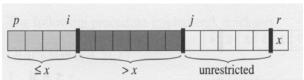


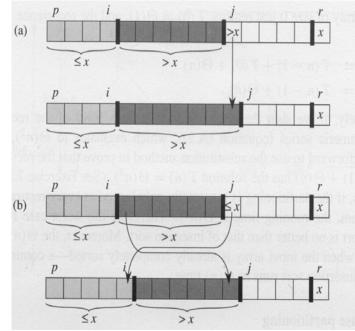
#### The operation of *Partition* on a sample array





# Two cases for one iteration of procedure *Partition*





Complexity: Partition on A[p...r] is  $\Theta(n)$ where n = r - p + 1

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### 7.2 Performance of quicksort

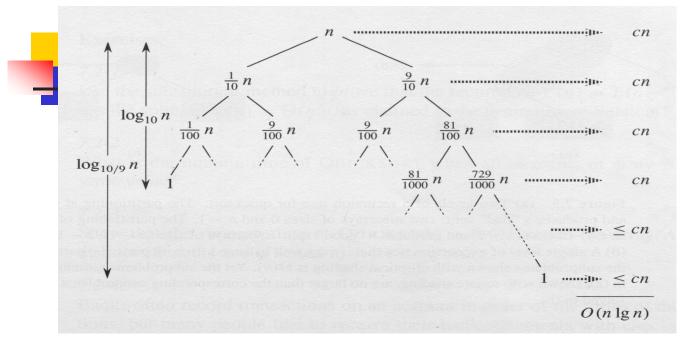
Worst-case partition:

$$T(n) = T(n-1) + \Theta(n)$$

$$= \sum_{k=1}^{n} \Theta(k) = \Theta(\sum_{k=1}^{n} k) = \Theta(n^{2})$$

Best-case partition:

$$T(n) = 2T(n/2) + \Theta(n)$$
  
$$\Rightarrow T(n) = \Theta(n \log n)$$



Balanced partition  $T(n) = \Theta(n \log n)$ 

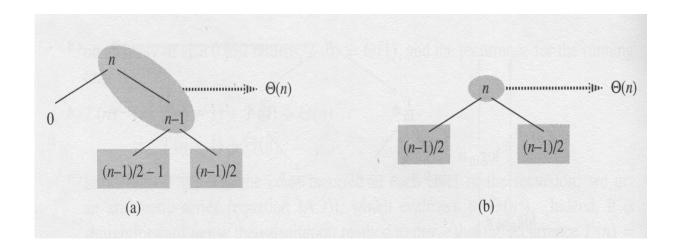
$$T(n) = T(9n/10) + T(n/10) + \Theta(n)$$
  
$$\Rightarrow T(n) = \Theta(n \log n)$$

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### Intuition for the average case $T(n) = \Theta(n \log n)$





### 7.3 Randomized versions of partition

RANDOMIZED\_PARTITION(A,p,r)

- 1  $i \leftarrow RANDOM(p,r)$
- 2 exchange  $A[p] \leftrightarrow A[i]$
- 3 **return** PARTITION(A,p,r)

RANDOMIZED\_QUICKSORT(A,p,r)

- 1 if p < r
- 2 then

$$q \leftarrow RANDOMIZED\_PARTITION(A, p, r)$$

- 3 RANDOMIZED QUICKSORT(A,p,q)
- Chapter 7 4 RANDOMIZED\_QUICKSORT(A,q+1,r)

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### 7.4 Analysis of quicksort

7.4.1 Worst-case analysis

$$T(n) = \max_{0 \le q \le n-1} (T(q) + T(n-q-1)) + \Theta(n)$$

guess  $T(n) \le cn^2$ 

$$T(n) \le \max_{0 \le q \le n-1} (cq^{2} + c(n-q-1)^{2}) + \Theta(n)$$

$$= c \max_{0 \le q \le n-1} (q^{2} + (n-q-1)^{2}) + \Theta(n)$$

$$\le cn^{2} - 2c(n-1) + \Theta(n)$$

$$\le cn^{2}$$

pick the constant c large enough so that the 2c(n-1) term dominates the  $\Theta(n)$  term.

$$\Rightarrow T(n) = \Theta(n^2)$$

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Show that  $q^2 + (n-q)^2$  achieves a maximum over

$$q = 1, 2, ...., n-1$$
 **when**  $q = 1$  **or**  $q = n-1$ 

**ans:** 
$$\not = f(q) = q^2 + (n-q)^2$$

一大微分: f'(q) = 2q - 2(n-q) = 4q - 2n

$$\Leftrightarrow f'(q) = 0 \Rightarrow 4q - 2n = 0 \Rightarrow q = \frac{n}{2}$$
 (極/小恒)

二次微分: f''(q)=4 (開口向上)

因為  $1 \le q \le n-1$  所以  $f(1) = f(n-1) = 1 + (n-1)^2$  (相對極大值)

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### 7.4.2 Expected running time

- Running time and comparsions
- Lemma 7.1
  - Let X be the number of comparisons performed in line 4 of *partition* over the entire execution of *Quicksort* on an *n*-element array. Then the running rime of *Quicksort* is *O*(*n*+X)



#### we define

$$X_{ij} = I \{z_i \text{ is compared to } z_j\},$$

$$X = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}.$$

$$E[X] = E\left[\sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}\right]$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} E[X_{ij}]$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \Pr \quad \{z_i \text{ is compared to } z_j\}$$

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Pr{z<sub>i</sub> is compared to z<sub>j</sub>} = Pr{z<sub>i</sub> or z<sub>j</sub> is first pivot chosen from Z<sub>ij</sub>} = Pr{z<sub>i</sub> is first pivot chosen from Z<sub>ij</sub>} + Pr{z<sub>j</sub> is first pivot chosen from Z<sub>ij</sub>} =  $\frac{1}{j-i+1} + \frac{1}{j-i+1}$ =  $\frac{2}{j-i+1}$ 

$$\therefore E[X] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{2}{j-i+1}.$$

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$$E[X] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{2}{j-i+1}$$

$$= \sum_{i=1}^{n-1} \sum_{k=1}^{n-i} \frac{2}{k+1}$$

$$< \sum_{i=1}^{n-1} \sum_{k=1}^{n} \frac{2}{k}$$

$$= \sum_{i=1}^{n-1} O(\lg n)$$

$$= O(n \lg n)$$

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### another analysis

$$T(n) = \frac{1}{n}(T(1) + T(n-1) + \sum_{q=1}^{n-1}(T(q) + T(n-q)) + \Theta(n)$$

$$T(1) = 1$$

$$T(n-1) = O(n^{2})$$

$$\Rightarrow \frac{1}{n}(T(1) + T(n-1)) = O(n)$$

$$T(n) = \frac{1}{n}(T(n) + T(n-1)) = O(n)$$

$$\frac{1}{n}(T(1) + T(n-1) + \sum_{q=1}^{n-1} (T(q) + T(n-q)) + \Theta(n)$$

$$= \frac{1}{n} (\sum_{q=1}^{n-1} T(q) + T(n-q)) + \Theta(n)$$

$$= \frac{2}{n} (\sum_{k=1}^{n-1} T(k)) + \Theta(n)$$



guess  $T(n) \le an \log n + b$ 

$$T(n) \le \frac{2}{n} \left( \sum_{k=1}^{n-1} ak \log k + b \right) + \Theta(n)$$
$$= \frac{2a}{n} \sum_{k=1}^{n-1} k \log k + \frac{2b}{n} (n-1) + \Theta(n)$$

We will prove 
$$\sum_{k=1}^{n-1} k \log k \le \frac{n^2 \log n}{2} - \frac{n^2}{8}$$

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$$T(n) \le \frac{2a}{n} (\frac{1}{2}n^2 \log n - \frac{n^2}{8}) + \frac{2b(n-1)}{n} + \Theta(n)$$

$$\le an \log n - \frac{an}{4} + 2b + \Theta(n)$$

$$= an \log n + b + (\Theta(n) + b - \frac{an}{4})$$

$$\le an \log n + b$$

Choose a large enough so that  $\frac{an}{4} \ge \Theta(n) + b$ .  $\Rightarrow T(n) = O(n \log n)$ .



$$\sum_{k=1}^{n-1} k \log k = \sum_{k=1}^{\lceil n/2 \rceil - 1} k \log k + \sum_{k=\lceil n/2 \rceil}^{n-1} k \log k$$

$$\leq (\log n - 1)^{\lceil n/2 \rceil - 1} k + \log n \sum_{k=\lceil n/2 \rceil}^{n-1} k$$

$$= \log n \sum_{k=1}^{n-1} k - \sum_{k=1}^{\lceil n/2 \rceil - 1} k$$

$$\leq \frac{n(n-1)\log n}{2} - \frac{1}{2} (\frac{n}{2} - 1) \frac{n}{2}$$

$$\leq \frac{n^2 \log n}{2} - \frac{n^2}{8}$$
if  $n \geq 2$ .

Another approach: Using 
$$\int x \ln x dx = \frac{1}{2}x^2 \ln x - \frac{1}{4}x^2$$

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