

Advanced Algorithmic Techniques (COMP523)

Recursion and Divide and Conquer Techniques #4

Recap and plan

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- **Last lecture:**
 - Finding the closest pair of points.
 - Integer Multiplication.

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 - Finding the closest pair of points.
 - Integer Multiplication.
- **This lecture:**
 - The Selection problem.
 - i.e., finding the i^{th} -order statistic in an array.

The selection problem

- **Definition:** The i^{th} -order statistic of a set of n (distinct) elements is the i^{th} smallest element.
 - i.e., the element which is larger than exactly $i-1$ other elements.

The Selection Problem: $\text{Selection}(\mathbf{A}[1, \dots, n], i)$

Input: A set of n (distinct) numbers (in an array \mathbf{A}) and a number i , with $1 \leq i \leq n$.

Output: The i^{th} -order statistic of the set.

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- Return the i -th element of the sorted array.
- Is sorting an overkill?

Divide and conquer

- Split the input into smaller inputs.
- Solve the problem for the smaller inputs recursively.
- Combine the solutions into a solution for the original problem.

A glance back at the QuickSort algorithm

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The Quicksort algorithm

- **Mergesort** was based on the **Merge** procedure for joining the sorted sub-arrays into a sorted array.
- **Quicksort** first divides the array into two parts, such that the first part is “smaller” than the second part.
 - This is done via the **Partition** procedure.
- Then it calls itself recursively.
- The two parts are joined, but this is trivial.

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Recall the **Partition** procedure

Revisiting the Partition procedure

Procedure **Partition**($A[i, \dots, j]$)

Running time **$O(n)$**

Choose a **pivot element x** of **A**

$k = i - 1$

For $h = i$ to $j - 1$ do

 If $A[h] \leq x$

$k = k + 1$

 Swap $A[k]$ with $A[h]$

 Swap $A[k + 1]$ with $A[j]$

Return $k + 1$

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- Then, we can reduce the search for the i -th element to one of the three subarrays.
- How can we choose the element x *appropriately*, such that the subarrays $A[1, \dots, x-1]$ and $A[x+1, \dots, n]$ are of (approximately) equal size?

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 - The median is the number that is larger than exactly $(n+1)/2 - 1$ numbers.
 - The median is the $[(n+1)/2]^{th}$ -order statistic.
- What is an algorithm for finding the median?
 - **Selection**($\mathbf{A}[1, \dots, n], (n+1)/2$)

Let's try to do that...

Algorithm **Selection**(**A**[1,...,n],*i*)

x = **Selection**(**A**[1,...,n],(n+1)/2)

y = **Partition**(**A**[1,...,n],**x**)

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$\mathbf{y} = \mathbf{Partition}(\mathbf{A}[1, \dots, n], \mathbf{x})$

Before you conquer, you need to divide!

Let's try to do that...

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Do you see a problem?

x = **Selection**(**A**[**1**,...,**n**],(**n**+**1**)/**2**)

y = **Partition**(**A**[**1**,...,**n**],**x**)

Before you conquer, you need to divide!

Are we stuck?

- We need to partition the array into two using a good pivot element (**the median**).
- Or otherwise the running time of the recursion will be bad!
- But to find the median, we need an algorithm for selection!

Are we stuck?

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- Or otherwise the running time of the recursion will be bad!
- But to find the median, we need an algorithm for selection!

A good pivot element

- Split the array **A** into sub-arrays with **5 elements** each.
 - The last one might have fewer elements.

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Run **InsertionSort**

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- For each one of those, find the **median**.
- Find the **median-of-medians**.

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Run **Selection**

This failed...

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...but this won't.

Algorithm **Selection**($\mathbf{A}[1, \dots, n], i$)

Split the array \mathbf{A} into $n/5$ arrays of size 5

For each subarray \mathbf{A}_i , find the **median**

Let $m_1, m_2, \dots, m_{n/5}$ be those medians

x = **Selection**($\mathbf{A}[1, \dots, n], (n+1)/2$) /*Find the median of medians */

y = **Partition**($\mathbf{A}[1, \dots, n], \mathbf{x}$) /*Partition the array using **x** as the pivot */

The **Selection** algorithm (not exactly pseudocode)

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$\mathbf{k}-1$ is the number of elements in the lower subarray.

If $i = \mathbf{k}$, return \mathbf{x}

If $i < \mathbf{k}$, return **Selection**($\mathbf{A}[1, \dots, \mathbf{k}-1], i$)

If $i > \mathbf{k}$, return **Selection**($\mathbf{A}[\mathbf{k}+1, \dots, n], i-\mathbf{k}$)

Zooming in

If $i = k$, return x

If $i < k$, return **Selection**($A[1, \dots, k-1], i$)

If $i > k$, return **Selection**($A[k+1, \dots, n], i-k$)

- We are looking for the i^{th} -order statistic.
- If $i=k$, then x is the answer - it is larger than $k-1$ elements.
- If $i \leq k$, the answer cannot be in the second part, as then i would be larger than at least $k-1$ elements.
- For the same reason, if $i \geq k$, the answer cannot be in the first part.

Running time

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T(|S_{max}|)

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$|\mathbf{S}_{\max}| = \max(\mathbf{k}-1, n-\mathbf{k})$

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Before we proceed, we have to bound $|S_{\max}|$.

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- Each one of these groups has at least (...) elements $> \mathbf{x}$.

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 - Because $x \leq$ their “baby median”.
 - Except possibly

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- At least **half of the** subarrays have “baby medians” $\geq x$.
- Each one of these groups has at least **3** elements $> x$.
 - Because $x \leq$ their “baby median”.
- Except possibly **the group containing x** and **the group that has fewer than 5 elements.**

Bounding the size of the subarrays

What is the total number of elements larger than x ?

$$3 \left(\left\lfloor \frac{1}{2} \cdot \left\lfloor \frac{n}{5} \right\rfloor \right\rfloor - 2 \right) \geq \frac{3n}{10} - 6$$

elements $> x$ in each of those groups

groups

groups who could be exceptions

groups with "baby medians" $> x$

This means that the size of the lower subarray is at most

$$7n/10 + 6$$

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- A symmetric argument shows that the size of the upper subarray is at most $7n/10 + 6$
- Back to the recurrence:

$$T(n) \leq T(n/5) + T(|S_{\max}|) + cn = T(n/5) + T(7n/10+6) + bn$$

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- We get that

$$\begin{aligned}T(n) &\leq c(n/5) + c(7n/10+6) + bn \\ &= 9cn/10 + 7c + bn \\ &= cn + (-cn/10 + 7c + bn)\end{aligned}$$

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- This is at most cn whenever $-cn/10 + 7c + bn \leq 0$, or equivalently, when $c \geq 10bn/(n-70)$.

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- This is at most cn whenever $-cn/10 + 7c + bn \leq 0$, or equivalently, when $c \geq 10bn/(n-70)$.
- If $n \geq 140$, then $n/(n-70) \leq 2$ and then, it suffices to have $c \geq 20b$.

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Solving the recurrence

- We want to show that there is some constant $c > 0$, such that $T(n) \leq cn$ for all $n > 0$.
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- We will prove the statement by induction.

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 - **Base case:** For every $n \leq 140$, $T(n) \leq cn$

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- We will prove the statement by induction.
 - **Base case:** For every $n \leq 140$, $T(n) \leq cn$
 - **Inductive Step:** Suppose that it holds for all n up to $k=140$. Then for $n=k+1$, we have $T(n) \leq cn + (-cn/10 + 7c + bn)$

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- We will prove the statement by induction.
 - **Base case:** For every $n \leq 140$, $T(n) \leq cn$
 - **Inductive Step:** Suppose that it holds for all n up to $k=140$. Then for $n=k+1$, we have $T(n) \leq cn + (-cn/10 + 7c + bn)$
 - This follows from the fact that $n > 140$ and $c \geq 20b$.

Bonus: The Master Theorem

Suppose $T(n) \leq \alpha T(\lceil n/b \rceil) + O(n^d)$

for some constants $\alpha > 0$, $b > 1$ and $d \geq 0$.

$$\text{Then, } T(n) = \begin{cases} O(n^d), & \text{if } d > \log_b \alpha \\ O(n^d \log_b n), & \text{if } d = \log_b \alpha \\ O(n^{\log_b \alpha}), & \text{if } d < \log_b \alpha \end{cases}$$

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Example: For MergeSort, $\alpha=b=2$ and $d=1$, we get **$O(n \log n)$** .