



Partition and Selection

Foundations of Algorithms
Guven

Reading Assignment

- Cormen, Chapter 7, 9



Outline

- Divide and Conquer
- Partition
- Selection
- Quickselect
- Deterministic Selection

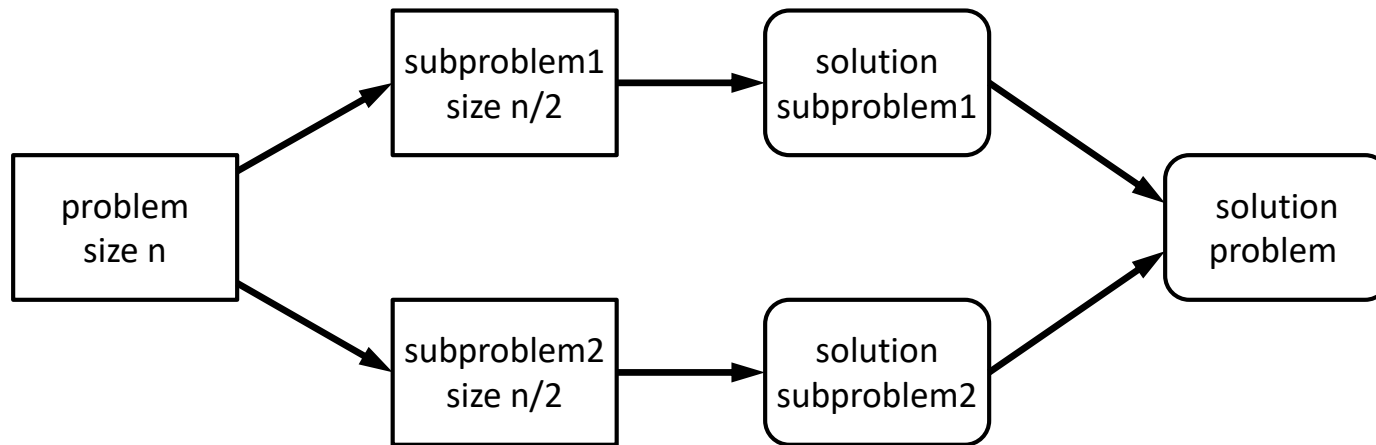


Divide and Conquer

- Algorithm design paradigm:
 1. Divide instance of the problem into two or more smaller instances (disjoint)
 2. Solve smaller instances recursively
 3. Obtain solution to original (larger) instance by combining these (smaller) solutions
- Multi-branched recursion
 - e.g. Fibonacci sequence: $F_n = F_{n-1} + F_{n-2}$ (true benefit DC?)
 - e.g. Fast Fourier Transform (true benefit DC)
 - Note, in practice recursion is avoided by direct iteration



Divide and Conquer Paradigm



- Divide and conquer \leftrightarrow e.g. Hadoop MapReduce
 - Not overlapping solutions
 - Not dynamic programming where the previous results are used in the next iteration
 - Each subproblem has an independent solution

Fibonacci Generation

- $F_0=0$, $F_1=1$, and $F_n=F_{n-1}+F_{n-2} \forall n \geq 2$
 - $\langle 0, 1, 1, 2, 3, 5, 8, 13, 21, \dots \rangle$
- A recursive solution

```
def fibr(n):  
    if n < 2:  
        return n  
    else:  
        return fibr(n-1)+fibr(n-2)
```

- Time complexity: $O(2^n)$
 - `fibr` looks like a divide and conquer approach
 - But it is not. Why? Because the solutions are overlapping



Quicksort

```
def quicksort(A,p,r):  
    if p<r:  
        q=partition(A,p,r)    # divide  
        quicksort(A,p,q-1)    # T(n/2)  
        quicksort(A,q+1,r)    # T(n/2)  
  
def partition(A,p,r):  
    x=A[r]; i=p-1  
    for j in range(p,r):    # Takes (r-p) operations  
        if A[j]<=x:  
            i+=1  
            A[i],A[j]=A[j],A[i]    # swap  
    A[i+1],A[r]=A[r],A[i+1]    # swap the pivot  
    return i+1
```

- Quicksort is also called partition-exchange sort algorithm

Quicksort Example

<i>i</i>	<i>j,p</i>									<i>r</i>
		1	2	3	4	5	4	3	2	3
	<i>i,p</i>	<i>j</i>							<i>r</i>	
		1	2	3	4	5	4	3	2	3
	<i>p</i>	<i>i</i>	<i>j</i>						<i>r</i>	
		1	2	3	4	5	4	3	2	3
	<i>p</i>			<i>i</i>	<i>j</i>					<i>r</i>
		1	2	3	4	5	4	3	2	3
	<i>p</i>				<i>i</i>	<i>j</i>				<i>r</i>
		1	2	3	4	5	4	3	2	3
	<i>p</i>					<i>i</i>	<i>j</i>			<i>r</i>
		1	2	3	4	5	4	3	2	3
	<i>p</i>						<i>i</i>	<i>j</i>		
		1	2	3	4	5	4	3	2	3
	<i>p</i>							<i>i</i>	<i>j</i>	<i>r</i>
		1	2	3	4	5	4	3	2	3
	<i>p</i>								<i>j</i>	<i>r</i>
		1	2	3	4	5	4	3	2	3
	<i>p</i>									<i>r</i>
		1	2	3	4	5	4	3	2	3
		1	2	3	3	5	4	3	2	4



In-class Exercise

- $A=[10,2,8,4,6,1,2,2,5]$
- Write down the partitions p , r , A at the end of first and second iterations
- Answer
 - 0 8 [2, 4, 1, 2, 2, 5, 10, 6, 8]
 - 0 4 [2, 1, 2, 2, 4, 5, 10, 6, 8]
 - 6 8 [1, 2, 2, 2, 4, 5, 6, 8, 10]



Selection Problem

- Given n elements from a totally ordered universe, find the k^{th} smallest
- Minimum: $k=1$, maximum: $k=n$
- Median: $k=\lfloor (n+1)/2 \rfloor$
- $O(n)$ comparisons for min or max
- $O(n \log n)$ comparisons by sorting
- $O(n \log k)$ comparisons with a **binary heap**



Selection Problem

- The input may or may not be ordered
 - It is orderable since drawn from an ordered universe
- Also named as i^{th} order statistic
- Applications of Selection
 - Pivot selection by median can be used for Quicksort
 - Incremental sorting
 - Order statistics



Recall Binary Heap

- A binary heap is a **complete** binary tree which satisfies the heap ordering property
 - min-heap: the value of each node is greater than or equal to the value of its parent, minimum-value element at the root
 - max-heap: the value of each node is less than or equal to the value of its parent, maximum-value element at the root
- Insertion $O(\log n)$
- Extraction $O(\log n)$
- Application: priority queue



Quickselect

- Idea: 3-way partition such that
 - smaller, equal and larger items in L, M and R, respectively
 - then recur in one of the subarrays - the one containing the kth smallest element
- Pick pivot $p \in A$ uniformly at random



Quickselect

```
def quickselect(A,k):
    p = A[int(len(A)/2)] # Pick the middle
    L,M,R = partition3way(A,p) # (n-1) comparisons
    if k<=len(L):
        return quickselect(L,k)
    elif k>len(L)+len(M):
        return quickselect(R,k-len(L)-len(M))
    else:
        return p
def partition3way(A,p):
    L,M,R=[],[],[]
    for a in A: # Can be improved like quicksort
        if p>a: L.append(a)
        elif p<a: R.append(a)
        else: M.append(a)
    return L,M,R
```



Analysis of Quickselect

- Without losing generality, assume distinct elements
- Candy bar split problem
 - After a "fair" split the average length of the longest bar is $3/4$
 - Expected value $E(x) = \text{integral } 1/2 \text{ to } 1 \text{ with uniform distribution}$
 - $E(x) = \int_{\frac{1}{2}}^1 xf(x)dx, f(x) = \begin{cases} 0, & \text{else} \\ 2, & \frac{1}{2} \leq x \leq 1 \end{cases}$
 - $E(x) = 3/4$
- $T(n) = T(3n/4) + O(n)$



Analysis of Quickselect

- Normal case: $T(n) = T(3n/4) + O(n)$
 - $T(n) \leq 4n$
 - Linear algorithm (!)
- Unlucky case: $T(n) = T(n-1) + O(n)$
 - $T(n) \leq n^2$
 - More costly than $O(n \log n)$ - sorting



Deterministic Selection

- Median of medians
 1. Group the array into $n/5$ groups of size 5 and find the median of each group
 2. Recursively, find the true median of the medians, call this p
 3. Use p as a pivot to split the array into subarrays L and R
 4. Recurse on the appropriate subarray



Deterministic Selection Analysis

- $O(n)$ for step 1
- $T(n/5)$ for step 2
- $3/10$ subarray $\leq p$, $3/10$ subarray $\geq p$
- $T(n) \leq cn + T(n/5) + T(7n/10)$
- Then, $T(n) \leq cn + T(9n/10)$
- Finally, $T(n) \leq 10n$
- Thus, the cost increases for the sake of determinism



Deterministic Selection Analysis

