Algorithms: Solutions 5

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number				X		Х	
homewor				X		X	
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					X		X
			X		X		Х
			X		X		Х
			X	X	X		X
			X	X	X		Х
Х	Х		X	X	X		Х
Х	Х		X	Х	X	X	Х
3	4	5	6	7	8	9	10
grades							
5							

Problem 1

Write an algorithm for finding the kth smallest element of a given array.

The following algorithm uses the same Partition procedure as Quick-Sort. The average-case time of this algorithm is $\Theta(n)$, but the worst case is $\Theta(n^2)$.

```
\begin{split} & \text{Select}(A, n, k) \\ & p \leftarrow 1 \\ & r \leftarrow n \\ & \textbf{while } p < r \\ & \textbf{do } q = \text{Partition}(A, p, r) \\ & \textbf{if } k \leq q \\ & \textbf{then } r \leftarrow q \\ & \textbf{else } p \leftarrow q + 1 \\ & \textbf{return } A[k] \end{split}
```

You may find a more sophisticated algorithm in Section 10.3 of the textbook; the worst-case time of the textbook algorithm is $\Theta(n)$.

Problem 2

Consider a computer environment where the control flow of a program can split three ways after a single comparison $a_i : a_j$, according to whether $a_i < a_j$, $a_i = a_j$, or $a_i > a_j$. Argue that the number of these three-way comparisons required to sort an *n*-element array is $\Omega(n \lg n)$.

Suppose that all numbers in the array are distinct. Then, a comparison sort never encounters the " $a_i = a_j$ " situation, and we may represent its control flow by a binary decision tree. The height of this tree is $\Omega(n \lg n)$; hence, the worst-case complexity of sorting an array of distinct numbers is $\Omega(n \lg n)$. Therefore, the complexity of sorting an arbitrary array is also $\Omega(n \lg n)$.

Problem 3

Suppose that A[1..n] is an array of integer numbers, and some value k occurs at least $\lfloor n/2 \rfloor + 1$ times in this array. Write an efficient algorithm for finding this value.

The "frequent" element is the median of the array, that is, it is the $\lfloor n/2 \rfloor$ -th smallest element. We can find it using the Select algorithm from Problem 1, with $k = \lfloor n/2 \rfloor$.

Problem 4

We consider an array A[1..n] and define a segment sum from p to r as follows:

$$sum(p, r) = \sum_{p \le i \le r} A[i].$$

Write a linear-time algorithm that determines the maximum over all segment sums.

```
\begin{aligned} \operatorname{Max-Segment}(A,n) \\ \operatorname{Local-Max} &\leftarrow 0 \\ \operatorname{Global-Max} &\leftarrow 0 \\ \operatorname{for} i &\leftarrow 1 \text{ to } n \\ \operatorname{do} \operatorname{Local-Max} &\leftarrow \max(A[i], \operatorname{Local-Max} + A[i]) \\ & \rhd \operatorname{Local-Max} \text{ is the maximum over the segments whose last element is } A[i]. \\ \operatorname{Global-Max} &\leftarrow \max(\operatorname{Local-Max}, \operatorname{Global-Max}) \\ & \rhd \operatorname{Global-Max} \text{ is the maximum over all segments in } A[1..i]. \end{aligned}
\operatorname{return} \operatorname{Global-Max}
```