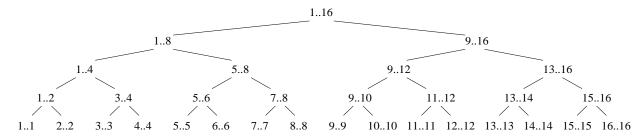
Analysis of Algorithms: Solutions 8

								X	
								X	
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								X	
						X		X	
number of						X		X	
homeworks		X	X		X	X	X	X	
		X	X		X	X	X	X	
		X	X	X	X	X	X	X	
	X 								
	1	2	3	4	5	6	7	8	9

The histogram shows the distribution of grades for the homeworks submitted on time.

Problem 1

Using Figure 16.2 in the textbook as a model, draw the recursion tree for the Merge-Sort procedure on a sixteen-element array. Explain why dynamic programming is ineffective for speeding up Merge-Sort.



The MERGE-SORT procedure does *not* have overlapping subproblems, that is, all nodes of the recursion tree are distinct. We cannot re-use the results of recursive calls and, hence, dynamic programming does not improve the efficiency.

Problem 2

Determine a longest common subsequence of $\langle 1, 0, 0, 1, 0, 1 \rangle$ and $\langle 0, 1, 0, 1, 1, 0, 1, 1 \rangle$. Using Figure 16.3 in the book as a model, draw the table constructed by the LCS-LENGTH algorithm for these two sequences (you do *not* need to show arrows in your table).

The longest common subsequences are $\langle 0, 0, 1, 0, 1 \rangle$, $\langle 1, 0, 1, 0, 1 \rangle$, and $\langle 1, 0, 0, 1, 1 \rangle$. The table computed by LCS-LENGTH is as follows:

```
6 | 0 1 2 3 4 4 4 5 5 5 5 | 0 1 2 3 3 3 4 4 4 4 4 4 4 4 4 4 4 4 4 5 5 5 6 7 8
```

Problem 3

Write an algorithm Increasing-Length (A, n) that determines the *length* of a longest increasing subsequence of an array A[n].

Let d[i] denote the length of a longest increasing subsequence whose last element is A[i]. For example, suppose that $A = \langle 1, 2, 1, 2, 3 \rangle$ and i = 3; then, the longest increasing subsequence that ends at A[3] is $\langle 1, 1 \rangle$, which implies that d[3] = 2.

Observe that this subsequence contains at least one element and, hence, $d[i] \geq 1$. Furthermore, if j < i and $A[j] \leq A[i]$, then $d[i] \geq d[j] + 1$, because we may construct a (d[j] + 1)-element increasing subsequence that ends at A[i], by adding A[i] to the longest increasing subsequence that ends at A[j]. These observations lead to the following algorithm, whose running time is $\Theta(n^2)$:

```
\begin{split} \text{Increasing-Length}(A,n) \\ d\text{-}max &\leftarrow 0 \\ \textbf{for } i \leftarrow 1 \textbf{ to } n \\ & \textbf{do } d[i] \leftarrow 1 \\ & \textbf{ for } j \leftarrow 1 \textbf{ to } i-1 \\ & \textbf{ do if } A[i] \geq A[j] \text{ and } d[i] < d[j]+1 \\ & \textbf{ then } d[i] \leftarrow d[j]+1 \\ & \textbf{ if } d\text{-}max < d[i] \\ & \textbf{ then } d\text{-}max \leftarrow d[i] \\ & \textbf{ return } d\text{-}max \end{split}
```