Algorithms: Solutions 10

	grades								
	2	3	4	5	6	7	8	9	10
	X 	X 	X 	X 	X 	X 	X 		X
	X		X	X	X	X			X
			X	X	X	X			
			X	X	X	X			
			X	X	X	X			
			X	X	X	X			
			X	X	X	X			
homewor		X	X	X					
number	of			X	X				
				X	X				
				X	X				
				X	X				
				X	X				
				X	X				
				X					
				X					
				X					

Problem 1

Assume that all characters in a pattern P[1..m] are distinct, and you need to find all occurrences of P in a text T[1..n]. Write an "accelerated" version of NAIVE-STRING-MATCHER, which solves this problem in O(n) time.

```
\begin{aligned} \text{FAST-Naive-Matcher}(T,P,n,m) \\ \text{initialize empty list } Shifts \\ i \leftarrow 1 \\ \textbf{while } s \leq n-m \\ \textbf{do if } i > m \\ \textbf{then add } s \text{ to } Shifts \\ s \leftarrow s+m \\ i \leftarrow 1 \\ \textbf{else if } P[i] = T[s+i] \\ \textbf{then } i \leftarrow i+1 \\ \textbf{else } s \leftarrow s+i \\ i \leftarrow 1 \end{aligned}
```

return Shifts

Problem 2

Write an algorithm that looks for a given $m \times m$ pattern in an $n \times n$ array of characters, based on the Rabin-Karp method.

We compute a separate Rabin-Karp numerical value for each row of the $m \times m$ pattern, and then compute numerical values for each row of the $n \times n$ array. For each possible position of the pattern in the array, we need to compare m numerical values, which correspond to the rows of the pattern. When all values match, we perform a character-by-character comparison of the pattern with the array.

Problem 3

Design an efficient algorithm for finding the longest common substring of two strings.

We denote the input strings by X[1..m] and Y[1..n], and define l[i,j] as the length of the longest common suffix of X[1..i] and Y[1..j]. We compute l[i,j] for every $i \leq m$ and every $j \leq n$; the maximal l[i,j] value is the length of the longest common substring.

The following algorithm finds the maximal l[i, j] value, uses it to identify the longest common substring, and prints out this substring. The algorithm runs in $\Theta(m \cdot n)$ time and requires $\Theta(m \cdot n)$ memory.

```
COMMON-SUBSTRING(X, Y, m, n)
i_{max} \leftarrow 0
l_{max} \leftarrow 0
for i \leftarrow 0 to m
     do l[i,0] \leftarrow 0
for j \leftarrow 1 to n
     do l[0,j] \leftarrow 0
for i \leftarrow 1 to m
     do for j \leftarrow 1 to n
              do if X[i] \neq Y[j]
                       then l[i,j] \leftarrow 0
                        else l[i, j] \leftarrow l[i - 1, j - 1] + 1
                               if l[i, j] > l_{max}
                                   then i_{max} \leftarrow i
                                            l_{max} \leftarrow l[i, j]
print X[(i_{max} - l_{max} + 1)..(i_{max})]
```

We can modify this algorithm to reduce its memory usage, without affecting the running time; the modified version uses an auxiliary array l[1..n], which takes only $\Theta(n)$ memory.

```
LOW-MEMORY-SUBSTRING(X, Y, m, n)
i_{max} \leftarrow 0
l_{max} \leftarrow 0
for j \leftarrow 1 to n
     do l[j] \leftarrow 0
for i \leftarrow 1 to m
     do old \leftarrow 0
           for j \leftarrow 1 to n
                 do temp \leftarrow l[j]
                      if X[i] \neq Y[j]
                         then l[j] \leftarrow 0
                         else l[j] \leftarrow old + 1
                                 if l[j] > l_{max}
                                     then i_{max} \leftarrow i
                                             l_{max} \leftarrow l[j]
                      old \leftarrow temp
print X[(i_{max} - l_{max} + 1)..(i_{max})]
```

Note that this algorithm for finding the longest common substring is *not* optimal. We can solve this problem in O(m+n) time, using the concept of a *suffix tree*.