Recursion

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This lecture will introduce a technique called recursion for designing algorithms. Its principle is:

When dealing with a subproblem (same problem but with a smaller input), consider it solved, and use the subproblem’s output to continue the algorithm design.

We will apply the technique to settle several problems in this course. Today, we will see two examples. In the first, we will re-discover binary search; in the second, we will design our first sorting algorithm.
An **array** of length $n$ is a sequence of $n$ elements such that

- they are stored consecutively in memory (i.e., the first element is immediately followed by the second, and then by the third, and so on);
- every element occupies the same number of memory cells.

![Array diagram]

<table>
<thead>
<tr>
<th>5</th>
<th>9</th>
<th>12</th>
<th>17</th>
<th>26</th>
<th>28</th>
<th>35</th>
<th>38</th>
<th>41</th>
<th>47</th>
<th>52</th>
<th>68</th>
<th>69</th>
<th>72</th>
<th>83</th>
<th>88</th>
</tr>
</thead>
</table>

array of length 16

array of length 12
With the concept of array, we now redefine the dictionary search problem:

**The Dictionary Search Problem (Redefined)**

**Problem Input:**
A set $S$ of $n$ integers has been arranged in ascending order in an array of length $n$. You are given the value of $n$ and another integer $v$ inside the CPU.

**Goal:**
Design an algorithm to determine whether $v$ exists in $S$. 
**Binary Search (Re-discovered)**

1. Compare \( v \) to the middle element \( e \) of the array. If \( v = e \), return “yes” and done.

2. Otherwise:
   2.1 If \( v < e \), we have a **subproblem**: check if \( v \) is in the portion of the array before \( e \);
   2.2 If \( v > e \), we have a **subproblem**: check if \( v \) is in the portion of the array after \( e \).

Considering the subproblem solved, we finish the algorithm.

**Think:** why does it work?
Recursion allows us to analyze the running time in an elegant manner.

Define $f(n)$ to be the maximum running time of binary search on $n$ elements. For $n = 1$, clearly:

$$f(1) = O(1)$$

For $n > 1$:

$$f(n) \leq O(1) + f(\lfloor n/2 \rfloor).$$
Analysis of Binary Search

So it remains to solve the recurrence \((c_1, c_2 \text{ are constants whose values we do not care})\):

\[
\begin{align*}
    f(1) & = c_1 \\
    f(n) & \leq c_2 + f([n/2])
\end{align*}
\]

Suppose, for now, that \(n\) is a power of 2. An easy way of doing so is the expansion method, which simply expands \(f(n)\) all the way down:

\[
\begin{align*}
    f(n) & \leq c_2 + f(n/2) \\
    & \leq c_2 + c_2 + f(n/2^2) \\
    & \leq c_2 + c_2 + c_2 + f(n/2^3) \\
    & \leq c_2 + \ldots + c_2 + f(1) \\
    & \leq c_2 \cdot \log_2 n \text{ of them} \\
    & = c_2 \cdot \log_2 n + c_1 = O(\log n).
\end{align*}
\]
Analysis of Binary Search

We can deal with general $n$ (not necessarily a power of 2) using a rounding approach. Let $n'$ be the least power of 2 that is larger than $n$. It thus holds that $n' < 2n$ (otherwise, $n'$ is not the least).

We then have:

\[
\begin{align*}
    f(n) & \leq f(n') \\
         & \leq c_2 \cdot \log_2 n' + c_1 \text{ (proved earlier)} \\
         & < c_2 \cdot \log_2 (2n) + c_1 \\
         & = c_2 (1 + \log_2 n) + c_1 \\
         & = c_2 \log_2 n + c_1 + c_2 = O(\log n).
\end{align*}
\]
Next, we switch our attention to the sorting problem, another classical problem in computer science.
The Sorting Problem

Problem Input:

A set $S$ of $n$ integers is given in an array of length $n$. The value of $n$ is inside the CPU (i.e., in a register).

Goal:

Produce an array that stores the elements of $S$ in ascending order.
Example

Input:

\[
\begin{array}{cccccccc}
16 & & & & & & & \\
\ldots & & & & & & & \\
\end{array}
\]

Output:

\[
\begin{array}{cccccccc}
16 & & & & & & & \\
\ldots & & & & & & & \\
\end{array}
\]
Selection Sort

1. Find the largest integer \( e_{\text{max}} \) in \( S \).

2. Swap \( e_{\text{max}} \) with the last (i.e., \( n \)-th) element of the array (after which \( e_{\text{max}} \) is at the end of the array).

3. We now have a subproblem: sort the first \( n - 1 \) elements.

Let us consider that the subproblem has been solved. Now, the entire array is in ascending order. We thus finish the algorithm.
Input:

After Step 2:

sort these 15 elements recursively
Analysis of Selection Sort

Let $f(n)$ be the maximum running time of selection sort when the problem size is $n$. We know:

$$f(1) = O(1)$$

For $n \geq 2$, we have:

$$f(n) \leq O(n) + f(n-1)$$

where the term $O(n)$ captures the cost of Steps 1 and 2, and $f(n-1)$ is the cost of Step 3.
Analysis of Selection Sort

So it remains to solve the recurrence ($c_1, c_2$ are constants):

\[
\begin{align*}
  f(1) &= c_1 \\
  f(n) &\leq c_2 n + f(n - 1)
\end{align*}
\]

Using the expansion method, we get:

\[
\begin{align*}
  f(n) &\leq c_2 n + f(n - 1) \\
  &\leq c_2 n + c_2(n - 1) + f(n - 2) \\
  &\leq c_2 n + c_2(n - 1) + c_2(n - 2) + f(n - 3) \\
  &\leq c_2 n + c_2(n - 1) + \ldots + c_2 \cdot 2 + f(1) \\
  &\leq c_2 n(n + 1)/2 + c_1 \\
  &= O(n^2).
\end{align*}
\]

We now conclude that selection sort runs in $O(n^2)$ worst-case time.