# Chapter 2. Getting Started

```
2.1-1.
A = \{31, 41, 59, 26, 41, 58\}
\Rightarrow A = \{26, 31, 41, 59, 41, 58\}
\Rightarrow A = \{26, 31, 41, 41, 59, 58\}
\Rightarrow A = \{26, 31, 41, 41, 58, 59\}
2.1-2.
  Algorithm 1: Insertion sort(A), nonincreasing order
   for j = 2 to A.length do
       \text{key} = A[j];
       // Insert A[j] into the sorted sequence A[1, ..., j - 1];
       while i = 0 and A[i] < key do
           A[i+1] = A[i];
         i = i - 1;
       end
       A[i+1] = key;
   end
2.1 - 3.
  Algorithm 2: Linear search(A, \nu)
   for j = 1 to A.length do
       if A[j] = \nu then
        | return j
       end
   end
   return NIL
```

At the start of each iteration of the for loop, the subarray  $A[1,\cdots,j-1]$ 

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The loop invariant of this algorithm is as follows:

does not contain  $\nu$ . The three necessary properties of the loop invariant can be proved as:

**Initialization**: This is trivial, since nonempty array does not contain anything.

**Maintenance**: Before the loop, the subarray  $A[1, \dots, j-1]$  does not contain  $\nu$ . If  $A[j] = \nu$ , the algorithm returns and the loop breaks. Otherwise,  $A[1, \dots, j]$  does not contain  $\nu$ , maintaining the loop invariant.

**Termination**: Finally, once the loop is completed, A does not contain  $\nu$ , so NIL should be returned.

#### 2.1 - 4.

#### **Algorithm 3:** Add two binary integers(A, B)

#### 2.2-1.

```
n^3/1000 - 100n^2 - 100n + 3 = \Theta(n^3).
```

#### 2.2 - 2.

### **Algorithm 4:** Selection sort(A)

```
\begin{array}{l} \mathbf{for} \ i = 1 \ to \ A.length - 1 \ \mathbf{do} \\ & \min = \infty, \ \mathrm{index} = \mathrm{i} + 1; \\ & \mathbf{for} \ j = i + 1 \ to \ A.length \ \mathbf{do} \\ & | \ \mathbf{if} \ min > A[j] \ \mathbf{then} \\ & | \ \min = \mathrm{A[j], \ index} = \mathrm{j}; \\ & | \ \mathbf{end} \\ & | \ \mathrm{end} \\ & | \ \mathrm{swap}(\mathrm{A[j], \ A[index]}); \\ & \mathbf{end} \end{array}
```

The loop invariant of selection sort is that, at the start of each iteration of the for loop, the subarray  $A[1, \dots, j-1]$  consists of the least j-1 elements

of A. It doesn't need to run for all n elements since after n-1 iteration it will be already guaranteed that A[n] is larger than any other elements in  $A[1, \dots, n-1]$ . The performance for the best case and the worst case are both  $\Theta(n^2)$ .

#### 2.2 - 3.

n/2 elements are needed to be checked on the average. In worst case n checks are required. They're both  $\Theta(n)$ .

# 2.2 - 4.

Making loop invariant to satisfy algorithm requirement would decrease the best-case running time.

#### 2.3-1.

 $A = \{3, 41, 52, 26, 38, 57, 9, 49\}$   $\Rightarrow A = \{3, 41, 26, 52, 38, 57, 9, 49\}$   $\Rightarrow A = \{3, 26, 41, 52, 9, 38, 49, 57\}$   $\Rightarrow A = \{3, 9, 26, 38, 41, 49, 52, 57\}$ 

2.3-2.

# **Algorithm 5:** Merge(A, p, q, r)

```
n_1 = q - p + 1;
n_2 = r - q;
let L[1, \dots, n_1] and R[1, \dots, n_1] be new arrays;
for i = 1 to n_1 do
L[i] = A[p + i - 1];
end
for j = 1 to n_2 do
R[i] = A[q + j];
end
i = 1, j = 1;
for k = p to r do
   if L[i] < R[j] or j = n_2 + 1 then
       A[k] = L[i];
       i = i + 1;
   else
       A[k] = R[j];
      j = j + 1;
   \quad \text{end} \quad
end
```

#### 2.3-3.

For n = 2, T(n) = 2.

Suppose the statement holds for  $n \leq 2^k$ . Then for  $n = 2^{k+1}$ ,  $T(n) = 2 \cdot (2^k \cdot k) + 2^{k+1} = 2^{k+1} \cdot (k+1)$ .

#### 2.3-4.

The recurrence formula for the worst-case running time of the modified version of insertion sort is T(n) = T(n-1) + (n-1).

2.3-5.

#### **Algorithm 6:** BinarySearch(A, p, q, $\nu$ )

```
if p > q then | return; m = (p+q)/2; if A[m] > \nu then | BinarySearch(A, p, m - 1, \nu); else if A[m] < \nu then | BinarySearch(A, m+1, n, \nu); else | return m; end
```

The worst-case running time of binary search is  $\Theta(\lg n)$ , because for each recursion the range being searched is halved.

#### 2.3-6.

No, the number of swaps required for each insertion makes the worst-case running time still  $\Theta(n^2)$ .

#### 2.3-7.

# Algorithm 7: TwoSum(S, x)

```
MergeSort(S);
left = 1, right = S.length;
while <math>left < right \text{ do}
| if S[left] + S[right] < x \text{ then}
| left = left + 1;
else if S[left] + S[right] > x \text{ then}
| right = right - 1;
else
| return true;
end
end
return false;
```

The worst-case running time of this algorithm is  $\Theta(n \lg n) + \Theta(n) = \Theta(n \lg n)$ .

```
2-1. a. \Theta(k^2) \cdot n/k = \Theta(nk).
```

- b. n/k sublists can be merged by merging the two adjacent sublists recursively; The time complexity becomes  $\Theta(n \lg n/k)$ .
- c.  $nk + n \lg(n/k) = n \lg n \Rightarrow k = \lg k$ , hence the largest value is k = 1.
- d. In practice, analyzing operation count is necessary to decide the optimal value of k.

2-2.

a. We need to prove that  $A'[1], \dots, A'[n]$  is an rearrangement of  $A[1], \dots, A[n]$ . b. The loop invariant is that, at the start of each iteration of the outer inner loop, A[j] is smallest among  $A[j, \dots, n]$ . The three necessary properties of the loop invariant can be proved as:

**Initialization**: This is trivial, since A[n] is smallest among A[n].

**Maintenance**: In the inner loop, if A[j] is smallest among  $A[j, \dots, n]$ , the element is not swapped and the loop invariant is maintained. Otherwise, A[j] and A[j+1] is swapped so that A[j+1] is placed in A[j], maintaining the loop invariant.

**Termination**: Finally, once the loop is completed, A[i+1] is the smallest element of  $A[i+1,\dots,n]$ .

c. The loop invariant is that, at the start of each iteration of the outer for loop,  $A[1, \dots, i-1]$  consists of the least i member among  $A[1, \dots, n]$ , insortedorder. The three necessary properties of the loop invariant can be proved as:

**Initialization**: This is trivial, since nonempty array does not contain anything.

**Maintenance**: In the inner loop, the smallest element in  $A[i+1,\dots,n]$  is repeatedly swapped so that it is placed in A[i], maintaining the loop invariant.

**Termination**: Finally, once the loop is completed,  $A[1, \dots, n]$  is sorted. d.  $\Theta(n^2)$ , same with the running time of insertion sort.

2.3.

a.  $\Theta(n)$ .

#### **Algorithm 8:** Naive polynomial evaluation(P)

```
y = 0;
\mathbf{for} \ i = n \ downto \ 0 \ \mathbf{do}
\begin{vmatrix} term = a_n; \\ \mathbf{for} \ j = 1 \ to \ n \ \mathbf{do} \\ | \ term = term \cdot x; \\ \mathbf{end} \\ y = y + term; \\ \mathbf{end} \\ \mathbf{return} \ y; \end{vmatrix}
```

The running time of this algorithm is  $\Theta(n^2)$ , significantly worse than Horner's rule.

- c. At termination, i = -1, so plugging i in the formula gives  $y = \sum_{k=0}^{n} a_k x^k$ .
- d. Direct calculation verifies correctness of the algorithm.

```
2.4.
```

- a. (2,3), (0,4), (1,4), (2,4), (3,4).
- b.  $\{n, n-1, \dots, 1\}$ . This have n(n-1)/2 inversions.
- c. The running time of insertion sort is proportional to number of inversions, because it is number of total swaps required.
- d. The number of inversion can be simply calculated by tweaking merge sort:

# **Algorithm 9:** MergeSort(A, p, r)

# Algorithm 10: Merge(A, p, q, r)

```
n_1 = q - p + 1;
n_2 = r - q;
count = 0;
let L[1, \dots, n_1] and R[1, \dots, n_1] be new arrays;
for i = 1 to n_1 do
L[i] = A[p + i - 1];
end
for j = 1 to n_2 do
R[i] = A[q + j];
\mathbf{end}
i = 1, j = 1;
for k = p to r do
   if L[i] < R[j] or j = n_2 + 1 then
      A[k] = L[i];
      i = i + 1;
   else
      A[k] = R[j];
      j = j + 1;
       count = count + 1;
   end
end
return count;
```

The algorithm runs in  $\Theta(n \lg n)$  time.