

Input: A sequence of n numbers $A = \langle a_1, \dots, a_n \rangle$ and a value v

Output: An index i s.t. $v = A[i]$ or NIL if v does not appear in A .

<Pseudo-code>

LinearSearch(A, v)

for $i = 1$ to $A.length$

if $A[i] == v$

return i .

return NIL.

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Algorithm

HWI.

"no collaborator"

< Use loop invariant to show at the start of each iteration of the for loop, the subarray.

$A[1, \dots, i-1]$ contains no " v " element >

① Initialization: Before the first loop, when $i = 1$, the subarray is empty.

Thus, the loop invariant holds.

② Maintenance: During each iteration, we compare $A[i]$ to v . If they are the same, we return i . Otherwise, the loop iteration continues. Thus, we know that at the end of each loop, the subarray $A[1, \dots, i]$ must not contain element v . Therefore, the loop invariant holds. Incrementing i for the next iteration of the for loop then preserves loop invariant.

③ Termination: There are two conditions that the algorithm terminates. First, when $A[i] == v$, from the maintenance property, we know that the subarray $A[1, \dots, i-1]$ contains no " v " element and we return i . Second, the condition causing the for loop to terminate is that $i = A.length = n$. Because we increment i by 1 in each loop iteration, we have $i = n+1$ at that time. Substituting $i = n+1$ into the sentence of loop invariant, we get that the subarray $A[1, \dots, n]$ contains no " v " element and we return NIL. Observing that $A[1, \dots, n]$ is the entire array, we conclude that $A[1, \dots, n]$ doesn't contain " v ". Hence, the algorithm is correct.

參考 "<https://walkccc.me/CLRS/Chap02/2.1/>" 與課本敘述.

2. The binary search algorithm halves the size of the remaining portion of the sequence and drops the unused portion, it yields 1 subproblem with $\frac{1}{2}$ the size of the original. Also, the divide action and the base case both take constant time. Thus, the recurrence form of time complexity is

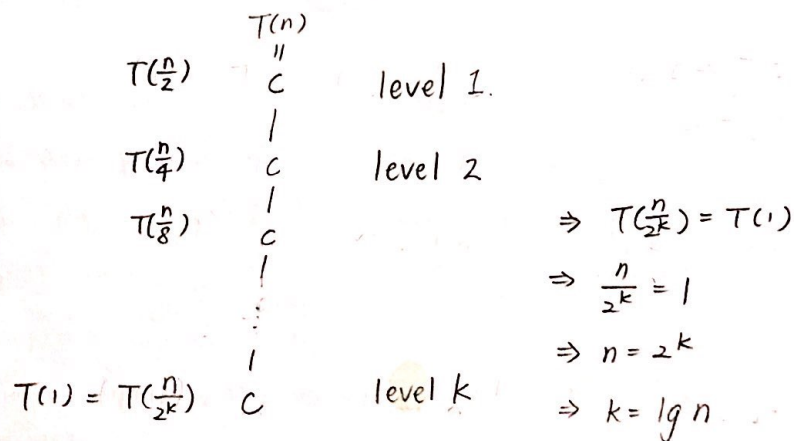
$$T(n) = \begin{cases} \Theta(1), & n=1 \\ T(\frac{n}{2}) + \Theta(1), & n>1 \end{cases}$$

↪ divide action

Assume that $C = \max$ (time of solving the problem when $n=1$, time of divide action)

$$T(n) = \begin{cases} C, & n=1 \\ T(\frac{n}{2}) + C, & n>1 \end{cases}$$

By using the recursion tree,



$$\Rightarrow T(n) = C \cdot \lg n$$

$$\Rightarrow T(n) = \Theta(\lg n)$$

3. 参考: <https://walkccc.me/CLRS/Chap02/Problems/2-2/>

- a. We need to show that A' contains the elements in A but in sorted order.
- b. for loop in lines 2-4, the loop invariant property is "at the start of each iteration of the for loop in lines 2-4, the subarray $A[j, \dots, n]$ contains the elements originally in $A[j, \dots, n]$ and the $A[j]$ element should be the smallest one in this subarray. Also, other elements may be moved into different position."

① Initialization: Before the first loop, $j = n$, the subarray contains only $A[n]$.

Thus, it is trivially the smallest element.

② Maintenance: During each iteration, we compare $A[j]$ with $A[j-1]$ and we move the smaller element to $A[j-1]$. Thus, after the iteration, the length of the subarray increases by one, and the first element is the smallest of the subarray.

③ Termination: The condition causing the for loop to terminate is that $j = 1$. Substituting it to the loop invariant, the subarray $A[1, \dots, n]$ contains the elements originally in $A[1, \dots, n]$ and $A[1]$ is the smallest element in the entire subarray.

7. c.

for loop in lines 1-4, the loop invariant property is that "at the start of each iteration of the for loop in lines 1-4, the subarray $A[1, \dots, i-1]$ contains the $i-1$ smallest elements in $A[1, \dots, n]$ in sorted order. Also, $A[i, \dots, n]$ contains the remaining $n-(i-1)$ elements in $A[1, \dots, n]$ that are larger than all elements in $A[1, \dots, i-1]$.

① Initialization: Before the first loop, when $i=1$, the subarray is empty. Thus, the loop invariant holds.

② Maintenance: At the beginning of the outer loop, the subarray $A[1, \dots, i-1]$ contains the $i-1$ elements in $A[1, \dots, n]$ in sorted order and $A[i, \dots, n]$ contains the elements greater than elements in $A[1, \dots, i-1]$. According to part b, after the inner loop, $A[i]$ will be the smallest element of the subarray $A[i, \dots, n]$. Thus, at the end of the outer loop (when the inner loop is finished), $A[1, \dots, i]$ contains the elements that are smaller than $A[i+1, \dots, n]$ in sorted order.

③ Termination: The condition causing the outer for loop to terminate is that $i=n$. Substituting into the loop invariant property, the array $A[1, \dots, n-1]$ contains the $n-1$ smallest elements in $A[1, \dots, n]$ in sorted order that are smaller than $A[n]$. Thus, the array $A[1, \dots, n]$ contains all elements in sorted order.

d. The worst-case of bubble sort happens when the entire array is in reverse order. In this situation, the outer loop will run $(n-1)$ iterations and the inner loop will run $[n-(i+1)+1] = (n-i)$ iterations. Thus the worst-case time complexity of bubble sort $= (n-1) + (n-2) + \dots + 1$.

$$= \frac{[1+(n-1)] \times (n-1)}{2} = \Theta(n^2), \text{ which is the same as the worst-case running time of}$$

insertion sort.

4. prove $\max(f(n), g(n)) = \Theta(f(n) + g(n))$ ^{参考} <https://atekthcan.github.io/CLRS103/E03.01-01/>

$\Leftrightarrow \exists C_1, C_2 > 0$ and $n_0 > 0$ s.t. $0 \leq C_1[f(n) + g(n)] \leq \max(f(n), g(n)) \leq C_2[f(n) + g(n)], \forall n \geq n_0$
^{for RHS} $f(n), g(n)$ are asymptotically nonnegative functions, we can assume that

for $n > n_1, f(n) > 0$ and for $n > n_2, g(n) > 0$.

choose $n_0 = \max(n_1, n_2)$, we know that $f(n) + g(n) \geq \max(f(n), g(n)) \forall n \geq n_0 = \max(n_1, n_2)$

^{for LHS} Also, $f(n) + g(n) \leq 2 \max(f(n), g(n)) \Leftrightarrow \frac{f(n) + g(n)}{2} \leq \max(f(n), g(n)), \forall n \geq n_0 = \max(n_1, n_2)$
 (since $\max(f(n), g(n))$ is either $f(n)$ or $g(n)$).

Combine the two results, we get $0 \leq \frac{f(n) + g(n)}{2} \leq \max(f(n), g(n)) \leq f(n) + g(n) \forall n \geq n_0$
 $n_0 = \max(n_1, n_2)$

Thus, $\max(f(n), g(n)) = \Theta(f(n) + g(n))$

5. By sterling's approximation, $n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \in \Theta(n^{n+\frac{1}{2}} \cdot e^{-n})$

$$(\log n)! \in \Theta((\log n)^{(\log n + \frac{1}{2})} e^{-\log n})$$

$$= \Theta((\log n)^{(\log n + \frac{1}{2})} n^{-\log e}) = \Theta(\log n \log n \frac{\log n^{\frac{1}{2}}}{n^{\log e}})$$

Some useful functions:

① $\lg^k n = (\lg n)^k$

④ $n^n > n! > a^n$

② $\lg \lg n = \lg(\lg n)$

③ $n^n \in \Omega(n!)$ ($n^n > n!$)

X 1. 2^{2^n}

X 2. $\log(n!) = \log 1 + \log 2 + \dots + \log n \leq n \cdot \log n$

$\log(n!) \in \Theta(n \log n)$

and $\log 1 + \dots + \log \frac{n}{2} + \dots + \log n \geq \log \frac{n}{2} + \dots + \log n$

$\geq \log \frac{n}{2} + \log \frac{n}{2} + \dots + \log \frac{n}{2}$

$= \frac{n}{2} \log \frac{n}{2} \quad (n \log n)$

X 3. $2^{2^{n+1}}$

X 4. $n^{\frac{1}{\log n}} = 2^{\log n \cdot \frac{1}{\log n}} = 2$

X 5. $2^{\sqrt{n}} = 2^{n^{\frac{1}{2}}} =$

X 6. $\log n$

X 7. $2^{2 \log n} = n^{2 \log 2} = n^2$

X 8. n^2

X 9. $n^{\frac{1}{2}}$

X 10. $n \log n$

X 11. 2^n

X 12. $(\log n)! > (\log n)^{(\log n + \frac{1}{2})} \cdot \frac{1}{n} = (\log n)^{\log n} \cdot \frac{(\log n)^{\frac{1}{2}}}{n}$

X 13. $n!$

X 14. $(\sqrt{2})^{\log n} = (2^{\frac{1}{2}})^{\log n} = (2^{\log n})^{\frac{1}{2}} = n^{\frac{1}{2}}$

X 15. 1

Ans: $2^{2^{n+1}} > 2^{2^n} > n! > 2^n > 2^{\sqrt{n}} > (\log n)! > n^2 = n^{\log n} >$

$n \log n = \log(n!) > n^{\frac{1}{2}} = (\sqrt{2})^{\log n} > \log n > n^{\frac{1}{\log n}} = 1$

6. 參考 "<https://walkccc.me/CLRS/Chap04/4.1/>"

iterativeFindMaximumSubarray(A)

$n = A.length$

$bestSum = -\infty$

$currentSum = -\infty$

for $j = 1$ to n

$currentEnd = j$

if $currentSum > 0$

$currentSum = currentSum + A[j]$

else

$currentBegin = j$

$currentSum = A[j]$

if $currentSum > bestSum$

$bestSum = currentSum$

$bestBegin = currentBegin$

$bestEnd = currentEnd$

return($bestBegin, bestEnd, bestSum$)

想三去: 如果現在的 $A[i], \dots, j+1$ 之和為負, 重新從 $j+1$ 開始計 maximum subarray, 因為新的 subarray 不可能加上一個負值使變得更大

Time complexity:

recursive: $\Theta(n \log n)$

iterative: $\Theta(n)$

7. (a) 参考 "https://ita.skanev.com/04/02/04.html"
 Suppose that we have $C = A \cdot B$, where A, B are two $n \times n$ matrices with $n = 3^m$.
 By Strassen's method, we can divide A, B, C into $\frac{n}{3} \times \frac{n}{3}$ matrix respectively. Then, we
 can compute each $\frac{n}{3} \times \frac{n}{3}$ submatrix in C by recursively compute k matrix products P_1, \dots, P_k .

$$\text{Thus, the recursive running time } T(n) = \begin{cases} \Theta(1), & \text{if } n=1 \\ k \cdot T(\frac{n}{3}) + \Theta(n^2), & \text{if } n>1. \end{cases}$$

By master theorem, if $n^2 = O(n^{\log_3 k - \epsilon})$, $T(n) = \Theta(n^{\log_3 k})$

Assuming $k \geq 10$ and $\epsilon = 0.01$, we get $T(n) = \Theta(n^{\log_3 k})$

$$\text{Thus, } T(n) = o(n^{\lg 7}) \Leftrightarrow n^{\log_3 k} < n^{\lg 7}$$

$$\Leftrightarrow \log_3 k < \lg 7$$

$$\Leftrightarrow \log k < \lg 7 \cdot \log 3$$

$$\Leftrightarrow k < e^{\lg 7 \cdot \log 3} = 1.849 \dots$$

\Rightarrow the largest value of k is ≥ 1 (which also satisfies our assumption) #

7. (b)

If $k = \geq 1$, the time complexity of this algorithm is $T(n) = \Theta(n^{\log_3 \geq 1})$ #

8. b.

$$T(n) = 3T(\frac{n}{3}) + \frac{n}{\lg n}$$

$$= 3 \left[3T(\frac{n}{9}) + \frac{n}{\lg n} \right] + \frac{n}{\lg n} = 9T(\frac{n}{9}) + \frac{n}{\lg n - \lg 3} + \frac{n}{\lg n}$$

$$= 9 \left[3T(\frac{n}{27}) + \frac{n}{\lg n} \right] + \frac{n}{\lg n - \lg 3} + \frac{n}{\lg n}$$

$$= 27T(\frac{n}{27}) + \frac{n}{\lg n - \lg 9} + \frac{n}{\lg n - \lg 3} + \frac{n}{\lg n}$$

$$= 3^{\log_3 n} T(1) + n \cdot \sum_{k=0}^{(\log_3 n)-1} \frac{1}{\lg(\frac{n}{3^k})}$$

$$= \Theta(n) + \Theta(n \log \log_3 n)$$

$$= \Theta(n \lg \lg n) \#$$

$$\times \sum_{k=0}^{\log_3 n - 1} \frac{1}{\lg(\frac{n}{3^k})} = \frac{1}{\lg n} + \frac{1}{\lg \frac{n}{3}} + \frac{1}{\lg \frac{n}{9}} + \dots + \frac{1}{\lg 3}$$

$$\text{also, } \lg n = \frac{\log_3 n}{\log_3 2} \Rightarrow \frac{1}{\lg n} = \frac{\log_3 2}{\log_3 n}$$

$$= \log_3 2 \left(\frac{1}{\log_3 n} + \frac{1}{\log_3 n - 1} + \frac{1}{\log_3 n - 2} + \dots + \frac{1}{\log_3 3} \right)$$

$$= \log_3 2 \left(\sum_{i=1}^{\log_3 n} \frac{1}{i} \right)$$

$$= \Theta(\log \log_3 n) = \Theta(\lg \lg n)$$

$$\downarrow \log_3 n = \frac{\lg n}{\lg 3} = \Theta(\lg n)$$

d.

$$T(n) = 3T(\frac{n}{3} - 2) + \frac{n}{2}$$

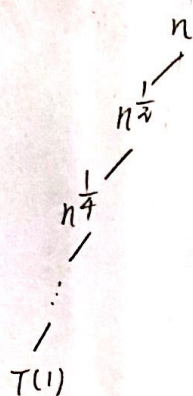
$$\text{when } n \text{ is large, } T(n) = 3T(\frac{n}{3}) + \frac{n}{2}$$

$$f(n) = \frac{n}{2} = \Theta(n^{\log_3 3}) = \Theta(n)$$

By master theorem, $T(n) = \Theta(n \lg n)$ #

J.

$$T(n) = \sqrt{n} T(\sqrt{n}) + n.$$



$$\begin{aligned} &\rightarrow n \times 1 \\ &\rightarrow n^{\frac{1}{2}} \times n^{\frac{1}{2}} = n \\ &\rightarrow n^{\frac{1}{4}} \times (n^{\frac{1}{4}} \times n^{\frac{1}{2}}) = n \\ &\vdots \\ &\rightarrow n. \end{aligned}$$

多/5: stack overflow,

"What would cause an algorithm to have $O(\log \log n)$ complexity?"

$$\text{height} = \lg \lg n.$$

*Height explanation: Since each time you take the square root of n , you halve the exponent in this equation. Therefore, there can be only $O(\log k)$ square roots applied before k drops to one or lower. Assume $n = 2^k$, $k = \log n$, $O(\log k) = O(\log \log n)$.

$$(k \text{ 最多能减 } \log k = k, n = 2^k \text{ 最多能减 } \log k = \log \log n = k)$$

$$\Rightarrow T(n) = \Theta(n \lg \lg n)$$

9.

多/5 "algorithm.cs.nthu.edu.tw/course/Extra-Info/Divide and Conquer - Supplement.pdf"

(a) Divide step:

Assume n is the power of 2

$$\text{Let } A_0(x) = a_0 + a_1 x + \dots + a_{\frac{n}{2}-1} x^{\frac{n}{2}-1}$$

$$A_1(x) = a_{\frac{n}{2}} + a_{\frac{n}{2}+1} x + \dots + a_n x^{\frac{n}{2}}$$

$$A(x) = A_0(x) + A_1(x) x^{\frac{n}{2}}$$

$$B_0(x) = b_0 + b_1 x + \dots + b_{\frac{n}{2}-1} x^{\frac{n}{2}-1}$$

$$B_1(x) = b_{\frac{n}{2}} + b_{\frac{n}{2}+1} x + \dots + b_n x^{\frac{n}{2}}$$

$$B(x) = B_0(x) + B_1(x) x^{\frac{n}{2}}$$

$$A(x)B(x) = A_0(x)B_0(x) + A_0(x)B_1(x)x^{\frac{n}{2}} + A_1(x)B_0(x)x^{\frac{n}{2}} + A_1(x)B_1(x)x^n$$

(The original problem of size n is divided into 4 problems of size $\frac{n}{2}$)

Conquer step:

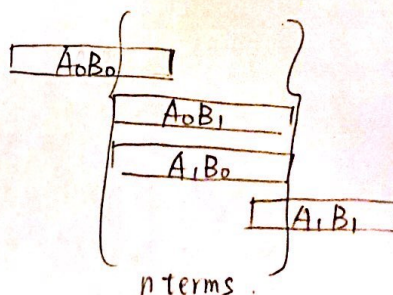
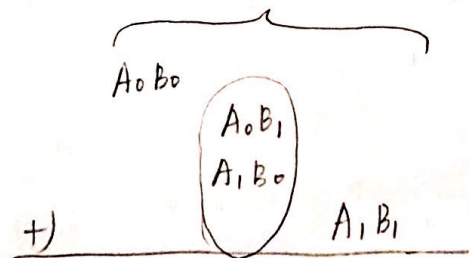
compute $A_0(x)B_0(x)$, $A_0(x)B_1(x)$, $A_1(x)B_0(x)$, $A_1(x)B_1(x)$ by recursively calling the algorithm 4 times.

Combine step:

We do the following calculation

$$A_0(x)B_0(x) + A_0(x)B_1(x)x^{\frac{n}{2}} + A_1(x)B_0(x)x^{\frac{n}{2}} + A_1(x)B_1(x)x^n$$

there are $(2n+1)$ terms to be added $\Rightarrow \Theta(n)$ time



Pseudocode:

PolyMultiply($A(x), B(x)$):

$$A_0(x) = a_0 + a_1 x + \dots + a_{\frac{n}{2}-1} x^{\frac{n}{2}-1}$$

$$A_1(x) = a_{\frac{n}{2}} + a_{\frac{n}{2}+1} x + \dots + a_n x^{\frac{n}{2}-1}$$

$$B_0(x) = b_0 + b_1 x + \dots + b_{\frac{n}{2}-1} x^{\frac{n}{2}-1}$$

$$B_1(x) = b_{\frac{n}{2}} + b_{\frac{n}{2}+1} x + \dots + b_n x^{\frac{n}{2}-1}$$

$$U(x) = \text{PolyMultiply}(A_0(x), B_0(x))$$

$$V(x) = \text{PolyMultiply}(A_0(x), B_1(x))$$

$$W(x) = \text{PolyMultiply}(A_1(x), B_0(x))$$

$$Z(x) = \text{PolyMultiply}(A_1(x), B_1(x))$$

$$\text{return } (U(x) + [V(x) + W(x)] x^{\frac{n}{2}} + Z(x) x^n).$$

(b)

$$T(n) = \begin{cases} 4T(\frac{n}{2}) + (2n+1), & \text{if } n > 1 \quad (\text{ignore cost of divide here, since it's miscellaneous compared to } (2n+1)) \\ 1, & \text{if } n = 1 \end{cases}$$

$$f(n) = 2n+1 = O(n^{\log_2 4 - \epsilon}) = O(n^{2-\epsilon})$$

$$\text{choose } \epsilon = 0.1$$

$$\text{By master theorem, } T(n) = \Theta(n^2) \#$$

(c)

In fact, we only need 3 terms in the conquer steps.

$$(A_0 B_0, A_0 B_1 + A_1 B_0, A_1 B_1)$$

This can be obtained by using 3 multiplications

$$Y = (A_0 + A_1)(B_0 + B_1)$$

$$U = A_0 B_0$$

$$Z = A_1 B_1$$

$$\text{we can get } A_0 B_1 + A_1 B_0 = Y - U - Z$$

(d)

PolyMultiplyBetter($A(x), B(x)$):

$$A_0(x) = a_0 + a_1 x + \dots + a_{\frac{n}{2}-1} x^{\frac{n}{2}-1}$$

$$A_1(x) = a_{\frac{n}{2}} + a_{\frac{n}{2}+1} x + \dots + a_n x^{\frac{n}{2}-1}$$

$$B_0(x) = b_0 + b_1 x + \dots + b_{\frac{n}{2}-1} x^{\frac{n}{2}-1}$$

$$B_1(x) = b_{\frac{n}{2}} + b_{\frac{n}{2}+1} x + \dots + b_n x^{\frac{n}{2}-1}$$

$$Y(x) = \text{PolyMultiplyBetter}(A_0(x) + A_1(x), B_0(x) + B_1(x))$$

$$U(x) = \text{PolyMultiplyBetter}(A_0(x), B_0(x))$$

$$Z(x) = \text{PolyMultiplyBetter}(A_1(x), B_1(x))$$

$$\text{return } (U(x) + [Y(x) - U(x) - Z(x)] x^{\frac{n}{2}} + Z(x) x^n)$$

(e)

$$T(n) = \begin{cases} 3T(\frac{n}{2}) + (2n+1), & \text{if } n > 1 \\ 1, & \text{if } n = 1 \end{cases}$$

$$f(n) = 2n+1 = O(n^{\log_2 3 - \epsilon})$$

$$\text{choose } \epsilon = 0.01 \quad (\frac{\log 3}{\log 2} = 1.585)$$

$$\text{By master theorem, } T(n) = \Theta(n^{\log_2 3})$$

$$= \Theta(n^{1.585}) \#$$