

1) a)

algo1(L[0...n-1])

1 if (n==1) return L[0]

else

T(n-1) = tmp = algo1(L[0...n-2])

1 { if (tmp <= L[n-1]) return tmp
else return L[n-1]

$$T(n) = \begin{cases} 1 & ; n=1; \\ T(n-1)+1; & n>1; \end{cases}$$

$$T(n) = T(n-1) + 1$$

$$T(n-1) = T(n-2) + 2$$

$$T(n-2) = T(n-3) + 3$$

=>

$$T(n) = T(n-k) + k$$

assume $n-k=1$

$$n=k+1$$

$$T(n) = \underbrace{T(1)}_1 + n - 1$$

$$= \boxed{T(n) = n}$$

b)

algo2(x[l...r])

1 { if (l==r) return x[l]

else

1 { flr = floor((l+r)/2)

T(n/2) { tmp1 = algo2(x[l...flr])

T(n/2) { tmp2 = algo2(x[flr+1...r])

1 { if (tmp1 <= tmp2) return tmp1
else return tmp2

$$T(n) = 2T(n/2) + 1$$

$$a=2$$

$$b=2$$

$$d=0$$

$$2 > 2^0$$

$$a > b^d$$

$$\Theta(n^{\log_2 2}) \quad a > b^d$$

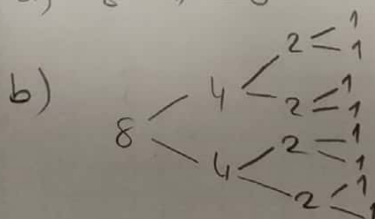
$$T(n) = \Theta(n^{\log_2 2})$$

$$T(n) = \boxed{\Theta(n)}$$

Result: In general time complexity equal but in the second function has more call for a value. ($a=(n)$ $b=(2n-1)$ both linear).

For example: $n=8$ is

a) 8 - 7 - 6 - 5 - 4 - 3 - 2 - 1 => n call



(2n-1) call

15 step to make operation

I choose the first one

2)

$$T(n) = T(n-1) + 1$$

$$T(n) = T(n-2) + 2$$

$$T(n) = T(n-3) + 3$$

$$T(n) = T(n-k) + k$$

$$n-k = 0$$

$$n = k$$

$$T(n) = \begin{cases} 1 & n = 0 \\ T(n-1) + 1 & n > 0 \end{cases}$$

$$T(n) = \underbrace{T(0)}_1 + n$$

$$T(n) \in \Theta(n)$$

Yes it is possible to design an algorithm has better complexity.
Using recurrence relation and substitution.

3) It could be start anywhere and finish every where in

for x, sChar in enumerate(string): (n)

if -start == sChar and end in string[x:]: (1)

for y, eChar in enumerate(string[x:]): (n-x)

print(eChar)

if (end == eChar) } (1)

break

print("\n")

$$\sum_{i=0}^n (n-i) \Rightarrow (n) + (n-1) + (n-2) + \dots + 1 \Rightarrow \frac{n \cdot (n+1)}{2} \Rightarrow T(n)$$

$$T(n) = \Theta(n^2)$$

4) Pseudo code :

function1 calDistTwoPoint (p1, p2) :

dist = 0.0

for c1, c2 in zip(p1, p2) // Iterating over coordinate points

dist += pow(c1 - c2, 2) // $(c_1 - c_2)^2 + \dots + (c_n - c_n)^2$

end for

return sqrt(dist)

end function

function2 minDist (points) :

minDist = math.inf

for i, p1 in enumerate(points):

for p2 in points [i+1:] :

cDist = calDistTwoPoint (p1, p2)

if cDist < minDist :

minDist = cDist

end if

end for

end for

return minDist

end function

k = dimension number
n = point number

Time Complexity : function 1 complexity depend on dimension so

calDistTwoPoint function $T(k) = \Theta(k)$

function 2 complexity depend on points numbers and calling function 1 each time

$$\sum_{i=0}^n (n-i) * T(k) \Rightarrow \frac{n^2 + n}{2} * \Theta(k) \Rightarrow \Theta(n^2) * \Theta(k)$$

$$n + (n-1) + (n-2) + \dots + 0$$

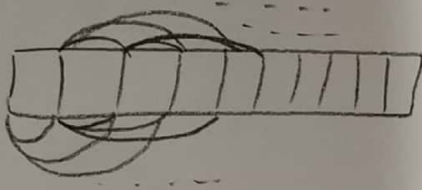
$$\frac{n \cdot (n+1)}{2}$$

$$\boxed{T(nk) \Rightarrow \Theta(n^2 k)}$$

5) a) I use 3 function for calculating this algorithm.

- 1) findAllSubSet();
- 2) calculate profit();
- 3) find Most Profitable cluster();

5.1) finding all consecutive subsets of array



- There is a list which we keep the all subset of given stops.

General algo \Rightarrow

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for i, stop in enumerate(stops):  $\Rightarrow (1)$ 
    tmp = [stop]  $\Rightarrow (1)$ 
    subsets.append(tmp.copy())  $(1 \times (n))$ 
    for _next in stops[i+1:]:  $\Rightarrow (n-i)$ 
        tmp.append(_next)  $\Rightarrow (1)$ 
        subsets.append(tmp.copy())  $\Rightarrow (1 \times (n))$ 

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$$T(n) = (n \times (1 + (1 \times (n))) + ((n-i) \times 1 \times (1 \times (n))))$$

$$T(n) = (n \times (n + \underbrace{(n-i) \times (n)}_{n^2}))$$

$$\boxed{T(n) = \Theta(n^3)}$$

S.2) calculate Profit (-subset)

for stop in -subset }
: } 1 } n $T(n) = \Theta(n)$

S.3) Find Most Profitable Cluster (-subsets);

profit = -inf } (1)
mostp = []

for -subset in -subsets : $\Rightarrow \left(\frac{n \cdot (n+1)}{2} \right) \Rightarrow (n^2)$
cprofit = calculate Profit (-subset) $\Rightarrow (n)$

If (cprofit > profit);
profit = cprofit } (1)
mostp = subset

return mostp;

$T(n) = \Theta(n^3)$

subsets = find All subsets () $\Rightarrow \Theta(n^3)$

most Pro = find Most Profitable Cluster () $\Rightarrow \Theta(n^3)$

$$\boxed{T(n) = \Theta(n^3)}$$

PS: I consider the array.copy() and iterating
over the substring is $T(n) = n$ if we take this as
constant (1) so the time complexity will be

$$T(n) = \Theta(n^2)$$

5) b) This is much more complexity solution because of the backtracking and recursive calls.

<u>n</u>	<u>counter of recursive calls</u>
1	2
2	5
3	34
4	125
5	461
6	1715
7	6432

I could not figure out totally dividing array
we are dividing -1 each time or for each element

$$T(n) = 2T(n-1) + 1$$

$$T(n) = 2[T(n-2)] + 1$$

$$T(n) = \Theta(2^n)$$

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