

1-)

$$a) \lim_{n \rightarrow \infty} \frac{(n^2 - 3n)^2}{5n^3} = \lim_{n \rightarrow \infty} \frac{n^4 - 6n^3 + 9n^2}{5n^3} = \lim_{n \rightarrow \infty} \frac{n - 6 + \frac{9}{n}}{5}$$

(Divide n^3)

As n approaches infinity, this simplifies to, $\lim_{n \rightarrow \infty} \frac{n}{5} = \infty$

Since the limit is infinity, $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$, $f(n) = \Omega(g(n))$

$$b) \lim_{n \rightarrow \infty} \frac{n^3}{\log_2 n^4} = \frac{1}{4} \cdot \lim_{n \rightarrow \infty} \frac{n^3}{\log_2 n} = \text{Apply L' Hospital}$$

$$= \frac{1}{4} \cdot \lim_{n \rightarrow \infty} \frac{3n^2}{\frac{1}{x \cdot \ln(2)}} = \frac{1}{4} \cdot \lim_{n \rightarrow \infty} 3 \ln 2 \cdot n^3 = \frac{1}{4} \cdot 3 \ln 2 \cdot \lim_{n \rightarrow \infty} n^3$$

$$\frac{1}{4} \cdot 3 \ln 2 \cdot \lim_{n \rightarrow \infty} n^3 = \infty, \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty, f(n) = \Omega(g(n))$$

$$c) \lim_{n \rightarrow \infty} \frac{5 \cdot \log_2(n^4)}{n \cdot \log_2(n^5)} = \lim_{n \rightarrow \infty} \frac{20 \cdot \log_2 n}{n \cdot 5 \cdot \log_2 n} = \lim_{n \rightarrow \infty} \frac{4}{n}$$

which approaches 0 as n becomes large, so $f(n) = O(g(n))$

$$d) \lim_{n \rightarrow \infty} \frac{n}{10n} = \lim_{n \rightarrow \infty} \frac{1}{10} = \frac{1}{10}$$

$0 < c = \frac{1}{10} < \infty$, $\frac{1}{10}$ is constant, therefore, $f(n) = \Theta(g(n))$

$$e) \lim_{n \rightarrow \infty} \frac{8 \cdot n^{5/2}}{n \cdot 3^{n/2}} = \lim_{n \rightarrow \infty} 8 \cdot \frac{n^{5/2}}{3^{n/2}} = 8 \cdot \lim_{n \rightarrow \infty} \frac{n^{5/2}}{3^{n/2}} \cdot \lim_{n \rightarrow \infty} \left(\frac{1}{n}\right)$$

$\lim_{n \rightarrow \infty} \frac{n^{5/2}}{3^{n/2}} = 0$, because, this is a comparison between a polynomial and exponential function. As n approaches infinity, the exponential term ($3^{n/2}$) grows much faster than polynomial term ($n^{5/2}$).

$$\lim_{n \rightarrow \infty} \frac{1}{n} = 0$$

Therefore, $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$, $f(n) = O(g(n))$

2-)

a-) The loop variable i is initialized once, so it's a constant time operation,

$O(1)$. The loop condition $i < \text{str-array-length}$ is checked $(n+1)$ times (where n is the length of the array.) checking the condition is a constant time operation, so this contributes $O(n)$. Inside the loop, we're assigning an empty string to each array element. string assignment is generally considered a constant time operation, so this also contributes $O(n)$.

The loop variable i is incremented (n) times which is $O(n)$.

Adding these up, the dominant term is $O(n)$, and since we ignore lower-order terms and constant factors in Big O notation

the worst-case time complexity of method A is $O(n)$.

b-) The first for loop calls method A, which is $O(n)$, n times. So, this loop contributes $O(n^2)$ complexity. The second for loop prints each element of the array, contributing $O(n)$ complexity. The combined time complexity is the sum of the complexities of these two loops, dominated by the first loop.

Time complexity $O(n^2)$, because the first loop is the dominant factor.

c-) The outer loop runs (n) times, where (n) is the length of the array. The inner loop runs (n) times for each iteration of the outer loop resulting in (n^2) total iterations of the inner loop.

Within the inner loop, method B is called, which has a time complexity of $O(n^2)$

To find overall time complexity: multiply the complexities of the nested loops with the complexity of the method call within the innermost loop:

$$(n) (\text{outer loop}) * (n) (\text{inner loop}) * O(n^2) (\text{method B}) = O(n^4)$$

Therefore, the worst-case time complexity of method C is $O(n^4)$

d-) This method prints each element and then resets it to an empty string. However, the $i--$ operation inside the loop will cause the loop to iterate indefinitely, leading to an infinite loop as the index i is decremented each time after incrementing. Time complexity not applicable (infinite loop)

e) The loop runs at most (n) times, where (n) is the length of the array. The comparison `str-array[i] == ""` is a constant time operation, $O(1)$. If an empty string is found early in the array, the loop breaks, resulting in fewer than (n) iterations. In the worst case, the empty string is not present, or it is the last element, so the loop will run (n) times. Therefore, the worst-case complexity of method E is $O(n)$.

3-) Algorithm `maxDifferenceSortedArray(A)`:

a)
 $n = \text{length}(A)$
if $n < 2$:
 return "Array must have at least two elements"
 $\text{max-diff} = A[n-1] - A[0]$
return max-diff

Since the array is sorted, the smallest element is $A[0]$, and the largest is $A[n-1]$. The maximum difference is simply the difference between these two.

Worst-case complexity: The time complexity is $O(1)$ because it performs a constant number of operations regardless of the size of array.

b) Algorithm `MaxDifferenceUnsortedArray(A)`:

$n = \text{length}(A)$
if $n < 2$:
 return "Array must have at least two elements"

$\text{min-val} = A[0]$

$\text{max-val} = A[0]$

for i from 1 to $n-1$:

 if $A[i] < \text{min-val}$:

$\text{min-val} = A[i]$

 if $A[i] > \text{max-val}$:

$\text{max-val} = A[i]$

$\text{max-diff} = \text{max-val} - \text{min-val}$

return max-diff

We initialize min-val and max-val with the first element. Then, we iterate through the array, updating min-val and max-val as we find smaller or larger elements, respectively. The maximum difference is the difference between max-val and min-val.

Worst-Case Complexity: The time complexity is $O(n)$, because we scan through all (n) elements of the array once.

Both algorithms meet the requirements of having a worst-case time complexity of linear time or better.