

Problem 5

(a)

Let $T(n, k)$ denote the runtime of MergeArrays, then let $S(p, q)$ denote the runtime of Merge. We say

$$T(n, k) = \sum_{i=1}^k S([i-1]n, n)$$

Then it follows that

$$(n, k) = \sum_{i=1}^k O([i-1]n + n) = \sum_{i=1}^k O([in - n + n) = \sum_{i=1}^k O(in)$$

Then we say

$$\sum_{i=0}^k O(in) = \sum_{i=0}^k iO(n) = O(n) \sum_{i=0}^k i = O(n) \frac{k(k-1)}{2} = O(n * k^2)$$

(b)

```
procedure MergeArrays(X_1[1..n], X_2[1..n], ... , X_k[1..n]) {
    midpoint = floor(k / 2);
    left_half = MergeArrays(X_1[1..n], X_2[1..n], ... , X_midpoint[1..n]);
    right_half = MergeArrays(X_midpoint[1..n], ... , X_k[1..n]);
    return Merge(left_half, right_half);
}
```

(c)

Let $T(n, k)$ denote the runtime of MergeArrays, and $S(p, q)$ denote the runtime of Merge. Then

$$\begin{aligned} T(n, k) &= O(1) + T(n, \frac{k}{2}) + T(n, \frac{k}{2}) + O(\frac{nk}{2} + \frac{nk}{2}) = O(nk) + 2T(n, \frac{k}{2}) \\ &= O(nk) + 2O(\frac{nk}{2}) + 4O(\frac{nk}{4}) + \dots + 2^{\log_2(k)} O(\frac{nk}{2^{\log_2(k)}}) \\ &= O(nk) + O(nk) + \dots + O(nk) = \log_2(n) O(nk) = O(nk \log n) \end{aligned}$$

Problem 6

(a)

Case 1: We have no increasing exponential subsequence to which we can append $A[i]$ If this is the case then $S(i) = 1$.

Case 2: Otherwise There exists some previous increasing exponential subsequence to which we can append $A[i]$. Choose the longest of these, and add 1 to its length.

$$S(i) = \max \left\{ 1, \max_{j \in \{1, \dots, i-1\}} \{S(j) + 1 : A[i] > 2A[j]\} \right\}$$

(b)

```
function LongestExpSubsequence(array[1..n]) {
    running_max = 1
    length = [] * n
    for i in {1, 2, ..., n} {
        length[i] = 1
        for j in {1, 2, ..., i-1} {
            if 2array[j] < array[i] {
                length[i] = max{length[i], 1 + length[j]}
            }
        }
    }
    running_max = max{running_max, length[i]}
```

```
}  
}
```

(c)

What is the running time of this algorithm.

The running time of this algorithm will be $O(n^2)$ because