

1. a) [A1 problem]

We use an extended version of Karatsuba's algorithm. We split the polynomials into upper and lower halves, denoted by $P_u(x)$, $P_l(x)$, $Q_u(x)$, $Q_l(x)$. We have that

$$P(x) = P_u(x)x^{n/2} + P_l(x)$$

$$Q(x) = Q_u(x)x^{n/2} + Q_l(x)$$

We use a divide and conquer algorithm as follows.

```
function polynomialMult(P, Q) {
    initialize Pu, Pl, Qu, Ql
    lower = polynomialMult(Pl, Ql)
    upper = polynomialMult(Pu, Qu)
    middle = polynomialMult(Pl + Pu, Ql + Qu)
    return upper * x^n + (middle - lower - upper) * x^n/2 + lower
}
```

Runtime: We have 3 recursive calls each with a subproblem of size $n/2$. Our remaining operations take $O(n)$ time, since we are adding and subtracting. Hence we have that the recurrence relation is $T(n) = 3T(n/2) + O(n)$, which resolves to $O(n^{\log_2(3)})$.

b) Our algorithm works as follows. For each integer a_k , we would like to find 2 other integers a_i and a_j that sum to a value greater than or equal to it. We do this by forming the polynomial

$$P(x) = x^{a_1} + \dots + x^{a_{k-1}} + x^{a_{k+1}} + \dots + x^{a_n}$$

We represent this as an array of coefficients of length n , with a 1 at each a_i from $i=1$ to n , excluding a_k , and 0 otherwise.

Then we compute $P(x)^2$. The coefficient of x^{a_i} is the number of pairs of integers that sum to a_i . Hence we iterate through all numbers from a_{k+1} to n , and sum their coefficients. We divide this value by 2, since we double count each pair. By running this for every a_k , we find the total number of triples such that $i < j < k$ and $a_i + a_j \geq a_k$.

Note that all the sum of all $a_i \leq n$.

```
function counting(a1...an) {
    result = 0
    for k=1 to n
        P = array of length n, 0-initialized
        for i=1 to k-1
            P[a_i] = 1
        P2 = polynomialMultiply(P, P)
        for i=a_k+1 to n
            result += P2[i]/2
    return result
}
```

Runtime: We iterate through each element in the array once. In each iteration, we create the polynomial, which takes $O(n)$ time. Polynomial multiplication takes $O(n^{\log_2(3)})$ time. Iterating from a_{k+1} to n takes $O(n)$ time. Hence our algorithm runs in $O(n^{\log_2(3)})$ time.

2. [A1 problem]

A) We have that $T(n) = T(2n/3) + T(n/3) + n^2 \geq n^2$ for all $n \geq 0$. Hence we have that $T(n) \in \Omega(n^2)$.

We show that $T(n) \in O(n^2)$. Consider the recurrence $S(n) = T(2n/3) + T(n/3) + n^2 = 2T(2n/3) + n^2$. Using the Master Theorem, we have $a = 2$, $b = 3/2$, and $c = 2$. We have that $\log_b(a) = \log_{3/2}(2) \approx 1.71 < c$. Hence by the Master Theorem, $S(n) \in \Theta(n^2)$ and thus $S(n) \in O(n^2)$. Since for all $n \geq 0$ we have that $T(n) \leq S(n)$, we have that $T(n) \in O(S(n))$. Hence we have that $T(n) \in O(n^2)$.

Since we have $T(n) \in \Omega(n^2)$ and $T(n) \in O(n^2)$, we have that $T(n) \in \Theta(n^2)$.

B) At level k of the recursion tree, there are $\sqrt{k} \cdot n^{1/4} \cdot \dots \cdot n^{1/2k} = n^{(1-2^{-k})}$ nodes with each node having $n^{1/2k}$ work. Hence the total work on level k is n .

When $k = \log \log n$, the subproblem size is $n^{1/2k} = 2$. Hence the total work done is at most $n \cdot k = n \log \log n$, so we have that $T(n) \in O(n \log \log n)$.

3. [similar to Tutorial 2 problem]

We first derive an algorithm to find the indices of the maximum subarray in a 1D array. We use Kadane's algorithm, which works as follows. We define a variable `currentSum` which stores the maximum sum ending at our current location. We also keep track of a global maximum sum. We iterate through the array, and add the value of the current element to `currentSum`. If this value is less than zero, we reset `currentSum` to 0. This allows us to discard negative subarrays.

```
function maxSum1D(A) {
    maxSum = INT_MIN
    currentSum = 0

    // Indices for maxSum and currentSum ranges
    maxSumStart = 0
    maxSumEnd = 0
    currentSumStart = 0

    for i=0 to n
        currentSum += A[i]
        if (maxSum < currentSum)
            maxSum = currentSum
            maxSumStart = currentSumStart
            maxSumEnd = i
        if (currentSum < 0):
            currentSum = 0
            currentSumStart = i + 1
    return maxSum, maxSumStart, maxSumEnd
}
```

Since we iterate through the array once, this algorithm runs in $O(n)$ time.

Using our `maxSum1D` algorithm, we now derive an algorithm for the maximum sum rectangle problem. We iterate through all possible upper and lower boundaries for a rectangle in `A`. For each pair (u, d) , we create an array `B`, where $B[i] = \text{sum}(A[u\dots d][i])$. Then we use the `maxSum1D` algorithm above to calculate the left and right indices of the maximum rectangle with upper and lower bounds (u, d) .

```
function maxRectangle(A) {
    maxRect = 0
    indices = (0,0,0,0)
    for u=1 to n
        B = [0]*n

        for d=u to n
            for i=1 to n: B[i] += A[d][i]
            maxRect = max(maxRect, max1DSum(B))
            update indices
    return maxRect, indices
}
```

Runtime: Iterating through all pairs (u, d) takes $O(n^2)$ time. For each pair (u, d) , we call create the array `B` and `max1DSum`, which takes $O(n)$ time. Hence our algorithm takes $O(n^3)$ time.

4.

We know that `preorder[0]` is always the root of the tree. We can find `preorder[0]` in `inorder`, at some index i . `inorder[i]` is traversed after its left subtree, so `inorder[start to i-1]` is the left subtree, and `inorder[i+1 to end]` is the right subtree. Hence we can build the tree by recursively building its children.

```
function buildTree(preorder, inorder) {
    if (inorder)
        root = preorder.popfront()
        i = index of root in inorder

        treeNode = new TreeNode(root)
        treeNode.left = buildTree(preorder, inorder[start to i-1])
        treeNode.right = buildTree(preorder, inorder[i+1 to end])

    return treeNode
}
```

Runtime: We iterate through each element of `preorder` once. For each iteration, it takes $O(n)$ time to find the index of `preorder[0]` in `inorder`. Hence our solution runs in $O(n^2)$ [can optimize by using a hashmap for `inorder`].

Correctness: Prove by induction.

Bonus

This algorithm is the same idea as the one above but we do it iteratively. It works as follows. We keep pushing nodes from `preorder` into a stack. For each node we push onto the stack, we create a `TreeNode` for it and push it onto a `TreeNode` stack. We push nodes from `preorder` until the top of the stack matches the first element of `inorder`. We then pop elements from the stack

until it no longer matches inorder.

```
function buildTree(preorder, inorder) {
    init stack, treeNodeStack
    createRightChild = false

    root = preorder.pop()
    treeRoot = new TreeNode(root)
    curTreeNode = treeRoot

    while (!preorder.empty())
        if (!stack.empty() and stack.top == inorder.top())
            treeRoot = treeNodeStack.pop()
            stack.pop()
            createRightChild = true
            inorder.pop()
        else
            node = preorder.pop()
            stack.push(node)
            treeNode = new TreeNode(node)
            if (createRightChild)
                createRightChild = false
                curTreeNode.right = treeNode
                curTreeNode = treeRoot.right
            else
                curTreeNode.left = treeNode
                curTreeNode = treeRoot.left

            treeNodeStack.push(treeNode)
    return treeRoot
}
```

Runtime: We pop every element from preorder and inorder once each. Hence this algorithm runs in $O(n)$ time.