

Quiz 2

- Do not open this quiz booklet until directed to do so. Read all the instructions on this page.
- When the quiz begins, write your name on every page of this quiz booklet.
- You have 75 minutes to earn 180 points. **However, 150 points is viewed as full mark. Anything above that are bonus points.** Do not spend too much time on any one problem. Read them all first, and attack them in the order that allows you to make the most progress.
- **This is a closed-book exam.** No calculators or programmable devices are permitted. No cell phones or other communications devices are permitted.
- Write your solutions **using non-erasable pens** in the space provided. If you need more space, write on the back of the sheet containing the problem. Pages may be separated for grading.
- Do not waste time and paper rederiving facts that we have studied. Simply cite them.
- When writing an algorithm, a **clear** description in English will suffice. Pseudo-code is also okay but not required.
- **Pay close attention to the instructions for each problem.** Depending on the problem, partial credit may be awarded for incomplete answers.
- Turn in your exam paper to your designated AI after the exam, and sign your name on the sign-in sheet.

| Problem | Parts | Points | Grade | Grader |
|---------|-------|--------|-------|--------|
| Name | 0 | 4 | | |
| 1 | 7 | 63 | | |
| 2 | 2 | 20 | | |
| 3 | 1 | 30 | | |
| 4 | 1 | 30 | | |
| 5 | 7 | 33 | | |
| Total | – | 180 | | |

Name: _____

- D. There is an order on the subproblems so that larger subproblems are reduced to smaller subproblems in the order.

Solution: B.

- (e)) Recall that the Fast Fourier Transform algorithm takes a polynomial $A(x) = a_0 + a_1x + a_2x^2 + \dots + a_{n-1}x^{n-1}$ as input and evaluates it at all n -th roots of unity. Which of the following statements is **not true** regarding the algorithm?
- A. n has to be in the form of $n = 2^k$ ($k = 0, 1, 2, \dots$) in the recursive algorithm.
 - B. Space complexity is $\Theta(n \log n)$.
 - C. The same algorithm (after small modification) can be used to compute the coefficients of $A(x)$ once given the evaluations at n -roots of unity.
 - D. Time complexity is $\Theta(n \log n)$ arithmetic operations.

Solution: B. At the i -th recursive call, the algorithm solves a problem of size $\frac{n}{2^i}$ and therefore uses space $O\left(\frac{n}{2^i}\right)$ (not counting space used by further recursive calls). Therefore the total space complexity is $\sum_{i=0}^{\log_2 n} O\left(\frac{n}{2^i}\right) = O(n)$.

- (f)) Consider the problem of multiplying two $n \times n$ matrices. The straightforward algorithm takes $O(n^3)$ time. Now we would like to design a divide-and-conquer algorithm that divides the problem into q subproblems, where in each subproblem the task is to multiply two $\frac{n}{2} \times \frac{n}{2}$ matrices. In order to make the algorithm run faster than $O(n^3)$ time, q has to be at most
- A. 2.
 - B. 3.
 - C. 7.
 - D. 8.

Solution: C.

- (g)) We have learnt the divide-and-conquer algorithm for 2-Dimensional Closest Pair of Points. Now let us consider the same problem in 3-D space (i.e. each point has 3 coordinates (x_i, y_i, z_i)). The following algorithm is a simple analogue of the algorithm for 2-D, where the difference is marked by **boldface**.
- i. Choose x_0 and split the given set of n points into 2 sets of $n/2$ points by the plane $x = x_0$.
 - ii. Solve the two subproblems recursively. Let ϵ be the smallest distance between pair of points within each subproblem.
 - iii. Let T be the set of points (x_i, y_i, z_i) such that $|x_i - x_0| < \epsilon$. Sort all points in T **according to $y_i + z_i$** .
 - iv. Check the distance between each point in T and its **100 neighbors** in the sorted list. Let δ be the smallest distance found.
 - v. Return $\min(\epsilon, \delta)$.

Choose the correct statement about this algorithm.

- A. It **can** be made to run in $O(n \log n)$ time and **gives** the correct answer.
- B. It **cannot** be made to run in $O(n \log n)$ time but **gives** the correct answer.
- C. It **can** be made to run in $O(n \log n)$ time but **does not give** the correct answer.
- D. It **cannot** be made to run in $O(n \log n)$ time and **does not give** the correct answer.

Solution: C.

Problem 2. Short Answer Problems [20 points] (2 parts)

(a) [10 points] Solve the following recurrence.

$$\begin{cases} T(n) = 4T(n/2) + n^2 & (n \geq 2) \\ T(n) = 1 & (n \leq 1) \end{cases}.$$

Your answer should be in the form of $\Theta(f(n))$ and you do not have to show the intermediate steps.

Solution: $T(n) = \Theta(n^2 \log n)$.

(b) [10 points] Solve the following recurrence.

$$\begin{cases} T(n) = T(n/2) + T(n/3) + T(n/6) + n & (n \geq 6) \\ T(n) = 1 & (n < 6) \end{cases}.$$

Your answer should be in the form of $\Theta(f(n))$ and you do not have to show the intermediate steps.

Solution: $T(n) = \Theta(n \log n)$.

Problem 3. Fast Fourier Transform with Base 3 [30 points]

We have learned the Fast Fourier Transform (FFT) algorithm to evaluate a given polynomial $A(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1}$ at the n n -th roots of unity $1, \omega, \omega^2, \dots, \omega^{n-1}$ (where $\omega = e^{\frac{2\pi i}{n}}$), using $O(n \log n)$ arithmetic operations. The algorithm we saw in class works only when $n = 2^k$ is an integer power of 2.

In this problem, you are asked to modify the FFT algorithm presented in class so that it works for the case when $n = 3^k$ is an integer power of 3, still using $O(n \log n)$ arithmetic operations.

Please note that our task is different from polynomial multiplication so the simple padding 0's trick may not work.

Solution:

The idea goes similar as the one presented in Lecture 13.

Here we introduce $A_1(x) = a_0 + a_3x + \cdots + a_{n-3}x^{n/3-1}$, $A_2(x) = a_1 + a_4x + \cdots + a_{n-2}x^{n/3-1}$, and $A_3(x) = a_2 + a_5x + \cdots + a_{n-1}x^{n/3-1}$. Hence, it holds that

$$A(w^j) = A_1((w^3)^j) + w^j \cdot A_2((w^3)^j) + w^{2j} \cdot A_3((w^3)^j).$$

Hence to evaluate $A(x)$ at $x = w^1, w^2, \dots, w^n = 1$, we can first evaluate $A_1(x), A_2(x), A_3(x)$ at $x = w^3, w^6, \dots, (w^3)^n$, and then use $O(n)$ time to get the final n evaluations of $A(x)$. Note that $w^{n+3} = w^3, \dots, w^{3n} = 1$. There are totally $n/3$ different values of $w^3, w^6, \dots, (w^3)^n$. And this reduces the problem of evaluating the degree- $(n-1)$ polynomial $A(x)$ at n n -th roots of unity to the three subproblems of evaluating degree- $(n/3-1)$ polynomials at $(n/3)$ $(n/3)$ -th roots of unity.

Let $T(n)$ denote the time complexity of evaluating $A(x)$ at $x = w^1, w^2, \dots, w^n = 1$. We have $T(n) = 3T(n/3) + O(n)$ from which we can see $T(n) = O(n \log n)$ according to the Master Theorem.

Problem 4. Merging sorted lists [30 points]

When learning the Merge-Sort algorithm, we saw how to merge 2 sorted lists in linear time. Now let us extend the merging algorithm to deal with more sorted lists.

Suppose you are given $k \geq 2$ lists as input, where each list L_i ($i = 1, 2, 3, \dots, k$) contains n/k numbers sorted in non-decreasing order. Please describe an algorithm that runs in $O(n \log k)$ time and merges the k sorted lists into one sorted list L of all n numbers. You do not have to formally prove the correctness of the algorithm. However, you should prove the runtime of your algorithm.

Your algorithm description is worth 18 points; the runtime analysis is worth 12 points. You will get the latter 12 points only when your algorithm is correct (or almost correct).

Solution: The following recursive algorithm merges a sorted lists (L_1, L_2, \dots, L_a) , where $\text{Merge}(P, Q)$ is the same algorithm we used in Merge-Sort.

Merge – Lists(a, L_1, L_2, \dots, L_a)

- IF $a = 1$ THEN RETURN L_1
- LET $m = \lfloor a/2 \rfloor$
- LET $P = \text{Merge – Lists}(m, L_1, L_2, \dots, L_m)$
- LET $Q = \text{Merge – Lists}(a - m, L_{m+1}, L_{m+2}, \dots, L_a)$
- RETURN $\text{Merge}(P, Q)$

Let $T(a)$ be the time complexity of Merge – Lists with a lists as input. We have $T(1) = O(n/k)$ as boundary condition. For $a \geq 1$, we have $T(a) = 2T(a/2) + O(an/k)$ (since there are an/k numbers in total for $\text{Merge}(P, Q)$ to process). Solving this recurrence, we get $T(a) = O((n/k) \cdot a \log a)$. Therefore, the time complexity to merge k lists is $T(k) = O(n \log k)$.

Problem 5. Segmented Least Errors [33 points] (7 parts) In class we discussed the Segmented Least Squares problem. Here is a quick recap of the problem. There are n data points $P_1, P_2, P_3, \dots, P_n$, and we would like to partition them into a few segments. Suppose we partition the data points into L segments as follows (where $a_0 = 1$ and $a_L = n + 1$),

$$[a_0, a_1), [a_1, a_2), [a_2, a_3), \dots, [a_{L-1}, a_L).$$

We define the total error to be

$$E = \sum_{i=1}^L e_{a_{i-1}, a_i} + cL$$

where c and e_{ij} 's are given as input. The goal is to find out the partition with minimum cost E .

- (a) [5 points] Let $\text{OPT}_E[j]$ be the optimal solution (i.e. minimum cost) for the subproblem defined on $P_1, P_2, P_3, \dots, P_j$. Please write down the Bellman Equation for $\text{OPT}_E[j]$ ($1 \leq j \leq n$) based on the optimal solution of smaller subproblems.

Solution:

$$\text{OPT}_E[j] = \min_{i: 1 \leq i \leq j} \{ \text{OPT}_E[i-1] + e_{i,j} + c \}.$$

Now let us consider the question when the total cost is non-linear in L . Suppose we would like to minimize the the new cost defined as follows

$$F = \sum_{i=1}^L e_{a_{i-1}, a_i} + c\sqrt{L}.$$

In order to solve this new problem, we need to use the technique of “adding a variable” as we did in the Knapsack problem. We define $\text{OPT}_F[j, \ell]$ to be the optimal solution (i.e. minimum cost) for the subproblem of partitioning $P_1, P_2, P_3, \dots, P_j$ into *exactly* ℓ segments.

- (b) [4 points] How many subproblems are there in total (up to the $\Theta(\cdot)$ notation)?

Solution: $\Theta(n^2)$ since $1 \leq \ell \leq j \leq n$.

- (c) [4 points] Suppose we have solved all subproblems. What is the optimal solution (i.e. minimum cost) to the original problem (to minimize F)?

Solution:

$$\min_{\ell: 1 \leq \ell \leq n} \text{OPT}_F[n, \ell].$$

- (d) [4 points] When $\ell = 1$, write the boundary condition for $\text{OPT}_F[j, 1]$ ($1 \leq j \leq n$).

Solution:

$$\text{OPT}_F[j, 1] = e_{1,j} + c.$$

- (e) [7 points] When $\ell > 1$, write the Bellman Equation for $\text{OPT}_F[j, \ell]$ ($\ell \leq j \leq n$). No proofs needed.

Solution:

$$\text{OPT}_F[j, \ell] = \min_{i: \ell \leq i \leq j} \left\{ \text{OPT}_F[i-1, \ell-1] + e_{i,j} + c \left(\sqrt{\ell} - \sqrt{\ell-1} \right) \right\}.$$

- (f) [4 points] What is a proper order to solve all the subproblems according to your Bellman Equation?

Solution: Either increasing order of j or increasing order of ℓ works.

- (g) [5 points] What is the time complexity of the Dynamic Programming algorithm based on your Bellman Equation (up to the $\Theta(\cdot)$ notation)?

Solution: $\Theta(n^3)$, since there are $\Theta(n^2)$ subproblems and it takes $\Theta(n)$ time to solve each subproblem.

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