

1. Solution to problem 1

- (a) Straightforward proof by induction.
- (b) This routine requires n additions and n multiplications. Let $n = 2^k$ and consider the polynomial $p(x) = x^n$. This polynomial can be evaluated at x using $\log n$ multiplications only by repeatedly squaring x .

2. Solution to Problem 2

- (a) Split the n -dimensional vector \mathbf{v} into vectors \mathbf{v}^t and \mathbf{v}^b which consist of the top and bottom $n/2$ entries of \mathbf{v} , respectively.

Similarly, split the n -dimensional vector $H_k \mathbf{v}$ into $(H_k \mathbf{v})^t$ and $(H_k \mathbf{v})^b$. Then

$$\begin{aligned}(H_k \mathbf{v})^t &= H_{k-1} \mathbf{v}^t + H_{k-1} \mathbf{v}^b = H_{k-1}(\mathbf{v}^t + \mathbf{v}^b) \\ (H_k \mathbf{v})^b &= H_{k-1} \mathbf{v}^t - H_{k-1} \mathbf{v}^b = H_{k-1}(\mathbf{v}^t - \mathbf{v}^b)\end{aligned}$$

To compute $H_k \mathbf{v}$ first compute $\mathbf{v}^t + \mathbf{v}^b$ and $\mathbf{v}^t - \mathbf{v}^b$ and then compute the products $H_{k-1}(\mathbf{v}^t + \mathbf{v}^b)$ and $H_{k-1}(\mathbf{v}^t - \mathbf{v}^b)$ recursively.

Running time: Let $T(n)$ be the time to multiply the $n \times n$ matrix H_k by an n -dimensional vector \mathbf{v} . Then $T(n) = cn + 2T(n/2) = O(n \log n)$.

3. Solution to Problem 3 (Inductive Proof)

- (a) Base case: For $n = 1$, $\Pr[\text{sample is the } i\text{-th item}] = \frac{1}{k} = 1$.
- (b) Hypothesis: Assume that for some $k \geq 1$, the probability of the sample being any of the k elements is $\frac{1}{k}$.
- (c) Step: We will show the statement for $k + 1$.

$$\Pr[\text{the sample is the } (k + 1)\text{-st item}] = \frac{1}{k + 1}$$

$$\Pr[\text{the sample is a different item from the } (k + 1)\text{-st}] = 1 - \frac{1}{k + 1} = \frac{k}{k + 1}$$

$$\Pr[\text{the sample is the } i\text{-th item}] = \frac{1}{k} \cdot \frac{k}{k + 1} = \frac{1}{k + 1}, \text{ for } i = 1, \dots, k$$

Therefore, the probability that any item is the sample is $\frac{1}{k+1}$.

- (b) The sample is the i^{th} item with probability =

$$\begin{cases} (\frac{1}{2})^{k-i+1} & i = 2, \dots, k \\ (\frac{1}{2})^{k-1} & i = 1 \end{cases}$$

4. Solution to Problem 4

- (a) Use standard algorithm for matrix multiplication to compute AB in $\Theta(n^3)$. Compare with C in $O(n^2)$.

(b) Consider the vector $\mathbf{v} = M\mathbf{x}$. Let v_i be the i -th entry of \mathbf{v} . For any fixed $1 \leq i \leq n$, we have

$$\Pr[\mathbf{v} \neq \mathbf{0}] \leq \Pr[v_i \neq 0].$$

So it suffices to prove that

$$\Pr[v_i \neq 0] \leq \frac{1}{2}.$$

Since M is non-zero, there is at least one entry in M that is not equal to 0; w.l.o.g., say entry $M_{ij} \neq 0$. Since v_i is given by the dot product of the i -th row of M with the vector \mathbf{x} , we have

$$v_i = \sum_{k=1}^n M_{ik} \cdot x_k = M_{ij} \cdot x_j + \sum_{\substack{k=1 \\ k \neq j}}^n M_{ik} \cdot x_k$$

It follows that v_i is 0 if and only if

$$M_{ij}x_j = - \sum_{\substack{k=1 \\ k \neq j}}^n M_{ik}x_k$$

Suppose that all the x_k 's have been set randomly and independently to $\{0, 1\}$ with probability $1/2$, **except** for x_j .

- If the right hand side above equals 0, then x_j must be set to 0 for v_i to equal 0. This happens with probability $1/2$.
- If the right hand side above does not equal zero, then certainly x_j should not be set to 0 for v_i to equal 0. Again this happens with probability $1/2$.

Hence $\Pr[v_i = 0] \leq \frac{1}{2}$ in either case.

(c) Set $M = AB - C$.

- If $AB = C$, then M is the all-zeros matrix. Hence for any \mathbf{x} , $\Pr[M\mathbf{x} = \mathbf{0}] = 1$.
- If $AB \neq C$, then M is a non-zero matrix. By part (b),

$$\Pr[AB\mathbf{x} = C\mathbf{x}] = \Pr[M\mathbf{x} = \mathbf{0}] \leq \frac{1}{2}.$$

Hence the randomized test for checking whether $AB = C$ is as follows:

Algorithm 1

VerifyMatrixProduct(A, B, C, n, k)

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1: for iteration  $i$  from 1 to  $k$  do
2:   Generate a random  $\mathbf{x}$  as in part (b)
3:   Compute  $C\mathbf{x}$ ,  $B\mathbf{x}$ ,  $A(B\mathbf{x})$ 
4:   Compute  $\mathbf{v}_i = AB\mathbf{x} - C\mathbf{x}$ 
5:   if  $\mathbf{v}_i \neq \mathbf{0}$  then
6:     return no
7:   end if
8: end for
9: return yes           // that is, return yes iff  $\mathbf{v}_i = \mathbf{0}$  for every  $1 \leq i \leq k$ 
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Fix an i . Line 2 inside the for loop takes $O(n)$ time; line 3 takes $O(n^2)$ time if we first compute $C\mathbf{x}$, $B\mathbf{x}$ and finally $A(B\mathbf{x})$; lines 4 and 5 take time $O(n)$. Hence, the running time of one iteration of the randomized algorithm is $O(n^2)$. Hence the total running time is $O(kn^2)$.

Success probability of the randomized test:

- If $AB = C$, then \mathbf{v}_i is always $\mathbf{0}$. Hence the algorithm *succeeds*—that is, it always returns the correct answer—with probability 1.
- If $AB \neq C$, then the algorithm *fails*—that is, outputs the wrong answer—with probability at most $\left(\frac{1}{2}\right)^k$. Thus the algorithm succeeds with probability at least $1 - \left(\frac{1}{2}\right)^k$.

1. Solution to recommended exercise 1

f	g	O	o	Ω	ω	Θ
$\log^5 n$	$10 \log^3 n$	n	n	y	y	n
$n^2 \log(2n)$	$n \log n$	n	n	y	y	n
$\sqrt{\log n}$	$\log \log n$	n	n	y	y	n
$n^2 + n^{1/3}$	$n^2 \log n + n^{5/2}$	y	y	n	n	n
$\sqrt{n} + 1500$	$n^{1/3} + \log n$	n	n	y	y	n
$\frac{3^n}{n^2}$	$2^n \log n$	n	n	y	y	n
$n^{\log n}$	2^n	y	y	n	n	n
2^n	$\frac{3^n}{n^{\log n}}$	y	y	n	n	n
n^n	$n!$	n	n	y	y	n
$\log n^n$	$\log n!$	y	n	y	n	y

2. Solution to recommended exercise 2

Give tight asymptotic bounds for the following recurrences.

- According to master theorem, $a = 4$, $b = 2$, $k = 2$. Thus $a = b^k$ and $T(n) = O(n^k \log n)$, hence

$$T(n) = O(n^2 \log n)$$

- According to master theorem, $a = 8$, $b = 2$, $k = 3$. Thus $a = b^k$ and $T(n) = O(n^k \log n)$, hence

$$T(n) = O(n^3 \log n)$$

- According to master theorem, $a = 11$, $b = 4$, $k = 2$. Thus $a < b^k$ and $T(n) = O(n^k)$, hence

$$T(n) = O(n^2)$$

- According to master theorem, $a = 7$, $b = 3$, $k = 1$. Thus $a > b^k$ and $T(n) = O(n^{\log_b a})$, hence

$$T(n) = O(n^{\log_3 7})$$

3. Solution to recommended exercise 3

The recurrence for this algorithm is

$$T(n) = 3T(2n/3) + \Theta(1) = \Theta(n^{\log_{3/2} 3}) = \Theta(n^{\frac{\log 3}{\log 1.5}}) = \omega(n^2)$$

Hence both insertion sort and merge Sort are faster than this algorithm.