

## In-Class Midterm ( Solution Ideas )

( 2:35 PM – 3:50 PM : 75 Minutes )

- This exam will account for either 15% or 30% of your overall grade depending on your relative performance in the midterm and the final. The higher of the two scores (midterm and final) will be worth 30% of your grade, and the lower one 15%.
- There are four (4) questions, worth 75 points in total. Please answer all of them in the spaces provided.
- There are 16 pages including four (4) blank pages and two (2) pages of appendices. Please use the blank pages if you need additional space for your answers.
- The exam is *open slides* and *open notes*.

**GOOD LUCK!**

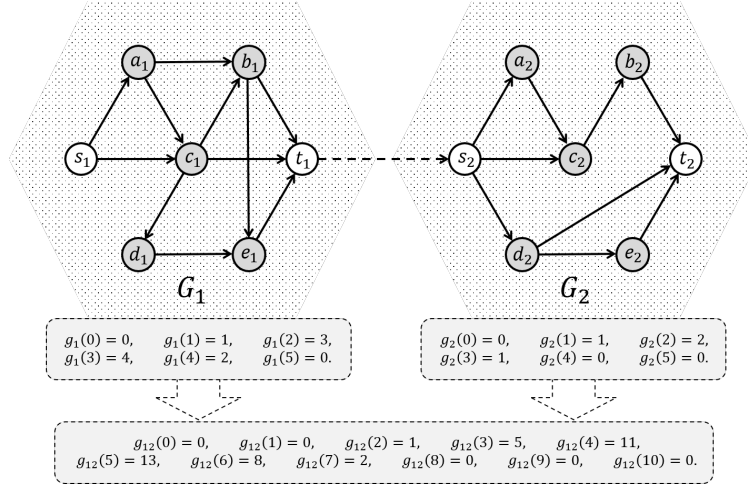
Question	Pages	Score	Maximum
1. Counting Paths	2–4		20
2. A Schönhage-Strassen-like Recurrence	6–8		25
3. Closest Pair of Points	10–11		20
4. An Impossible Priority Queue	13		10
Total			75

NAME: \_\_\_\_\_

**QUESTION 1. [ 20 Points ] Counting Paths.** Suppose you are given two directed graphs<sup>1</sup>  $G_1$  and  $G_2$  containing  $n + 2$  nodes each for some  $n \geq 0$ . For  $i \in \{1, 2\}$ ,  $G_i$  includes two special nodes — a *source node*  $s_i$  with no incoming edges<sup>2</sup> and a *target node*  $t_i$  with no outgoing edges<sup>3</sup>. These two nodes are called *external nodes* while the rest are called *internal nodes*. The figure below shows an example with  $n = 5$  in which the internal nodes are colored grey and the external nodes are white. Let  $g_i(k)$  denote the number of paths in  $G_i$  that go from  $s_i$  to  $t_i$  and pass through exactly  $k$  internal (i.e., grey) nodes. For example, in the figure below  $g_1(3) = 4$  which represents the following 4 paths:

$$\begin{aligned} & s_1 \rightarrow a_1 \rightarrow b_1 \rightarrow e_1 \rightarrow t_1, \\ & s_1 \rightarrow a_1 \rightarrow c_1 \rightarrow b_1 \rightarrow t_1, \\ & s_1 \rightarrow c_1 \rightarrow b_1 \rightarrow e_1 \rightarrow t_1 \\ & \text{and } s_1 \rightarrow c_1 \rightarrow d_1 \rightarrow e_1 \rightarrow t_1. \end{aligned}$$

Suppose for  $0 \leq k \leq n$ , all  $g_1(k)$  and  $g_2(k)$  values are known to you.



Now suppose you connect  $G_1$  and  $G_2$  by putting an edge directed from  $t_1$  to  $s_2$ . For  $0 \leq k \leq 2n$ , let  $g_{12}(k)$  denote the number of paths from  $s_1$  to  $t_2$  that pass through exactly  $k$  internal (i.e., grey) nodes. The figure above shows an example in which  $g_{12}(3) = 5$  representing the following 5 paths:

$$\begin{aligned} & (s_1 \rightarrow c_1 \rightarrow t_1) \rightarrow (s_2 \rightarrow c_2 \rightarrow b_2 \rightarrow t_2), \\ & (s_1 \rightarrow c_1 \rightarrow t_1) \rightarrow (s_2 \rightarrow d_2 \rightarrow e_2 \rightarrow t_2), \\ & (s_1 \rightarrow a_1 \rightarrow b_1 \rightarrow t_1) \rightarrow (s_2 \rightarrow d_2 \rightarrow t_2), \\ & (s_1 \rightarrow a_1 \rightarrow c_1 \rightarrow t_1) \rightarrow (s_2 \rightarrow d_2 \rightarrow t_2) \\ & \text{and } (s_1 \rightarrow c_1 \rightarrow b_1 \rightarrow t_1) \rightarrow (s_2 \rightarrow d_2 \rightarrow t_2). \end{aligned}$$

<sup>1</sup>e.g., road networks with one-way roads

<sup>2</sup>e.g., incoming roads

<sup>3</sup>e.g., outgoing roads

1(a) [ **5 Points** ] For any given integer  $k \in [0, 2n]$ , show that  $g_{12}(k)$  can be computed from  $g_1$ 's and  $g_2$ 's in  $\mathcal{O}(n)$  time.

**Solution.** Observe that for any  $k \in [0, 2n]$ ,

$$g_{12}(k) = \sum_{i = \max\{0, k-n\}}^{\min\{k, n\}} g_1(i) \times g_2(k-i)$$

Then clearly, for any given  $k \in [0, 2n]$ , the following code fragment computes  $g_{12}(k)$ .

1.  $c \leftarrow 0$
2. **for**  $i \leftarrow \max\{0, k-n\}$  **to**  $\min\{k, n\}$  **do**
3.      $c \leftarrow c + g_1(i) \times g_2(k-i)$
4.  $g_{12}(k) \leftarrow c$

Observe that the **for** loop above iterates  $t = \min\{k, n\} - \max\{0, k-n\} + 1$  times with each iteration taking  $\Theta(1)$  time, and so the code fragment runs in  $\mathcal{O}(t)$  time. But  $t = n - (k-n) + 1 = n+1 = \mathcal{O}(n)$  when  $k > n$ , and since  $k \leq 2n$ ,  $t = k - 0 + 1 = k+1 = \mathcal{O}(n)$  otherwise.

Thus the computation of  $g_{12}(k)$  requires  $\mathcal{O}(n)$  time.

1(b) [ **15 Points** ] Show that for  $0 \leq k \leq 2n$ , one can compute all  $g_{12}(k)$  values simultaneously in  $\mathcal{O}(n \log n)$  time.

**Solution.** In  $\mathcal{O}(n)$  time we construct the following two polynomials of degree at most  $n$  each.

$$\mathcal{G}_1(z) = g_1(0) + g_1(1)z + g_1(2)z^2 + \dots + g_1(n)z^n = \sum_{0 \leq i \leq n} g_1(i)z^i$$

$$\mathcal{G}_2(z) = g_2(0) + g_2(1)z + g_2(2)z^2 + \dots + g_2(n)z^n = \sum_{0 \leq j \leq n} g_2(j)z^j$$

Let  $\mathcal{G}_{12}(z) = \mathcal{G}_1(z)\mathcal{G}_2(z)$ . Then clearly  $\mathcal{G}_{12}(z)$  is of degree at most  $2n$ . Let

$$\mathcal{G}_{12}(z) = c_0 + c_1z + c_2z^2 + \dots + c_{2n}z^{2n} = \sum_{0 \leq k \leq 2n} c_k z^k.$$

Since  $\mathcal{G}_{12}(z)$  is the product of  $\mathcal{G}_1(z)$  and  $\mathcal{G}_2(z)$ , clearly, for  $k \in [0, 2n]$ ,

$$c_k = \sum_{i = \max\{0, k-n\}}^{\min\{k, n\}} g_1(i) \times g_2(k-i) = g_{12}(k).$$

So if we compute the product  $\mathcal{G}_1(z)\mathcal{G}_2(z)$ , for each  $k \in [0, 2n]$ , the coefficient  $c_k$  of  $z^k$  in the product will give us the value of  $g_{12}(k)$ .

We know that two polynomials of degree at most  $n$  can be multiplied in  $\mathcal{O}(n \log n)$  time using FFT. Hence, all coefficients of  $\mathcal{G}_{12}(z)$ , and thus  $g_{12}(k)(= c_k)$  for all  $k \in [0, 2n]$  can be computed in  $\mathcal{O}(n \log n)$  time.

Use this page if you need additional space for your answers.

**QUESTION 2. [ 25 Points ] A Schönhage-Strassen-like Recurrence.** Consider the following recurrence (for  $n \geq 2$ ) which is similar to the recurrence that arises during the analysis of the *Schönhage-Strassen algorithm* for multiplying large integers.

$$T(n) = \begin{cases} \Theta(1) & \text{if } 2 \leq n \leq 8, \\ n^{\frac{2}{3}}T\left(n^{\frac{1}{3}}\right) + n^{\frac{1}{3}}T\left(n^{\frac{2}{3}}\right) + \Theta(n \log n) & \text{otherwise.} \end{cases}$$

2(a) [ 4 Points ] Show that the recurrence above can be rewritten as follows, where  $T(n) = nS(n)$ .

$$S(n) = \begin{cases} \Theta(1) & \text{if } 2 \leq n \leq 8, \\ S\left(n^{\frac{1}{3}}\right) + S\left(n^{\frac{2}{3}}\right) + \Theta(\log n) & \text{otherwise.} \end{cases}$$

**Solution.** Dividing both sides of the given recurrence for  $T(n)$  by  $n$ ,

$$\frac{T(n)}{n} = \begin{cases} \frac{\Theta(1)}{n} & \text{if } 2 \leq n \leq 8, \\ \frac{T\left(n^{\frac{1}{3}}\right)}{n^{\frac{1}{3}}} + \frac{T\left(n^{\frac{2}{3}}\right)}{n^{\frac{2}{3}}} + \Theta(\log n) & \text{otherwise.} \end{cases}$$

But  $\frac{T(n)}{n} = S(n)$ ,  $\frac{T\left(n^{\frac{1}{3}}\right)}{n^{\frac{1}{3}}} = S\left(n^{\frac{1}{3}}\right)$ ,  $\frac{T\left(n^{\frac{2}{3}}\right)}{n^{\frac{2}{3}}} = S\left(n^{\frac{2}{3}}\right)$ , and for  $n \in [2, 8]$ ,  $\frac{\Theta(1)}{n} = \Theta(1)$ . Hence, the recurrence above can be rewritten as follows.

$$S(n) = \begin{cases} \Theta(1) & \text{if } 2 \leq n \leq 8, \\ S\left(n^{\frac{1}{3}}\right) + S\left(n^{\frac{2}{3}}\right) + \Theta(\log n) & \text{otherwise.} \end{cases}$$

2(b) [ 4 Points ] Show that the recurrence in 2(a) can be rewritten as follows, where  $P(x) = S(2^x)$ .

$$P(x) = \begin{cases} \Theta(1) & \text{if } 1 \leq x \leq 3, \\ P\left(\frac{x}{3}\right) + P\left(\frac{2x}{3}\right) + \Theta(x) & \text{otherwise.} \end{cases}$$

**Solution.** Let  $n = 2^x$ , where  $x = \log_2 n$ . Then from the recurrence for  $S(n)$  we get:

$$S(2^x) = \begin{cases} \Theta(1) & \text{if } 2 \leq 2^x \leq 8, \\ S\left(2^{\frac{x}{3}}\right) + S\left(2^{\frac{2x}{3}}\right) + \Theta(x) & \text{otherwise.} \end{cases}$$

But  $S(2^x) = P(x)$ ,  $S\left(2^{\frac{x}{3}}\right) = P\left(\frac{x}{3}\right)$ ,  $S\left(2^{\frac{2x}{3}}\right) = P\left(\frac{2x}{3}\right)$ , and  $2 \leq 2^x \leq 8 \Rightarrow 1 \leq x \leq 3$ . Hence, the recurrence above can be rewritten as follows.

$$P(x) = \begin{cases} \Theta(1) & \text{if } 1 \leq x \leq 3, \\ P\left(\frac{x}{3}\right) + P\left(\frac{2x}{3}\right) + \Theta(x) & \text{otherwise.} \end{cases}$$

2(c) [ **9 Points** ] Solve the recurrence from part 2(b) to show that  $P(x) = \Theta(x \log x)$ .

**Solution.** The given recurrence is in the Akra-Bazzi form since for this recurrence:

$k = 2 \geq 1$  is an integer constant,

$$a_1 = 1 \geq 0, b_1 = \frac{1}{3} \in (0, 1),$$

$$a_2 = 1 \geq 0, b_2 = \frac{2}{3} \in (0, 1),$$

$x \geq 1$  is a real number,

$$x_0 = 3 \geq \max \left\{ \frac{1}{b_i}, \frac{1}{1 - b_i} \right\}, \text{ for } i \in \{1, 2\},$$

and  $g(x) = \Theta(x) = \Theta(x^1 \log^0 x)$ , which satisfies the polynomial growth condition.

Now in order to find the Akra-Bazzi solution for this recurrence, we need to find the unique real number  $p$  for which  $a_1 b_1^p + a_2 b_2^p = 1 \Rightarrow \left(\frac{1}{3}\right)^p + \left(\frac{2}{3}\right)^p = 1$ . This gives us  $p = 1$ .

Hence, the Akra-Bazzi solution:

$$\begin{aligned} P(x) &= \Theta \left( x^p \left( 1 + \int_1^x \frac{g(u)}{u^{p+1}} du \right) \right) \\ &= \Theta \left( x \left( 1 + \int_1^x \frac{u}{u^2} du \right) \right) && \{\cdot : p = 1\} \\ &= \Theta \left( x \left( 1 + \int_1^x \frac{1}{u} du \right) \right) \\ &= \Theta \left( x \left( 1 + [\ln x + c]_1^x \right) \right) && \left\{ \cdot : \int \frac{du}{u} = \ln u + c, \text{ where } c \text{ is a constant} \right\} \\ &= \Theta \left( x (1 + \ln x) \right) \\ &= \Theta(x \ln x) && \{\cdot : \ln x = \omega(1)\} \\ &= \Theta(x \log x) && \{\cdot : \ln x = \Theta(\log x) \text{ for any constant base of } \log x\} \end{aligned}$$

2(d) [ **8 Points** ] Use your results from part 2(c) to show that  $T(n) = \Theta(n \log n \log \log n)$ .

**Solution.** From part 2(b) we know:  $S(n) = P(\log n)$ .

But from part 2(c), we have:  $P(x) = \Theta(x \log x)$ .

Hence,  $S(n) = P(\log n) = \Theta(\log n \log \log n)$ .

Again from part 2(a) we know:  $T(n) = nS(n)$ .

Hence, using the solution for  $S(n)$  we have:  $T(n) = \Theta(n \log n \log \log n)$ .



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**QUESTION 3. [ 20 Points ] Closest Pair of Points.** Consider the algorithm CLOSEST-PAIR given below that finds the closest pair of points among a given set of points in the plane.

CLOSEST-PAIR(  $P, n$  )

**Input:** A set  $P = \{p_1 = (x_1, y_1), p_2 = (x_2, y_2), \dots, p_n = (x_n, y_n)\}$  of  $n$  points in the plane. Assume for simplicity that (a)  $n = 2^k$  for some integer  $k > 0$ , (b) all  $x_i$ 's are distinct, and (c) all  $y_i$ 's are distinct.

**Output:** Two distinct points  $p_i, p_j \in P$  such that the distance between  $p_i$  and  $p_j$  is the smallest among all pairs of points in  $P$ .

**Algorithm:**

1. **if**  $n = 2$  **then return**  $\langle p_1, p_2 \rangle$
2. **else**
3. Find a value  $x$  such that exactly  $\frac{n}{2}$  points in  $P$  have  $x_i < x$ , and the other  $\frac{n}{2}$  points have  $x_i > x$
4. Let  $L$  be the subset of  $P$  containing all points with  $x_i < x$
5. Let  $R$  be the subset of  $P$  containing all points with  $x_i > x$
6.  $\langle p_L, q_L \rangle \leftarrow \text{CLOSEST-PAIR}( L, \frac{n}{2} )$
7.  $\langle p_R, q_R \rangle \leftarrow \text{CLOSEST-PAIR}( R, \frac{n}{2} )$
8.  $d_L \leftarrow$  distance between  $p_L$  and  $q_L$
9.  $d_R \leftarrow$  distance between  $p_R$  and  $q_R$
10.  $d \leftarrow \min \{ d_L, d_R \}$
11. Scan  $P$  and remove each  $p_i = (x_i, y_i) \in P$  with  $x_i < x - d$  or  $x_i > x + d$
12. Sort the remaining points of  $P$  in increasing order of  $y$ -coordinates
13. Scan the sorted list, and for each point compute its distance to the 7 subsequent points in the list.  
Let  $\langle p_M, q_M \rangle$  be the closest pair of points found in this way.
14. Let  $\langle p, q \rangle$  be the closest pair among  $\langle p_L, q_L \rangle$ ,  $\langle p_R, q_R \rangle$  and  $\langle p_M, q_M \rangle$
15. **return**  $\langle p, q \rangle$

3(a) [ 10 Points ] Argue that for a set of  $n$  points, steps 3–5 take  $\mathcal{O}(n)$  time while steps 8–15 take  $\mathcal{O}(n \log n)$  time.

**Solution.** In step 3, we use the deterministic SELECT algorithm we saw in the class to find elements  $x'$  and  $x''$  such that  $\text{rank}(x', \{x_1, x_2, \dots, x_n\}) = \frac{n}{2}$  and  $\text{rank}(x'', \{x_1, x_2, \dots, x_n\}) = \frac{n}{2} + 1$ . This will take  $\mathcal{O}(n)$  time. Then we set  $x = \frac{x' + x''}{2}$ . Clearly, exactly  $\frac{n}{2}$  points in  $P$  will have  $x_i < x$ , and the other  $\frac{n}{2}$  points will have  $x_i > x$ . Steps 4–5 require scanning the points of  $P$  once, and hence take  $\mathcal{O}(n)$  time. Thus steps 3–5 take  $\mathcal{O}(n)$  worst-case time.

Steps 8–10 take only  $\mathcal{O}(1)$  time. Step 11 requires scanning  $P$  once and hence takes  $\mathcal{O}(n)$  time. Step 12 sorts the points of  $P$  which can be done in  $\mathcal{O}(n \log n)$  worst-case time using mergesort or heapsort (not standard quicksort as it takes  $\Theta(n^2)$  time in the worst case). Step 13 scans the sorted list in  $\mathcal{O}(n)$  time. Finally, steps 14–15 require only  $\mathcal{O}(1)$  time. Overall, steps 8–15 run in  $\mathcal{O}(n \log n)$  worst-case time.

Observe that since input size is  $n$  and one cannot sort  $n$  real numbers in less than  $\Theta(n \log n)$  time, the two bounds above are tight, i.e.,  $\Theta(n)$  and  $\Theta(n \log n)$ , respectively.

3(b) [ **10 Points** ] Let  $T(n)$  be the running time of CLOSEST-PAIR on a set of  $n$  points. Write a recurrence relation for  $T(n)$  and solve it.

**Solution.** Observe that CLOSEST-PAIR is a divide-and-conquer algorithm. We know from part 3(a) that for a set of  $n$  points the cost of divide (steps 3–5) is  $\Theta(n)$ , and the cost of combine (steps 8–15) is  $\Theta(n \log n)$ . Thus cost of divide and combine,  $f(n) = \Theta(n) + \Theta(n \log n) = \Theta(n \log n)$ . Also the algorithm divides a problem of size  $n$  into  $a = 2$  subproblems of size  $\frac{n}{b}$  each, where  $b = 2$ , and recursively solves (i.e., conquers) them in steps 6 and 7.

Hence, for  $n = 2^k$ , where  $k > 0$  is an integer, the recurrence for  $T(n)$  can be written as follows.

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 2, \\ 2T\left(\frac{n}{2}\right) + \Theta(n \log n) & \text{otherwise.} \end{cases}$$

Though the recurrence above has a base case size of  $n = 2$  and not  $n = 1$ , Master Theorem still applies. To see why, observe that since  $n$  is an even number,  $T(n)$  can also be described in terms of  $\frac{n}{2}$  (i.e., number of pairs of points) instead of  $n$  (i.e., number of points). Then  $T(n) = T'\left(\frac{n}{2}\right)$ , where for  $n = 2^l$  with integer  $l \geq 0$  the recurrence for  $T'(n)$  can be written as follows.

$$T'(n) = \begin{cases} \Theta(1) & \text{if } n = 1, \\ 2T'\left(\frac{n}{2}\right) + \Theta(n \log n) & \text{otherwise.} \end{cases}$$

One can apply case 2 of Master Theorem (see appendix) with  $a = b = 2$  and  $k = 1$ , and obtain  $T'(n) = \Theta\left(n^{\log_b a} \log^{k+1} n\right) = \Theta\left(n \log^2 n\right)$ .

Hence,  $T(n) = T'\left(\frac{n}{2}\right) = \Theta\left(n \log^2 n\right)$ .

Use this page if you need additional space for your answers.

**QUESTION 4. [ 10 Points ] An Impossible Priority Queue.** Consider a (comparison-based) priority queue  $Q$  that supports the following operations.

MAKE-QUEUE( $Q$ ): Create an empty queue  $Q$ .

INSERT( $Q, x$ ): Insert item  $x$  into  $Q$ .

INCREASE-KEY( $Q, x, k$ ): Increase the key of item  $x$  to  $k$  assuming  $k \geq$  current key of  $x$ .

FIND-MIN( $Q$ ): Return a pointer to an item in  $Q$  containing the smallest key.

DELETE-MIN( $Q$ ): Delete an item with the smallest key from  $Q$  and return a pointer to it.

4(a) [ 10 Points ] Suppose  $Q$  supports INSERT and INCREASE-KEY operations in  $\mathcal{O}(1)$  amortized time each, and DELETE-MIN operations in  $\mathcal{O}(\log n)$  worst-case time each, where  $n$  is the number of items in  $Q$ . It also supports the MAKE-QUEUE operation and every FIND-MIN operation in  $\mathcal{O}(1)$  worst-case time.

Argue that such a priority queue cannot exist.

**Solution.** We show below that if such a priority queue exists one can use it to sort a set of  $n$  numbers in  $\mathcal{O}(n)$  time which is impossible since no comparison-based sorting algorithm can sort  $n$  numbers asymptotically faster than  $\Theta(n \log n)$  time.

Let  $A[1 : n]$  be an array of  $n$  distinct unsorted numbers. We sort the numbers as follows.

1. MAKE-QUEUE( $Q$ ) *{create an empty priority queue}*
2. **for**  $i \leftarrow 1$  **to**  $n$  **do**
3.     create item  $x$  with  $key(x) = A[i]$
4.     INSERT( $Q, x$ ) *{insert  $A[i]$  into  $Q$ }*
5. **for**  $i \leftarrow 1$  **to**  $n$  **do**
6.      $x \leftarrow$  FIND-MIN( $Q$ ) *{find the item  $x$  with the smallest key in  $Q$ }*
7.      $A[i] \leftarrow key(x)$  *{ $A[i]$  now stores the  $i$ -th smallest number in the original input}*
8.     INCREASE-KEY( $Q, x, +\infty$ ) *{increase  $x$ 's key to something larger than the largest number in the input}*

Clearly, the algorithm above puts the input numbers (given in  $A$ ) back in  $A[1 : n]$  in sorted order. Now let us compute its running time. Step 1 takes  $\mathcal{O}(1)$  worst-case time. Since step 3 takes  $\mathcal{O}(1)$  worst-case time and step 4 takes  $\mathcal{O}(1)$  amortized time, the **for** loop in steps 2–4 takes  $\mathcal{O}(n)$  worst-case time. Steps 6 and 7 take  $\mathcal{O}(1)$  worst-case time and step 8 takes  $\mathcal{O}(1)$  amortized time. Hence, the **for** loop in steps 5–8 runs in  $\mathcal{O}(n)$  worst-case time. Thus sorting the  $n$  numbers requires only  $\mathcal{O}(n)$  time in the worst-case using this priority queue which clearly violates the known  $\Theta(n \log n)$  lower bound for comparison-based sorting. Hence, such a priority queue cannot exist.

Use this page if you need additional space for your answers.

## APPENDIX: RECURRENCES

**Master Theorem.** Let  $a \geq 1$  and  $b > 1$  be constants, let  $f(n)$  be a function, and let  $T(n)$  be defined on the nonnegative integers by the recurrence

$$T(n) = \begin{cases} \Theta(1), & \text{if } n \leq 1, \\ aT\left(\frac{n}{b}\right) + f(n), & \text{otherwise,} \end{cases}$$

where,  $\frac{n}{b}$  is interpreted to mean either  $\lfloor \frac{n}{b} \rfloor$  or  $\lceil \frac{n}{b} \rceil$ . Then  $T(n)$  has the following bounds:

**Case 1:** If  $f(n) = \mathcal{O}(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$ .

**Case 2:** If  $f(n) = \Theta(n^{\log_b a} \log^k n)$  for some constant  $k \geq 0$ , then  $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$ .

**Case 3:** If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and  $af\left(\frac{n}{b}\right) \leq cf(n)$  for some constant  $c < 1$  and all sufficiently large  $n$ , then  $T(n) = \Theta(f(n))$ .

**Akra-Bazzi Recurrences.** Consider the following recurrence:

$$T(x) = \begin{cases} \Theta(1), & \text{if } 1 \leq x \leq x_0, \\ \sum_{i=1}^k a_i T(b_i x) + g(x), & \text{otherwise,} \end{cases}$$

where,

1.  $k \geq 1$  is an integer constant,
2.  $a_i > 0$  is a constant for  $1 \leq i \leq k$ ,
3.  $b_i \in (0, 1)$  is a constant for  $1 \leq i \leq k$ ,
4.  $x \geq 1$  is a real number,
5.  $x_0$  is a constant and  $\geq \max \left\{ \frac{1}{b_i}, \frac{1}{1-b_i} \right\}$  for  $1 \leq i \leq k$ , and
6.  $g(x)$  is a nonnegative function that satisfies a polynomial growth condition (e.g.,  $g(x) = x^\alpha \log^\beta x$  satisfies the polynomial growth condition for any constants  $\alpha, \beta \in \mathbb{R}$ ).

Let  $p$  be the unique real number for which  $\sum_{i=1}^k a_i b_i^p = 1$ . Then

$$T(x) = \Theta \left( x^p \left( 1 + \int_1^x \frac{g(u)}{u^{p+1}} du \right) \right).$$

## APPENDIX: COMPUTING PRODUCTS

**Integer Multiplication.** Karatsuba's algorithm can multiply two  $n$ -bit integers in  $\Theta(n^{\log_2 3}) = \mathcal{O}(n^{1.6})$  time (improving over the standard  $\Theta(n^2)$  time algorithm).

**Matrix Multiplication.** Strassen's algorithm can multiply two  $n \times n$  matrices in  $\Theta(n^{\log_2 7}) = \mathcal{O}(n^{2.81})$  time (improving over the standard  $\Theta(n^3)$  time algorithm).

**Polynomial Multiplication.** One can multiply two  $n$ -degree polynomials in  $\Theta(n \log n)$  time using the FFT (Fast Fourier Transform) algorithm (improving over the standard  $\Theta(n^2)$  time algorithm).