

CS 341 — Alorithms

Kevin James

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1 Solving Recurrences

eg. mergesort

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

which we can solve to find

$$T(n) \in \Theta(n \log n)$$

eg. stoogesort($A[1..n]$)

```

if n <= 1:
    return
if A[1] > A[2]:
    swap(A[1], A[2])
stoogesort(A[1..⌊2n/3⌋])
stoogesort(A[⌊n/3⌋+1..n])
stoogesort(A[1..⌊2n/3⌋])

```

Let $T(n)$ be the runtime of stoogesort on n elements. Then

$$T(n) = \begin{cases} 3T\left(\frac{2n}{3}\right) + \Theta(1) & \text{if } n > 1 \\ \Theta(1) & \text{if } n = 1 \end{cases}$$

There are three methods for solving this recurrence:

1. Recursion Tree method: Expand for k iterations to get a tree of terms, set k to reach the base case, sum across rows, then over all levels.
2. Master method: Look up the answer. The Master Theorem: Let $T(n) = aT\left(\frac{n}{b}\right) + f(n)$ if $n > n_0$ and c if $n = n_0$. Set $d = \log_b a$ and pick a small constant $\varepsilon > 0$. Case 1: $f(n) = O(n^{d-\varepsilon}) \implies T(n) = \Theta(n^d)$. Case 2: $f(n) = \Theta(n^d) \implies T(n) = \Theta(n^d \log n)$. Case 3: $\lim_{n \rightarrow \infty} f(n)/n^{d+\varepsilon} = \infty \implies T(n) = \Theta(f(n))$.
3. Substitution method: Guess the form of the solution $T(n) \leq x$, then verify your guess by induction and fill in any constants.

Example:

$$\begin{aligned}
 T(n) &= 2T\left(\frac{n}{2}\right) + n^2, 7 \\
 T(n) &\leq cn^2 \\
 n = 1 : T(n) &= 7 \leq cn^2, c \geq 7 \\
 T\left(\frac{n}{2}\right) &\leq c\left(\frac{n}{2}\right)^2 \\
 T(n) &= 2T\left(\frac{n}{2}\right) + n^2
 \end{aligned}$$

$$\begin{aligned}
&\leq 2c \binom{n}{2} + n^2 \\
&= \frac{2cn^2}{4} + n^2 \\
&= \left(\frac{c}{2} + 1\right)n^2 \\
&\leq cn^2, \frac{c}{2} + 1 \leq c, c \geq 2
\end{aligned}$$

We can then pick $c = 7$ and solve as

$$T(n) \leq 7n^2 \implies T(n) \in O(n^2)$$

We can also note that $T(n) \geq n^2$, so $T(n) \in \Theta(n^2)$.

Example:

$$\begin{aligned}
T(n) &= 3T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + 4T\left(\left\lfloor \frac{n}{4} \right\rfloor\right) + 1, 1 \\
T(n) &\leq cn^2 \\
n = 1 : T(n) = 1 &\leq cn^2, c \geq 1 \\
T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) &\leq c \left\lfloor \frac{n}{2} \right\rfloor^2 \\
T(n) &= 3T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + 4T\left(\left\lfloor \frac{n}{4} \right\rfloor\right) + 1 \\
&\leq 3c \left\lfloor \frac{n}{2} \right\rfloor^2 + 4c \left\lfloor \frac{n}{4} \right\rfloor^2 + 1 \\
&\leq 3c \left(\frac{n}{2}\right)^2 + 4c \left(\frac{n}{2}\right)^2 + 1 \\
&= \left(\frac{3}{4} + \frac{4}{16}\right)cn^2 + 1 \\
&= cn^2 + 1
\end{aligned}$$

but since we could not get rid of the constant, we try

$$\begin{aligned}
T(n) &\leq cn^2 - c_0 \\
n = 1 : T(n) = 1 &\leq cn^2 - c_0, c \geq 1 + c_0 \\
T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) &\leq c \left\lfloor \frac{n}{2} \right\rfloor^2 - c_0 \\
T(n) &= 3T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + 4T\left(\left\lfloor \frac{n}{4} \right\rfloor\right) + 1 \\
&\leq 3\left(c \left\lfloor \frac{n}{2} \right\rfloor^2 - c_0\right) + 4\left(c \left\lfloor \frac{n}{4} \right\rfloor^2 - c_0\right) + 1 \\
&\leq 3c \left(\frac{n}{2}\right)^2 - 3c_0 + 4c \left(\frac{n}{2}\right)^2 - 4c_0 + 1
\end{aligned}$$

$$\begin{aligned}
&= \left(\frac{3}{4} + \frac{4}{16}\right)cn^2 - 7c_0 + 1 \\
&= cn^2 - c_0, -7c_0 + 1 \leq -c_0, c_0 \geq \frac{1}{6}
\end{aligned}$$

Pick $c_0 = \frac{1}{6}, c = \frac{7}{6}$ and we see that

$$T(n) \leq \frac{7}{6}n^2 - \frac{1}{6} \implies T(n) \in O(n^2)$$

2 Algorithm Design Techniques

2.1 Divide and Conquer

Divide your problem into subproblems of the same type, then use recursion to solve each problem and combine the results.

Problem (Maxima): Given a set P of n points in 2D, we say point p dominates point q if and only if p has both a greater x and y value than q . We say point q is maximal if and only if $q \in P$ and no point in P dominates q . Find all maximal points.

Solutions:

- Brute Force: for each $q \in P$, check if no points dominate q . Total time: $\Theta(n^2)$
- Divide and Conquer: divide into two subarrays of size $\frac{n}{2}$.
- Divide by Medians: instead of dividing by size, divide by a median vertical line.

```

maxima(sorted[p_1..p_n]):
1. if n == 1: return p_1
2. [q_1..q_l] = maxima([p_1..p_{n/2}])
3. [s_1..s_m] = maxima([p_{n/2}..p_n])
4. i = 1
5. while q_i.y > s_1.y
6.   i += 1
7. return [q_1..q_l, s_1..s_m]

```

which is an $O(n \log n)$ algorithm.

Problem (Closest Pair): Given set P of n points in 2D, find a pair $p, q \in P$ such that these points have a smaller distance between them than any other pair of points in P , ie. $d(p, q) = \sqrt{(p.x - q.x)^2 + (p.y - q.y)^2}$.

Solutions:

- Brute Force Algorithm: $\Theta(n^2)$
- Shamos' Algorithm: $\Theta(n \log^2 n)$. Note that if we refine the Shamos algorithm by pre-sorting P also by y -coordinate at the beginning, we find that our complexity becomes $O(n \log n)$.

```

def bruteForce(P):
    distance = infinity
    for each p in P:
        for each q in P (q neq P):
            distance = min(distance, d(p,q))
    return distance

def shamos(P):
    if n <= 10:
        return bruteForce(P)
    x_m = median x_coordinate
    P_L = { p in P : p.x < x_m }
    P_R = { p in P : p.x > x_m }
    d_L = shamos(P_L)
    d_R = shamos(P_R)
    d = min(d_L, d_R)
    <p_1..p_m> = sorted_by(points in { p in P : x_m - d <= p.x <= x_m + d }, y)
    for i = 1 to m do
        j = i + 1
        while p_j.y <= p_i.y + d:
            d = min(d, d(p_i, p_j))
            j++
    return d

```

2.1.1 Multiplying Large Numbers

Given two n -bit numbers $A = a_{n-1}, a_{n-2}, \dots, a_0$, $b = b_{n-1}, b_{n-2}, \dots, b_0$ in binary, compute $AB = c_{n-1}, c_{n-2}, \dots, c_0$.

The “Elementary School” algorithm for this involves doing

```

      1011
x 1101
-----
      1011
     1011
    1011
   -----
  10001111

```

which is $O(n)$ shifts and $O(n)$ additions, thus making it an $O(n^2)$ algorithm.

The “Karatsuba and Ofman” algorithm involves a different process:

```

A' = [a_{n-1}..a_{n/2}]
A" = [a_{n/2 - 1}..a_0]
A = [A'..A"]

```

$$B' = [b_{\{n-1\}} \dots b_{\{n/2\}}]$$

$$B'' = [b_{\{n/2 - 1\}} \dots b_0]$$

$$B = [B' \dots B'']$$

$$AB = A'B'2^n + (A'B'' + A''B')2^{(n/2)} + A''B''$$

which gives us a complexity of $O(n^2)$.