

Solving Recurrences

1. Techniques

- Divide and Conquer (recursion tree) – powerful
- Master Method/Master Theorem – to solve “divide”
- Annihilators – to solve “Minus”: $T(n) = T(n - a) + T(n - b) \dots + f(n)$
- Ultimate tech: Guess and Check!

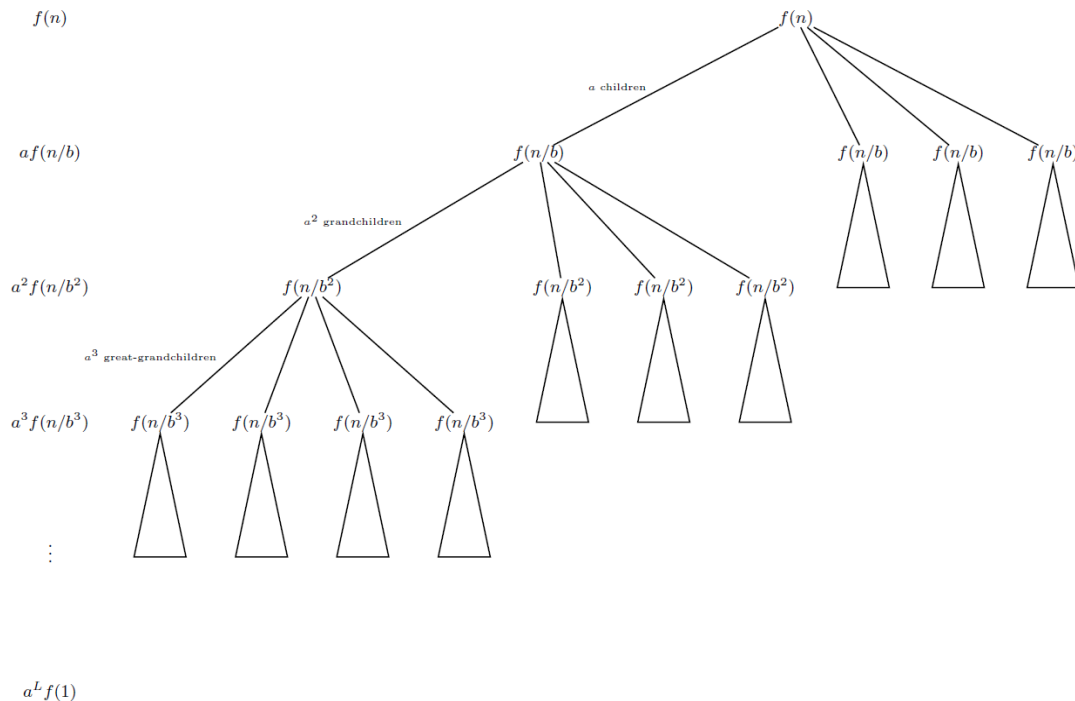
2. Master Theorem

The Master Theorem. The recurrence $T(n) = aT(n/b) + f(n)$ can be solved as follows.

- If $af(n/b)/f(n) < 1$, then $T(n) = \Theta(f(n))$.
- If $af(n/b)/f(n) > 1$, then $T(n) = \Theta(n^{\log_b a})$.
- If $af(n/b)/f(n) = 1$, then $T(n) = \Theta(f(n) \log_b n)$.
- If none of these three cases apply, you're on your own.

- It works when $f(n)$ is polynomial
- We use divide and conquer to prove it's correct

3. Using Divide and Conquer to prove Master Theorem



- Draw the recursion tree for $T(n) = f(n) + aT(n/b) + f(n)$, the tree's depth is $\log_b n$
- $T(n) = f(n) + af\left(\frac{n}{b}\right) + a^2f\left(\frac{n}{b^2}\right) + \dots + a^i f\left(\frac{n}{b^i}\right) + \dots + a^H f\left(\frac{n}{b^H}\right)$

- $H = \log_b n$, because on the last layer $\frac{n}{b^H} = 1$
- We consider this as a geometric series: because $f(n)$ is polynomial, and $f(n) = \Theta(n^k) \rightarrow \text{common ratio} = \Theta\left(\frac{a}{b^k}\right)$
- If $a \frac{f(\frac{n}{b})}{f(n)} < 1$, or the common ratio is between 0 and 1; $f(n)$ will be the largest term, so $T(n) = \Theta(f(n))$.
- If $a \frac{f(\frac{n}{b})}{f(n)} > 1$, or the common ratio is greater than 1, and the last term becomes the largest. So $T(n) = \Theta(a^{\log_b n})$
- If $a \frac{f(\frac{n}{b})}{f(n)} = 1$, we have a lot of $f(n)$ here, so $T(n) = \Theta(\log_b n \cdot f(n))$

4. Examples

- Merge Sort $T(n) = 2T\left(\frac{n}{2}\right) + n$
 - $f(n)$ is polynomial, we can use Master
 - $a = 2, b = 2 \rightarrow \Theta(n \lg n)$
- $T(n) = 4T\left(\frac{n}{2}\right) + n \lg n$
 - $f(n)$ is not polynomial, and $a f\left(\frac{n}{b}\right) = 2n \lg n - 2n$
 - We can say $2n \lg n > 2n \lg n - 2n > c_1 n \lg n$, $c_1 > 1$ when n is really large.
 - Thus, we can use the second case in Master $\rightarrow \Theta(4^{\lg n}) = \Theta(n^2)$
- $T(n) = T\left(\frac{3n}{4}\right) + T\left(\frac{n}{4}\right) + n$
 - Divide and Conquer
 - (Draw the tree) And we can see each complete level sum up to n .
 - Shortest branch has depth $\log_4 n$, longest has depth $\log_{\frac{4}{3}} n$.
 - $n \log_4 n \leq T(n) \leq n \log_{\frac{4}{3}} n \rightarrow \Theta(?)$
- $T(n) = 2T\left(\frac{n}{2}\right) + n / \lg n$
 - $f(n)$ is not polynomial, and $a f\left(\frac{n}{b}\right) = \frac{n}{\lg n - 1}$
 - We can't say it's geometric, so let's try Divide and conquer
 - We can see each layer sum up to $\frac{n}{\lg n - i}$ for the i^{th} level
 - $H < \lg n - 1 (\approx \lg n - 2)$
 - So it will sum up to $\Theta(?)$
- Hanoi Tower $T(n) = 2T(n-1) + 1$
 - Divide and Conquer gives us $\Theta(2^n - 1)$
 - Talk a little bit about annihilator if there's time
- $T(n) = \sqrt{n} \times T(\sqrt{n}) + n$
 - Let's guess! And check the upper bound and lower bound.
 - Check whether $b \times g(n) \leq T(n) \leq a \times g(n)$
 - Try $\Theta(n^2)$, first let's check if it's okay for upper bound:

Insert $T(n) = O(n^2)$, we have $T(n) \leq \sqrt{n} \times a \times n + n = a \times n\sqrt{n} + n \leq a \times n^2$. When n is large enough, we can find this constant a for sure, so n^2 is good for an upper bound.

Then let's check if it's okay for lower bound:

Insert $T(n) = \Omega(n^2)$, we have $T(n) \geq \sqrt{n} \times b \times n + n = b \times n\sqrt{n} + n \not\geq b \times n^2$. When n is large enough, we can't find this constant b for sure, so n^2 is not good for a lower bound, which means we need something grows more slowly.

- Let's try $\Theta(n)$:

$$T(n) \leq \sqrt{n} \times a \times \sqrt{n} + n = (a + 1) \times n \not\leq a \times n$$

We can see $\Theta(n)$ grows too slowly so need something faster.

- Maybe $\Theta(n \log n)$? Keep narrowing it down!