Outline

- Compare the growth rate of functions
- 2 Limit Test
- 3 L'Hospital Rule
- 4 Stirling Formula
- Summations
- Integration Method
- Solving Linear Recursive Equations

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- By using the definitions, we can directly show whether $T_1(n) = O(T_2(n))$, or $T_1(n) = \Omega(T_2(n))$. However, it is not easy to prove the relationship of two functions in this way.

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Limit Test

Limit Test is a powerful method for comparing functions.

Limit Test

Let $T_1(n)$ and $T_2(n)$ be two functions. Let $c=\lim_{n\to\infty} \frac{T_1(n)}{T_2(n)}$.

- **1** If c is a constant > 0, then $T_1(n) = \Theta(T_2(n))$.
- 2 If c = 0, then $T_1(n) = o(T_2(n))$.
- 3 If $c = \infty$, then $T_1(n) = \omega(T_2(n))$.
- 4 If c does not exists (or if we do not know how to compute c), the limit test fails.

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- If c does not exists (or if we do not know how to compute c), the limit test fails.

Proof of (1): $c = \lim_{n \to \infty} \frac{T_1(n)}{T_2(n)}$ means: $\forall \epsilon > 0$, there exists $n_0 \ge 0$ such that for any $n \ge n_0$: $\left| \frac{T_1(n)}{T_2(n)} - c \right| \le \epsilon$; or equivalently: $c - \epsilon \le \frac{T_1(n)}{T_2(n)} \le c + \epsilon$. Let $\epsilon = c/2$ and let $c_1 = c - \epsilon = c/2$ and $c_2 = c + \epsilon = 3c/2$, we have

$$c_1T_2(n) \leq T_1(n) \leq c_2T_2(n)$$

for all $n \ge n_0$. Thus $T_1(n) = \Theta(T_2(n))$ by definition.

Example 1

$$T_1(n) = 10n^2 + 15n - 60, T_2(n) = n^2$$

$$\lim_{n \to \infty} \frac{T_1(n)}{T_2(n)} = \lim_{n \to \infty} \frac{10n^2 + 15n - 60}{n^2} = \lim_{n \to \infty} (10 + \frac{15}{n} - \frac{60}{n^2}) = 10 + 0 - 0 = 10$$

Since 10 is a constant > 0, we have $T_1(n) = \Theta(T_2(n)) = \Theta(n^2)$ by the statement 1 of Limit Test (as expected).

Log function

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$$lg = log_2 n$$

$$log n = log_{10} n$$

$$ln n = log_e n$$

($\ln n$ is the log function with the natural base e = 2.71828...).

Log base change formula

For any
$$1 < a, b$$
, $\log_b n = \frac{\log_a n}{\log_a b} = \log_b a \cdot \log_a n$.

Proof: Let $k = \log_b n$. By definition: $n = b^k$.

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This implies: $\log_b n = k = \frac{\log_a n}{\log_a b}$.

Let n = a in this formula and note $1 = \log_a a$:

$$\log_b a = \frac{\log_a a}{\log_a b} = \frac{1}{\log_a b}$$

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Let n = a in this formula and note $1 = \log_a a$:

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This proves the second part of the formula.



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L'Hospital Rule

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• If $\lim_{n\to\infty} f(n) = 0$ and $\lim_{n\to\infty} g(n) = 0$, then

$$\lim_{n \to \infty} \frac{f(n)}{g(n)} = \lim_{n \to \infty} \frac{f'(n)}{g'(n)}$$

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• If $\lim_{n\to\infty} f(n) = \infty$ and $\lim_{n\to\infty} g(n) = \infty$, then

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=\lim_{n\to\infty}\frac{f'(n)}{g'(n)}$$

Example 2

$$T_1(n) = n^2 + 6$$
, $T_2(n) = n \lg n$. (Recall: $\lg n = \log_2 n$.)

$$\begin{split} \lim_{n \to \infty} \frac{T_1(n)}{T_2(n)} &= \lim_{n \to \infty} \frac{n^2 + 6}{n \lg n} = \lim_{n \to \infty} \frac{n + \frac{6}{n}}{\lg n} \\ &= \lim_{n \to \infty} \frac{1 - \frac{6}{n^2}}{\frac{1}{\ln 2 \cdot n}} \text{ (by L'Hospital Rule)} \\ &= \ln 2 \lim_{n \to \infty} (n - \frac{6}{n}) = \ln 2(\infty - 0) = \infty \end{split}$$

By Limit Test, we have $n^2 + 6 = \omega(n \lg n)$.



Example 3

 $T_1(n) = (\ln n)^k$, $T_2(n) = n^{\epsilon}$, where k > 0 is any (large) constant and $\epsilon > 0$ is any (small) constant. (Recall: $\ln n = \log_{\epsilon} n$.)

$$\begin{split} &\lim_{n\to\infty}\frac{T_1(n)}{T_2(n)}=\lim_{n\to\infty}\frac{(\ln n)^k}{n^\epsilon} \text{ (use L'Hospital Rule)}\\ &=\lim_{n\to\infty}\frac{k(\ln n)^{k-1}\times (1/n)}{\epsilon n^{(\epsilon-1)}}\\ &=\frac{k}{\epsilon}\lim_{n\to\infty}\frac{(\ln n)^{k-1}}{n^\epsilon} \text{ (use L'Hospital Rule again and simplify)}\\ &=\frac{k(k-1)}{\epsilon^2}\lim_{n\to\infty}\frac{(\ln n)^{k-2}}{n^\epsilon} \text{ (use L'Hospital Rule }k \text{ times)}\\ &\dots\\ &=\frac{k(k-1)\cdots 2\cdot 1}{\epsilon^k}\lim_{n\to\infty}\frac{1}{n^\epsilon}=0 \end{split}$$

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So by Limit Test, $(\ln n)^k = o(n^\epsilon)$ for any k and ϵ . For example, take k=100 and $\epsilon=0.01$, we have $(\ln n)^{100}=o(n^{0.01})$.

Example 4

 $T_1(n)=n^k$, $T_2(n)=a^n$, where k>0 is any (large) constant and a>1 is any constant bigger than 1.

$$\begin{split} &\lim_{n\to\infty}\frac{T_1(n)}{T_2(n)}=\lim_{n\to\infty}\frac{n^k}{a^n}\text{ (using L'Hospital Rule)}\\ &=\lim_{n\to\infty}\frac{k\cdot n^{k-1}}{\ln a\cdot a^n}=\frac{k}{\ln a}\lim_{n\to\infty}\frac{n^{k-1}}{a^n}\text{ (using L'Hospital Rule }k\text{ times)}\\ &=\frac{k(k-1)\cdots 2\cdot 1}{(\ln a)^k}\lim_{n\to\infty}\frac{n^0}{a^n}\\ &=\frac{k(k-1)\cdots 2\cdot 1}{(\ln a)^k}\lim_{n\to\infty}\frac{1}{a^n}=0 \end{split}$$

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So by Limit Test, $n^k = o(a^n)$ for any k > 0 and a > 1. For example, take k = 1000 and a = 1.001, we have $n^{1000} = o((1.001)^n)$.



Example 5

 $T_1(n) = \log_a n$, $T_2(n) = \log_b n$, where a > 1 and b > 1 are any two constants bigger than 1.

By the Log Base Change Formula: $\log_b n = \log_b a \cdot \log_a n$

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Since $\frac{1}{\log_b a} > 0$ is a constant, we have $\log_a n = \Theta(\log_b n)$ by Limit Test.

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Since $\frac{1}{\log_b a} > 0$ is a constant, we have $\log_a n = \Theta(\log_b n)$ by Limit Test.

So: the growth rates of the \log functions are the same for any base > 1.



Example 6

 $T_1(n) = a^n$, $T_2(n) = b^n$, where 1 < a < b are any two constants.

We have: $\lim_{n\to\infty} \frac{T_1(n)}{T_2(n)} = \lim_{n\to\infty} \frac{a^n}{b^n} = \lim_{n\to\infty} (\frac{a}{b})^n = 0$.

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The list of common functions:

The following list shows the functions commonly used in algorithm analysis, in the order of increasing growth rate (a,b,c,d,k,ϵ) are positive constants, $\epsilon < 1, k > 1, d > 1$ and a < b):

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$$c, \log_d n, (\log_d n)^k, n^\epsilon, n, n^k, a^n, b^n, n!, n^n$$

in the sense that if f(n) and g(n) are any two consecutive functions in the list, we have f(n) = o(g(n))

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$$T_1(n) = n!$$
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$$\frac{a^n}{n!} = \underbrace{\frac{a}{1} \cdot \frac{a}{2} \cdots \frac{a}{2\lceil a \rceil}}_{2\lceil a \rceil \text{ terms}} \cdot \underbrace{\frac{a}{2\lceil a \rceil + 1} \cdots \frac{a}{n}}_{(n-2\lceil a \rceil) \text{ terms}}$$

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The first part is a constant c > 0. In the second part, each term < 1/2. So:

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$$0 \le \lim_{n \to \infty} \frac{a^n}{n!} \le c \cdot \lim_{n \to \infty} (\frac{1}{2})^{(n-2\lceil a \rceil)} = 0$$



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By Limit Test: $a^n = o(n!)$.



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Stirling Formula

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$$\sqrt{2\pi n} \left(\frac{n}{e}\right)^n e^{\frac{1}{12n+1}} \leq n! \leq \sqrt{2\pi n} \left(\frac{n}{e}\right)^n e^{\frac{1}{12n}}$$

or:
$$n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \cdot \left(1 + \Theta\left(\frac{1}{n}\right)\right)$$

When n = 10;

- n! = 3628800.
- $\sqrt{2\pi n} \left(\frac{n}{e}\right)^n = 3598696, 99\%$ accurate.

Example 7 (another solution)

$$T_1(n) = n!$$
 and $T_2(n) = a^n$ $(a > 1)$

$$\lim_{n \to \infty} \frac{n!}{a^n} \ge \lim_{n \to \infty} \sqrt{2\pi n} \left(\frac{n}{ae}\right)^n = \lim_{n \to \infty} \sqrt{2\pi n} \cdot \lim_{n \to \infty} \left(\frac{n}{ae}\right)^n$$

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The first limit is ∞ . The second limit goes to ∞^{∞} . So it's also ∞ . Thus $\lim_{n\to\infty}\frac{n!}{a^n}=\infty$ and $n!=\omega(a^n)$ by limit test.

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Example 8

$$T_1(n) = n^n \text{ and } T_2(n) = n!$$

By using similar method as in Example 7, we can show: $n! = o(n^n)$



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Summations

Consider the following simple program:

- 1: for i = 1 to n do
- 2: the loop body takes $\Theta(i^k)$ time
- 3: end for

What's the runtime of this program? It should be:

$$T(n) = \sum_{i=1}^{n} \Theta(i^k) = c \sum_{i=1}^{n} i^k$$
 (for some constant c)

Thus, it is important to know the sum of the form $\sum_{i=1}^{n} i^{k}$.



Summation formulas

$$\sum_{i=1}^{n} i^{1} = 1 + 2 \dots + n = \frac{n(n+1)}{2} = \Theta(n^{2})$$
 (1)

$$\sum_{i=1}^{n} i^2 = 1^2 + 2^2 \dots + n^2 = \frac{n(n+1)(2n+1)}{6} = \Theta(n^3)$$
 (2)

$$\sum_{i=1}^{n} i^3 = 1^3 + 2^3 \dots + n^3 = \frac{n^2(n+1)^2}{4} = \Theta(n^4)$$
 (3)

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$$\sum_{i=1}^{n} i^3 = 1^3 + 2^3 \dots + n^3 = \frac{n^2(n+1)^2}{4} = \Theta(n^4)$$
 (3)

In general, for any k > 0, the following is true.

$$\sum_{k=1}^{n} i^{k} = \Theta(n^{k+1}) \tag{4}$$



Summations:

By using these formulas, we can compute the runtime of nested loops.

Example

```
\begin{array}{l} \text{for } i=1 \text{ to } n \text{ do} \\ \text{for } j=i \text{ to } n \text{ do} \\ \text{for } k=i \text{ to } j \text{ do} \\ (\dots \text{ loop body takes } \Theta(1) \text{ time.}) \\ \text{end for} \\ \text{end for} \\ \text{end for} \end{array}
```

Since the inner loop body takes $\Theta(1)$ time, we only need to count the number D(n) of the inner loop iterations. Then $T(n) = D(n) \cdot \Theta(1) = \Theta(D(n))$.

$$D(n) = \sum_{i=1}^{n} \sum_{j=i}^{n} \sum_{k=i}^{j} 1 = \sum_{i=1}^{n} \sum_{j=i}^{n} (j-i+1)$$

Calculate D(n)

To calculate the second sum, let t = j - i + 1. When j = i, t = 1. When j = n, t = n - i + 1. Thus

$$\sum_{j=i}^{n} (j-i+1) = \sum_{t=1}^{n-i+1} t = 1 + 2 + \dots + (n-i+1) = \frac{(n-i+2)(n-i+1)}{2}$$

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Next we calculate: $\sum_{i=1}^{n} \frac{(n-i+2)(n-i+1)}{2}$. Let s=n-i+1. When i=1, s=n. When i=n, s=1. Thus:

$$\sum = \sum_{s=1}^{n} \frac{(s+1)s}{2} = \frac{1}{2} \{ \sum_{s=1}^{n} s^2 + \sum_{s=1}^{n} s \}$$
$$= \frac{1}{2} \{ \frac{n(n+1)(2n+1)}{6} + \frac{n(n+1)}{2} \} = \Theta(n^3)$$



More Summations:

The following summation formulas are useful.

$$\sum_{i=0}^{n} a^{i} = 1 + a + a^{2} + \dots + a^{n} = \begin{cases} \frac{1 - a^{n+1}}{1 - a} &= \Theta(1) & \text{if } 0 < a < 1\\ n + 1 &= \Theta(n) & \text{if } a = 1\\ \frac{a^{n+1} - 1}{a - 1} &= \Theta(a^{n}) & \text{if } 1 < a \end{cases}$$
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How to compute H(n)?



Outline

- Compare the growth rate of functions
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Integration Method

Let f(x) be an increasing function. Then for any $a \le b$:

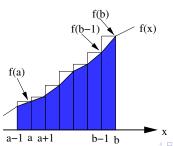
$$\int_{a-1}^{b} f(x)dx \le \sum_{i=a}^{b} f(i) \le \int_{a}^{b+1} f(x)dx$$

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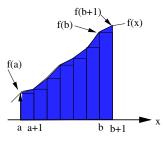
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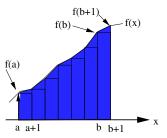
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Similarly:

Let f(x) be a decreasing function. Then for any $a \le b$:

$$\int_{a-1}^{b} f(x)dx \ge \sum_{i=a}^{b} f(i) \ge \int_{a}^{b+1} f(x)dx$$

Example

For any k > 0, $f(x) = x^k$ is an increasing function. Let a = 1 and b = n.

$$\int_{0}^{n} x^{k} dx \le \sum_{i=1}^{n} i^{k} \le \int_{1}^{n+1} x^{k} dx$$

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By limit test, both lower and upper bounds $= \Theta(n^{k+1})$. Thus $\sum_{i=1}^{n} i^k = \Theta(n^{k+1})$.



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Note: $\lim_{n\to\infty} (\ln n - \sum_{i=1}^n \frac{1}{i}) = c$, where c = 0.577... is Euler constant.



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How to compute Fib_n directly from n?

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Example 2: $f_0 = 1$, $f_1 = 2$, $f_2 = 4$ and for all $n \ge 0$, $f_{n+3} = 3f_{n+1} - 2f_n$ Then $\{f_n\}$ is a linear recursive sequence of order 3 where $c_2 = 0$, $c_1 = 3$ and $c_0 = -2$.



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- Say $\alpha_1 = \alpha_2 = \ldots = \alpha_t$ repeats t times.
- Then in the solution formula, the portion

$$\cdots a_1(\alpha_1)^n + a_2(\alpha_2)^n + \cdots + a_t(\alpha_t)^n \cdots$$

is replaced by:

$$\cdots a_1(\alpha_1)^n + a_2n^1(\alpha_1)^n + \cdots + a_tn^{t-1}(\alpha_1)^n \cdots$$

Other steps are the same.



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- Thus: $F_n = \frac{8}{9} + \frac{4}{3}n + \frac{1}{9}(-2)^n$

