

Algorithm

Mathematical Background

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Floor and Ceiling Functions

• The *floor* function maps any real number x onto the greatest integer less than or equal to x:

$$\begin{bmatrix} 3.2 \end{bmatrix} = \begin{bmatrix} 3 \end{bmatrix} = 3$$
$$\begin{bmatrix} -5.2 \end{bmatrix} = \begin{bmatrix} -6 \end{bmatrix} = -6$$

• The *ceiling* function maps x onto the least integer greater than or equal to x:

- The <cmath> library implements these as
 - double floor(double); double ceil(double);

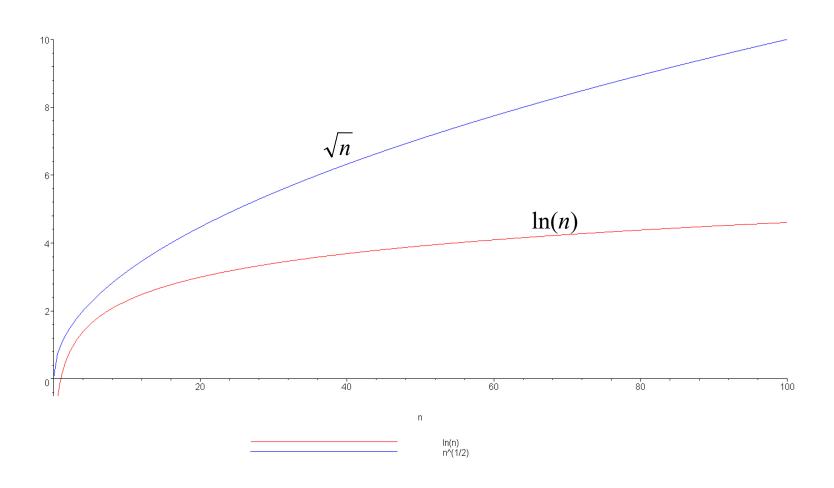
- If $n = e^m$, we define $m = \ln(n)$
 - It is always true that $e^{\ln(n)} = n$
- Exponentials grow faster than any non-constant polynomial

$$\lim_{n\to\infty}\frac{e^n}{n^d}=\infty$$

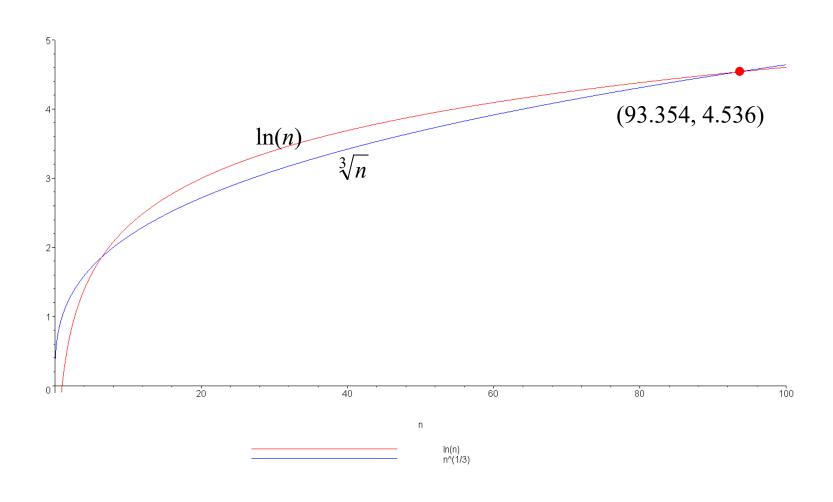
- for any d > 0
- Thus, their inverses—logarithms—grow slower than any polynomial

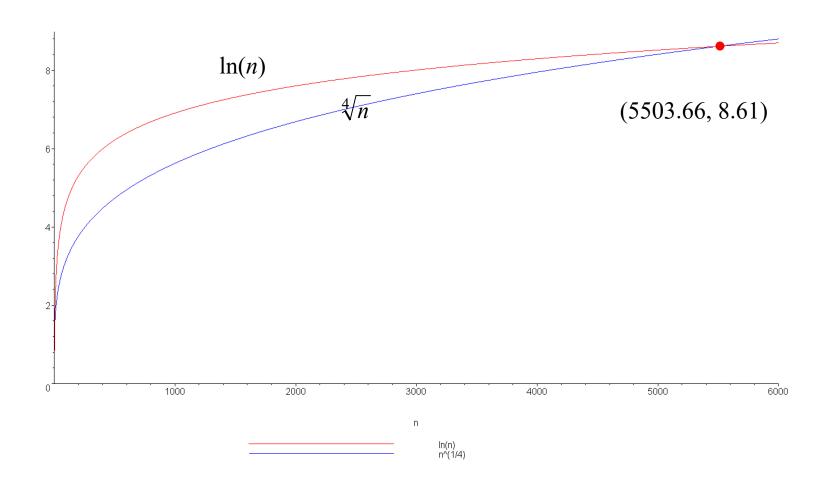
$$\lim_{n\to\infty}\frac{\ln(n)}{n^d}=0$$

Example: $f(n) = n^{1/2} = \sqrt{n}$ is strictly greater than $\ln(n)$



 $f(n) = n^{1/3} = \sqrt[3]{n}$ grows slower but only up to n = 93



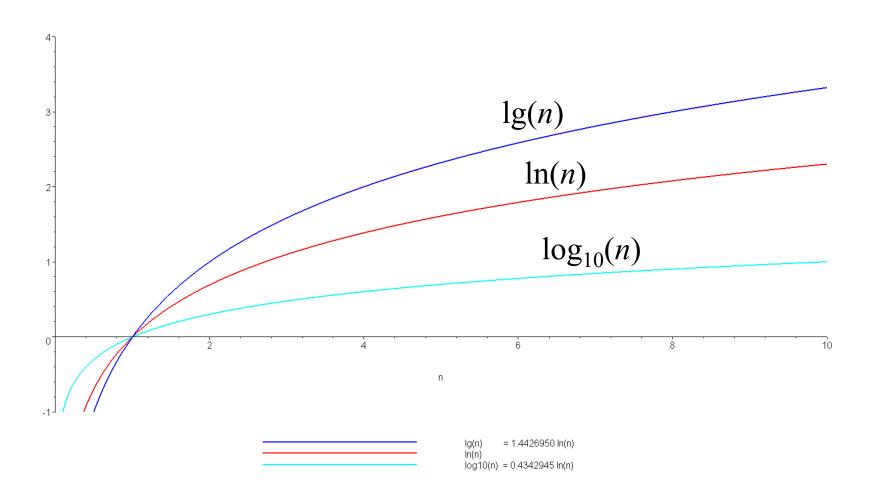


- We have compared logarithms and polynomials
 - How about $\log_2(n)$, $\ln(n)$, and $\log_{10}(n)$
- You have seen the formula:

$$\log_b(n) = \frac{\ln(n)}{\ln(b)}$$

- where, ln(b) is a constant
- All logarithms are scalar multiples of each others

- A plot of $\log_2(n) = \lg(n), \ln(n), \text{ and } \log_{10}(n)$



- Note: the base-2 logarithm $log_2(n)$ is written as lg(n)
- It is an industry standard to implement the natural logarithm $\ln(n)$ as

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double log( double );
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• The *common* logarithm $\log_{10}(n)$ is implemented as

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double log10( double );
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You should also be aware of the relationship:

$$lg(2^{10}) = lg(1024) = 10$$
 $lg(2^{20}) = lg(1048576) = 20$
 $lg(10^3) = lg(1000) \approx 10 \text{ kilo}$
 $lg(10^6) = lg(1000000) \approx 20 \text{ mega}$
 $lg(10^9) \approx 30 \text{ giga}$
 $lg(10^{12}) \approx 40 \text{ tera}$

L'Hôpital's Rule

If you are attempting to determine

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

■ But both $\lim_{n\to\infty} f(n) = \lim_{n\to\infty} g(n) = \infty$, it follows

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=\lim_{n\to\infty}\frac{f^{(1)}(n)}{g^{(1)}(n)}$$

Note: $f^{(k)}(n)$ is the k^{th} derivative

Repeat as necessary

Arithmetic Series

Each term in an arithmetic series is increased by a constant value (usually 1):

$$0+1+2+3+\cdots+n=\sum_{k=0}^{n}k=\frac{n(n+1)}{2}$$

Proof 1: write out the series twice and add each column

$$1 + 2 + 3 + \dots + n - 2 + n - 1 + n
+ n + n - 1 + n - 2 + \dots + 3 + 2 + 1
(n+1) + (n+1) + (n+1) + \dots + (n+1) + (n+1) + (n+1)$$

$$= n (n+1)$$

Since we added the series twice, we must divide the result by 2

Arithmetic Series

Each term in an arithmetic series is increased by a constant value (usually 1):

$$0+1+2+3+\cdots+n=\sum_{k=0}^{n}k=\frac{n(n+1)}{2}$$

- Proof 2: By mathematical induction
 - The statement is true for n = 0: $\sum_{i=0}^{0} k = 0 = \frac{0 \cdot 1}{2} = \frac{0(0+1)}{2}$
 - Assume that the statement is true for an arbitrary n: $\sum_{k=0}^{n} k = \frac{n(n+1)}{2}$

For
$$n+1$$
, we have:
$$\sum_{k=0}^{n+1} k = (n+1) + \sum_{i=0}^{n} k = (n+1) + \frac{n(n+1)}{2}$$
$$= \frac{(n+1)2 + (n+1)n}{2}$$
$$= \frac{(n+1)(n+2)}{2}$$

$$\sum_{k=0}^{n} k^2 = \frac{n(n+1)(2n+1)}{6}$$

$$\sum_{k=0}^{n} k^3 = \frac{n^2 (n+1)^2}{4}$$

However, it is easier to see the pattern

$$\sum_{k=0}^{n} k = \frac{n(n+1)}{2} \approx \frac{n^2}{2}$$

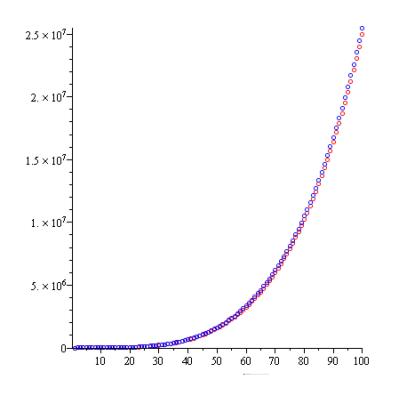
$$\sum_{k=0}^{n} k = \frac{n(n+1)}{2} \approx \frac{n^2}{2} \qquad \sum_{k=0}^{n} k^2 = \frac{n(n+1)(2n+1)}{6} \approx \frac{n^3}{3}$$

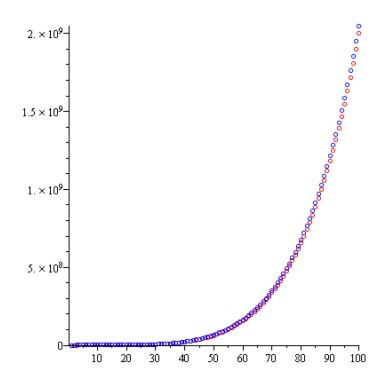
$$\sum_{k=0}^{n} k^{3} = \frac{n^{2} (n+1)^{2}}{4} \approx \frac{n^{4}}{4}$$

We can generalize this formula

$$\sum_{k=0}^{n} k^d \approx \frac{n^{d+1}}{d+1}$$

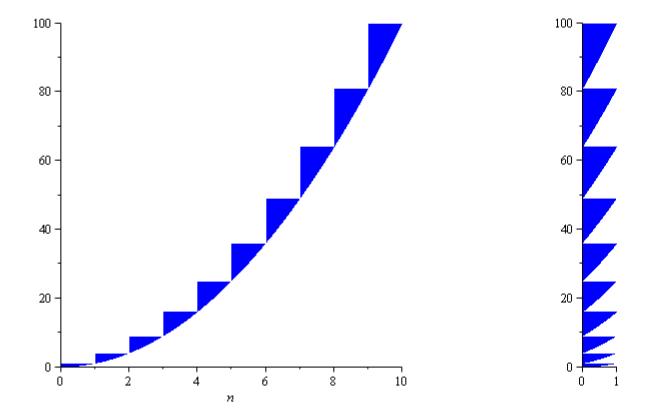
• Demonstrating with d=3 and d=4





- How large is the error?
 - Assuming d > 1, shifting the errors, we can see that they would be:

$$\frac{n^d}{2} \le \sum_{k=0}^n k^d - \frac{n^{d+1}}{d+1} < n^d < n^{d+1}$$



- The ratio between the error and the actual value goes to 0:
 - In the limit, as $n \to \infty$, the ratio between the sum and the approximation goes to 1:

$$\lim_{n\to\infty} \frac{\frac{n^{d+1}}{d+1}}{\sum_{k=0}^{n} k^d} = 1$$

The relative error of the approximation goes to 0

• The geometric series with common ratio r:

$$\sum_{k=0}^{n} r^{k} = \frac{1 - r^{n+1}}{1 - r}$$

• If |r| < 1, then it is also true that

$$\sum_{k=0}^{\infty} r^k = \frac{1}{1-r}$$

• Elegant proof: using $1 = \frac{1-r}{1-r}$

$$\sum_{k=0}^{n} r^{k} = \frac{(1-r)\sum_{k=0}^{n} r^{k}}{1-r}$$

$$= \frac{\sum_{k=0}^{n} r^{k} - r\sum_{k=0}^{n} r^{k}}{1-r}$$

$$= \frac{(1+r+r^{2}+\cdots+r^{n}) - (r+r^{2}+\cdots+r^{n}+r^{n+1})}{1-r}$$

$$= \frac{1-r^{n+1}}{1-r}$$

- Proof by induction:
 - The formula is correct for n = 0: $\sum_{k=0}^{0} r^k = r^0 = 1 = \frac{1 r^{0+1}}{1 r}$
 - Assume that the formula $\sum_{i=0}^{n} r^{i} = \frac{1-r^{n+1}}{1-r}$ is true for an arbitrary n; then

$$\sum_{k=0}^{n+1} r^k = r^{n+1} + \sum_{k=0}^{n} r^k = r^{n+1} + \frac{1 - r^{n+1}}{1 - r} = \frac{(1 - r)r^{n+1} + 1 - r^{n+1}}{1 - r}$$
$$= \frac{r^{n+1} - r^{n+2} + 1 - r^{n+1}}{1 - r} = \frac{1 - r^{n+2}}{1 - r} = \frac{1 - r^{(n+1)+1}}{1 - r}$$

• and therefore, by the process of mathematical induction, the statement is true for all $n \geq 0$

• A common geometric series with the ratios $r = \frac{1}{2}$ and r = 2

$$\sum_{i=0}^{n} \left(\frac{1}{2}\right)^{i} = \frac{1 - \left(\frac{1}{2}\right)^{n+1}}{1 - \frac{1}{2}} = 2 - 2^{-n} \qquad \sum_{i=0}^{\infty} \left(\frac{1}{2}\right)^{i} = 2$$

$$\sum_{k=0}^{n} 2^{k} = \frac{1 - 2^{n+1}}{1 - 2} = 2^{n+1} - 1$$

Combinations

- Given n distinct items, in how many ways can you choose k of these?
 - I.e., "In how many ways can you combine k items from n?"
 - For example, given the set $\{1, 2, 3, 4, 5\}$, I can choose three of these in any of the following ways:

$$\{1, 2, 3\}, \{1, 2, 4\}, \{1, 2, 5\}, \{1, 3, 4\}, \{1, 3, 5\}, \{1, 4, 5\}, \{2, 3, 4\}, \{2, 3, 5\}, \{2, 4, 5\}, \{3, 4, 5\}$$

The number of ways such items can be chosen is written

$$\binom{n}{k} = \frac{n!}{k! (n-k)!}$$

- is read as "n choose k"
- A recursive definition: $\binom{n}{k} = \binom{n-1}{k} + \binom{n-1}{k-1}$

Combinations

You have also seen this in expanding polynomials:

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$$

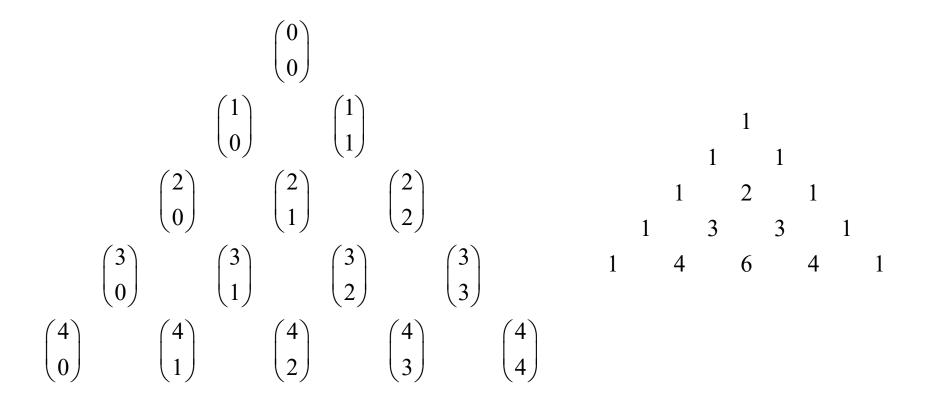
For example,

$$(x+y)^4 = \sum_{k=0}^4 {4 \choose k} x^k y^{4-k}$$

$$= {4 \choose 0} y^4 + {4 \choose 1} x y^3 + {4 \choose 2} x^2 y^2 + {4 \choose 3} x^3 y + {4 \choose 4} x^4$$

$$= y^4 + 4xy^3 + 6x^2 y^2 + 4x^3 y + x^4$$

Pascal's Triangle



Any Question?