

Growth of functions

CSE 3318 – Algorithms and Data Structures
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Math Background & book reference

L'Hopital rule:

If $\lim_{n \rightarrow \infty} f(n)$ and $\lim_{n \rightarrow \infty} g(n)$ are both 0 or $\pm\infty$
and if $\lim_{n \rightarrow \infty} \frac{f'(n)}{g'(n)}$ is a constant or $\pm\infty$

$$\text{Then } \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{f'(n)}{g'(n)}$$

- Limits
 - From the Limits cheat sheet see:
 - Properties,
 - Basic limit evaluations at $\pm\infty$ (focus on the +),
 - **Polynomials at infinity (in Evaluation techniques)**
 - [L'Hopital's rule on wikipedia](#)
- Derivatives
 - needed to Apply L'Hopital's rule
 - From the Derivatives cheat sheet see:
 - “Basic Properties and Formulas” and
 - “Common Derivatives” (especially for: polynomial, logarithmic and exponential functions)
- Logarithm properties
 - See the class cheat sheet
- Cheat sheets and other useful links are on the [Slides and Resources webpage](#).
 - Math cheat sheets and links: [Integrals](#), [Derivatives](#), [Limits](#) [Algebraic properties of equality](#), [L'Hospital's rule on wikipedia](#).
- **Book: Chapter 3.2 has a useful math review**

Book

- Read chapter 3
 - Including 3.2 which has useful math review

Motivation

Understand the formal meaning of $\Theta, \Omega, O, \omega, o$.

Be able to understand (read and calculate) Big-Oh notation. E.g.

- Alg 1 is $O(N^2)$
- Alg 2 is $\Theta(N \lg N)$
- Alg 3 is $\Omega(N)$

Function plots

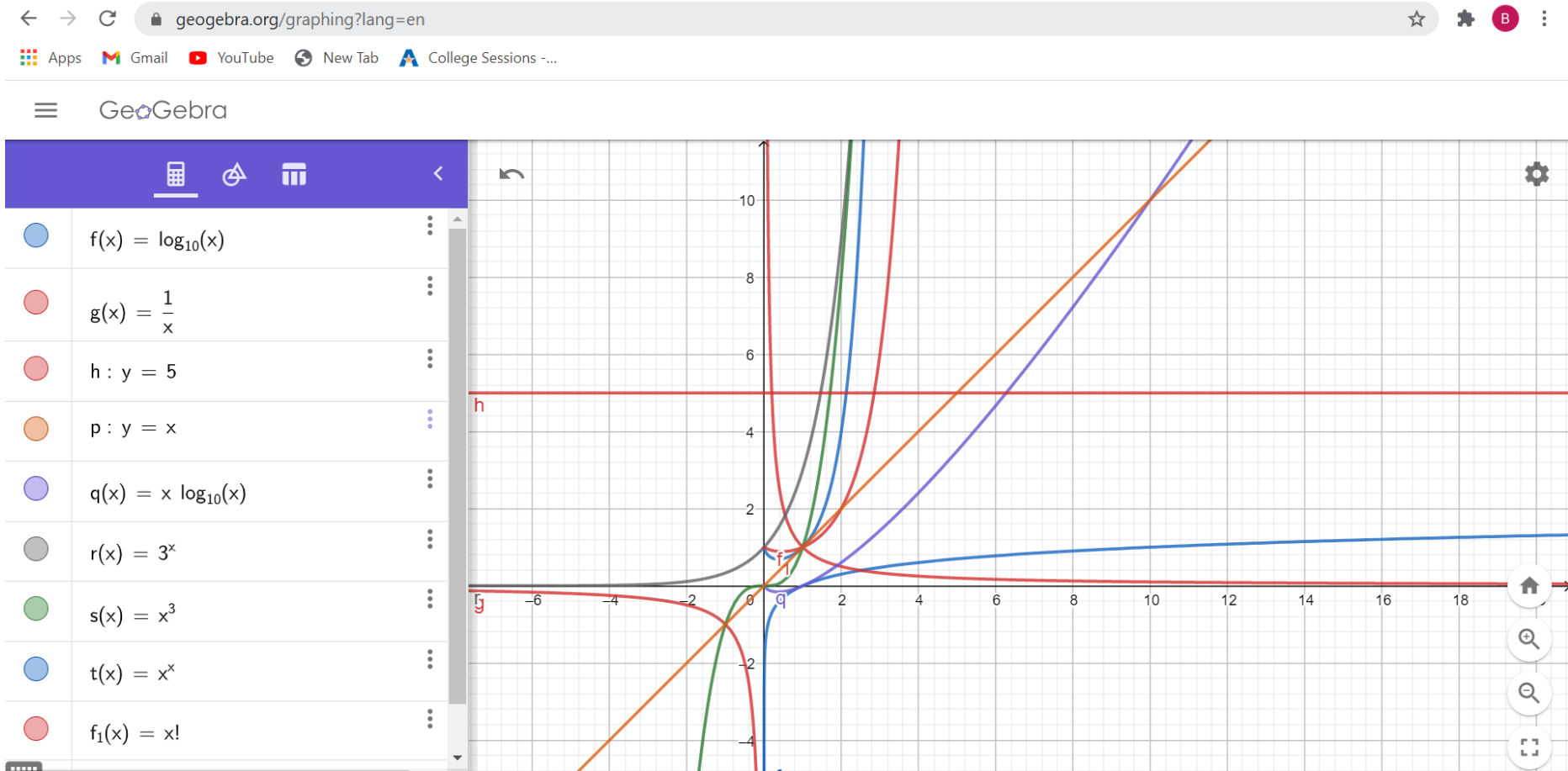


Image from Biruk Kebede

Asymptotic Bounds Notation

- Theta – is called an *asymptotic tight* bound
- Goal: we want to express lower and upper bounds as well. E.g.:
 - Selection sort will take time strictly proportional to n^2 $\in \Theta(n^2)$
 - Insertion sort will take time at most proportional n^2 $\in O(n^2)$
 - Use big-Oh for upper bounding complex functions of n .
 - Note that we can still say that the **worst case** for insertion sort is $\Theta(n^2)$.
 - Any sorting algorithm will take time at least proportional to n . $\in \Omega(n)$
- Math functions that are:
 - $\Theta(n^2)$:
 - $O(n^2)$:
 - $\Omega(n^2)$:

Abuse notation:

$f(n) = O(g(n))$

instead of:

$f(n) \in O(g(n))$

Informal definition:

$f(n)$ grows 'proportional' to $g(n)$ if:

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c \neq 0$$

(c is a non-zero constant)

= Θ tight bound

\leq O upper bound
(big-Oh – bigger)

\geq Ω lower bound

Asymptotic Bounds and Notation (CLRS)

chapter 3)

- $f(n) = \Theta(g(n)) \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c \neq 0$ (limit is a non-zero constant)
 - $g(n)$ is an **asymptotic tight** bound for $f(n)$.
 - $f(n) = \Theta(g(n)) \Leftrightarrow$ there exist positive constants c_0, c_1 and n_0 such that $c_0 g(n) \leq f(n) \leq c_1 g(n) \quad \forall n \geq n_0$.
- $f(n) = O(g(n)) \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$ or c (limit is a constant)
 - $g(n)$ is an **asymptotic upper** bound for $f(n)$.
 - $f(n) = O(g(n)) \Leftrightarrow$ there exist positive constants c_0 and n_0 such that $f(n) \leq c_0 g(n)$, for all $n \geq n_0$.
- $f(n) = \Omega(g(n)) \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$ or c
 - $g(n)$ is an **asymptotic lower** bound for $f(n)$.
 - $f(n) = \Omega(g(n)) \Leftrightarrow$ there exist positive constants c_0 and n_0 such that $c_0 g(n) \leq f(n)$ for all $n \geq n_0$.
- Typically, $f(n)$ is the running time of an algorithm. This can be a complicated function.
- We try to find a $g(n)$ that is **simple** (e.g. n^2), and such that $f(n) = O(g(n))$. E.g.
 $7n^2 + 10n + 26 = O(n^2)$

Asymptotic Bounds and Notation (CLRS)

chapter 3)

“little-oh”: **o**

- **Theorem:** if $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \mathbf{0}$, then $f(n) \in o(g(n))$
- $f(n)$ is **$o(g(n))$** if **for any** positive constant c_0 , there exists n_0 s.t.: **$f(n) < c_0 g(n)$** for all $n \geq n_0$.
- $g(N)$ is an **asymptotic upper** bound for $f(N)$ (**but NOT tight**).
- E.g.: $n = o(n^2)$, $n = o(n \lg n)$, $n^2 = o(n^4)$,...

“little-omega”: **ω**

- **Theorem:** if $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$, then $f(n) \in \omega(g(n))$
- $f(N)$ is **$\omega(g(n))$** if **for any** positive constant c_0 , there exists n_0 s. t.: **$c_0 g(n) < f(n)$** for all $n \geq n_0$.
- $g(n)$ is an **asymptotic lower** bound for $f(n)$ (**but NOT tight**).
- E.g.: $n^2 = \omega(n)$, $n \lg n = \omega(n)$, $n^3 = \omega(n^2)$,...

Theta vs Big-Oh

- The Theta notation is more strict than the Big-Oh notation:
 - $n^2 = O(n^3)$ is true
 - $n^2 = \Theta(n^3)$ is false

Properties of O , Ω and Θ

1. $f(n) = O(g(n)) \Rightarrow g(n) = \Omega(f(n))$
2. $f(n) = \Omega(g(n)) \Rightarrow g(n) = O(f(n))$
3. $f(n) = \Theta(g(n)) \Rightarrow g(n) = \Theta(f(n))$
4. *If $f(n) = O(g(n))$ and $f(n) = \Omega(g(n)) \Rightarrow f(n) = \Theta(g(n))$*
5. *If $f(n) = \Theta(g(n)) \Rightarrow f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$*

Transitivity (proved in future slides):

6. *If $f(n) = O(g(n))$ and $g(n) = O(h(n))$, then $f(n) = O(h(n))$.*
7. *If $f(n) = \Omega(g(n))$ and $g(n) = \Omega(h(n))$, then $f(n) = \Omega(h(n))$.*

Simplifying Big-Oh Notation

- Let $f(n) = 35n^2 + 41n + \lg(n) + 1532$.
- We say that $f(n) = O(n^2)$.
- Also correct, but too detailed (do not use them):
 - $f(n) = O(n^2+n)$
 - $f(n) = O(35n^2 + 41n + \lg(n) + 1532)$.

Polynomial functions

- If $f(n)$ is a polynomial function, then it is Θ of the dominant term.
- E.g. $f(n) = 15n^3 + 7n^2 + 3n + 20$,
find $g(n)$ s.t. $f(n) = \Theta(g(n))$:
 - find the dominant term: $15n^3$
 - Ignore the constant, left with: n^3
 - $\Rightarrow g(n) = n^3$
 - $\Rightarrow f(n) = \Theta(n^3)$

You cannot use the dominant term method if $f(n)$ is a summation that has a number of terms that depends on n .

E.g.: $f(n) = n^2 + (n-1)^2 + \dots + 2^2 + 1$

See Summations for techniques for solving these.

Big-Oh Hierarchy

- $1 = O(\lg(n))$
- $\lg(n) = O(n)$
- $n = O(n^2)$
- If $0 \leq c \leq d$, then $n^c = O(n^d)$.
 - Higher-order polynomials always get larger than lower-order polynomials, eventually.
- For any d , if $c > 1$, $n^d = o(c^n)$. (e.g. $n^{100} = o(2^n)$)
 - Exponential functions always get larger than polynomial functions, eventually.
- You can use these facts in your assignments.
- You can apply transitivity to derive other facts, e.g., that $\lg(n) = O(n^2)$.

$1/n, 1, \lg n, n^\epsilon, \sqrt{n}, n, n \lg n, n^2, n^3, n^d, c^n, n!, n^n$ (where $0 < \epsilon < 0.5$)

$n!$

- Compare the following functions (in terms of o, ω)

$$n!, \quad 2^n, \quad n^n$$

- We can upper and lower bound $n!$

$$n! = o(n^n) \quad n^n \text{ is a strictly upper bound for } n!$$

$$n! = \omega(2^n) \quad 2^n \text{ is a strictly lower bound for } n!$$

- Extra material:

- We have a Θ bound on $\lg(n!)$:

$$\lg(n!) = \lg 1 + \lg 2 + \lg 3 + \dots + \lg(n-1) + \lg n = \Theta(n \lg n)$$

- But not on $n!$: $n! \neq \Theta(n^n)$
- To understand why, see this case, consider:

$$\lg(n^2) = \Theta(\lg(n)) \quad \text{but} \\ n^2 \neq \Theta(n)$$

Useful logarithm properties

- $c^{\lg(n)} = n^{\lg(c)}$
 - Proof: apply lg on both sides and you get two equal terms:

$$\begin{aligned}\lg(c^{\lg(n)}) &= \lg(n^{\lg(c)}) \quad \Rightarrow \\ \lg(n) * \lg(c) &= \lg(n) * \lg(c)\end{aligned}$$

- This equality helps identify “false exponentials”. E.g. $3^{\lg(n)}$ *may look like an exponential growth, but is really polynomial: $n^{\lg(3)}$.*
- Can we also say that $c^n = n^c$?
 - NO!

Proofs using the c definition: O

- Let $f(n) = 35n^2 + 41n + \lg(n) + 1532$.

Show (using the definition) that $f(n) = O(n^2)$.

- Proof:

Want to find n_0 and c_0 s.t., for all $n \geq n_0$: $f(n) \leq c_0 n^2$.

Upperbound each term in the $f(n)$ expression.

Version 1:

- Pick c_0 large enough to cover the coefficients of all the terms:

$$\begin{aligned} f(n) &= 35n^2 + 41n + \lg(n) + 1532 \leq \\ &\leq 35n^2 + 41n^2 + n^2 + 1532n^2 = 1609n^2 \\ &\Rightarrow \text{use } c_0 = 1609 \text{ and } n_0 = 1 \end{aligned}$$

Version 2:

- Upper bound each term by n^2 for large n (e.g. $n \leq 1532$)

$$\begin{aligned} f(n) &= 35n^2 + 41n + \lg(n) + 1532 \\ &\leq 35n^2 + n^2 + n^2 + n^2 = 38n^2 \quad \text{for all } n \geq 1536 \\ &\Rightarrow f(n) = 35n^2 + 41n + \lg(n) + 1532 \leq 38n^2, \text{ for all } n \geq 1536 \quad (\text{here } c_0 = 38, n_0 = 1536) \end{aligned}$$

Proofs using the c definition: Ω , Θ

- Let $f(n) = 35n^2 + 41n + \lg(n) + 1532$.

- Proof that $f(n) = \Omega(n^2)$:

Want to find n_1 and c_1 s.t., for all $n \geq n_1$: $f(n) \geq c_1 n^2$.

– Use: $c_1 = 1, n_1 = 1$

$$f(n) = 35n^2 + 41n + \lg(n) + 1532 \geq n^2, \text{ for all } n \geq 1$$

- Proof that $f(n) = \Theta(n^2)$:

Version 1: use *property 4, page 10*:

We have proved $f(n) = O(n^2)$ and $f(n) = \Omega(n^2)$, therefore $f(n) = \Theta(n^2)$ holds

Version 2: We found $c_0 = 38, n_0 = 1536$ and $c_1 = 1, n_1 = 1$ s.t.:

$$f(n) = 35n^2 + 41n + \lg(n) + 1532 \leq 38n^2, \text{ for all } n \geq 1536$$

$$f(n) = 35n^2 + 41n + \lg(n) + 1532 \geq n^2, \text{ for all } n \geq 1$$

$$\Rightarrow n^2 \leq f(n) \leq 38n^2, \text{ for all } n \geq 1536 \Rightarrow f(n) = \Theta(n^2)$$

Using Limits: Example 1

- Suppose that we are given this running time:
 $f(n) = 35n^2 + 41n + \lg(n) + 1532$.
- Use the limits theorem to show that $f(n) = O(n^2)$.

Summary

- Definitions
- Properties: transitivity, reflexivity, ...
- Using limits
- Big-Oh hierarchy
- Example problems
- Asymptotic notation for two parameters
- $a^{\log_b(n)} = n^{\log_b(a)}$ ($a^n \neq n^a$) (note \log_b in the exponent)

EXTRA
NOT REQUIRED

Example Problem 1

- Is $n = O(\sin(n) n^2)$?
- Answer:

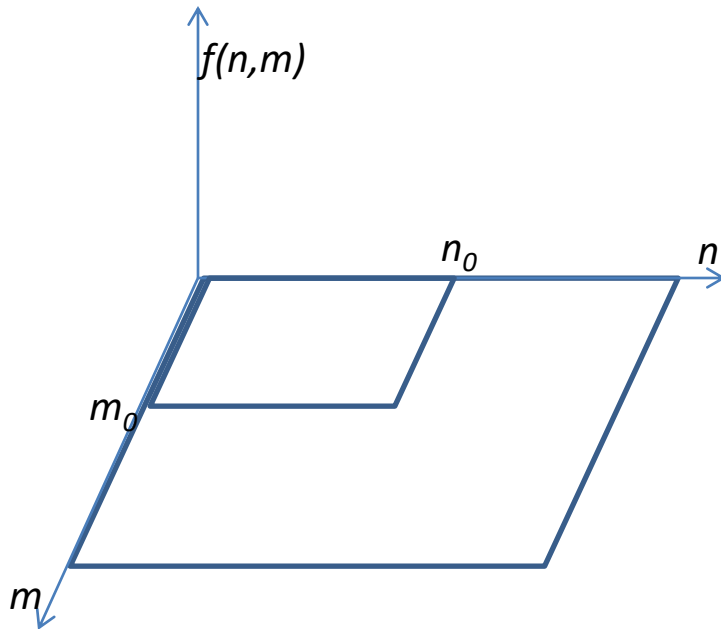
Example Problem 2

- Show that $\max(f(n), g(n))$ is $\Theta(f(n) + g(n))$
 - Show O :
 - Show Ω :

Asymptotic notation for two parameters (CLRS)

$f(n,m)$ is $O(g(n,m))$ if there exist constants c_0 , n_0 and m_0 such that:

$$f(n,m) \leq c_0 g(n,m) \text{ for all pairs } (n,m) \text{ s.t.} \\ \text{either } n \geq n_0 \text{ or } m \geq m_0$$



Using Limits

- if $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c$ is a **non-zero** constant, then $f(n) = \underline{\hspace{1cm}}$ ($g(n)$).

- In this definition, both zero and infinity are excluded.
- In this case we can also say that $g(n) = \Theta(f(n))$.

This can easily be proved using the limit or the reflexivity property of Θ .

- if $\lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = c$ is a constant, then $f(n) = \underline{\hspace{1cm}}$ ($g(n)$).

- "constant" includes zero, but cannot be infinity.

- if $\lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = \infty$ then $f(n) = \underline{\hspace{1cm}}$ ($g(n)$).

- $g(n)$ grows much faster than $f(n)$

- if $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c$ is a constant, then $f(n) = \underline{\hspace{1cm}}$ ($g(n)$).

- "Constant" includes zero, but cannot be infinity.

Using Limits

- if $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c$ is a **non-zero** constant, then $f(n) = \Theta(g(n))$.
 - In this definition, both zero and infinity are excluded.
 - In this case we can also say that $g(n) = \Theta(f(n))$.

This can easily be proved using the limit or the reflexivity property of Θ .

- if $\lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = c$ is a constant, then $f(n) = \Omega(g(n))$.
 - "constant" includes zero, but cannot be infinity.
- if $\lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = \infty$ then $f(n) = o(g(n))$.
 - $g(n)$ grows much faster than $f(n)$
- if $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c$ is a constant, then $f(n) = O(g(n))$.
 - "Constant" includes zero, but cannot be infinity.

Asymptotic Notation in Expressions (if needed)

In the recurrence formulas and proofs, you may see these notations (see CLRS, page 49):

- $f(n) = 2n^2 + \Theta(n)$
 - *There is a function $h(n)$ in $\Theta(n)$ s.t. $f(n) = 2n^2 + h(n)$*
- $2n^2 + \Theta(n) = \Theta(n^2)$.
 - *For any function $h(n)$ in $\Theta(n)$, there is a function $g(n)$ in $\Theta(n^2)$ s.t. $2n^2 + h(n) = g(n)$.*
 - *For any function $h(n)$ in $\Theta(n)$, $2n^2 + h(n)$ is in $\Theta(n^2)$.*

Big-Oh Transitivity - Proof

- If $f(n) = O(g(n))$ and $g(n) = O(h(n))$, then $f(n) = O(h(n))$.

Proof:

We want to find c_3 and n_3 s. t. $f(n) \leq c_3 h(n)$, for all $n \geq n_3$.

We know:

$f(n) = O(g(n)) \Rightarrow$ there exist c_1, n_1 , s.t. $f(n) \leq c_1 g(n)$, for all $n \geq n_1$

$g(n) = O(h(n)) \Rightarrow$ there exist c_2, n_2 , s.t. $g(n) \leq c_2 h(n)$, for all $n \geq n_2$

$\Rightarrow f(n) \overset{n \geq n_1}{\leq} c_1 g(n) \overset{n \geq n_2}{\leq} c_1 c_2 h(n)$, for all $n \geq \max(n_1, n_2)$

\Rightarrow Use: $c = c_1 * c_2$, and $n \geq \max(n_1, n_2)$

Extra

Using Substitutions

- If $\lim_{x \rightarrow \infty} h(x) = \infty$, and $h(x)$ is monotonically increasing then:

$$f(\textcolor{red}{x}) = O(g(\textcolor{red}{x})) \Rightarrow f(\textcolor{red}{h(x)}) = O(g(\textcolor{red}{h(x)})).$$

(This can be proved)

- How do we use that?
- For example, prove that:

$$(\lg n)^{10} = O(n)$$

$$(\textit{for} : n^2 (\lg n)^{10} = O(n^3))$$

Proof: Use substitution: $h(n) = \lg(n)$

and: $y^{10} = O(2^y)$

($y = h(n)$)