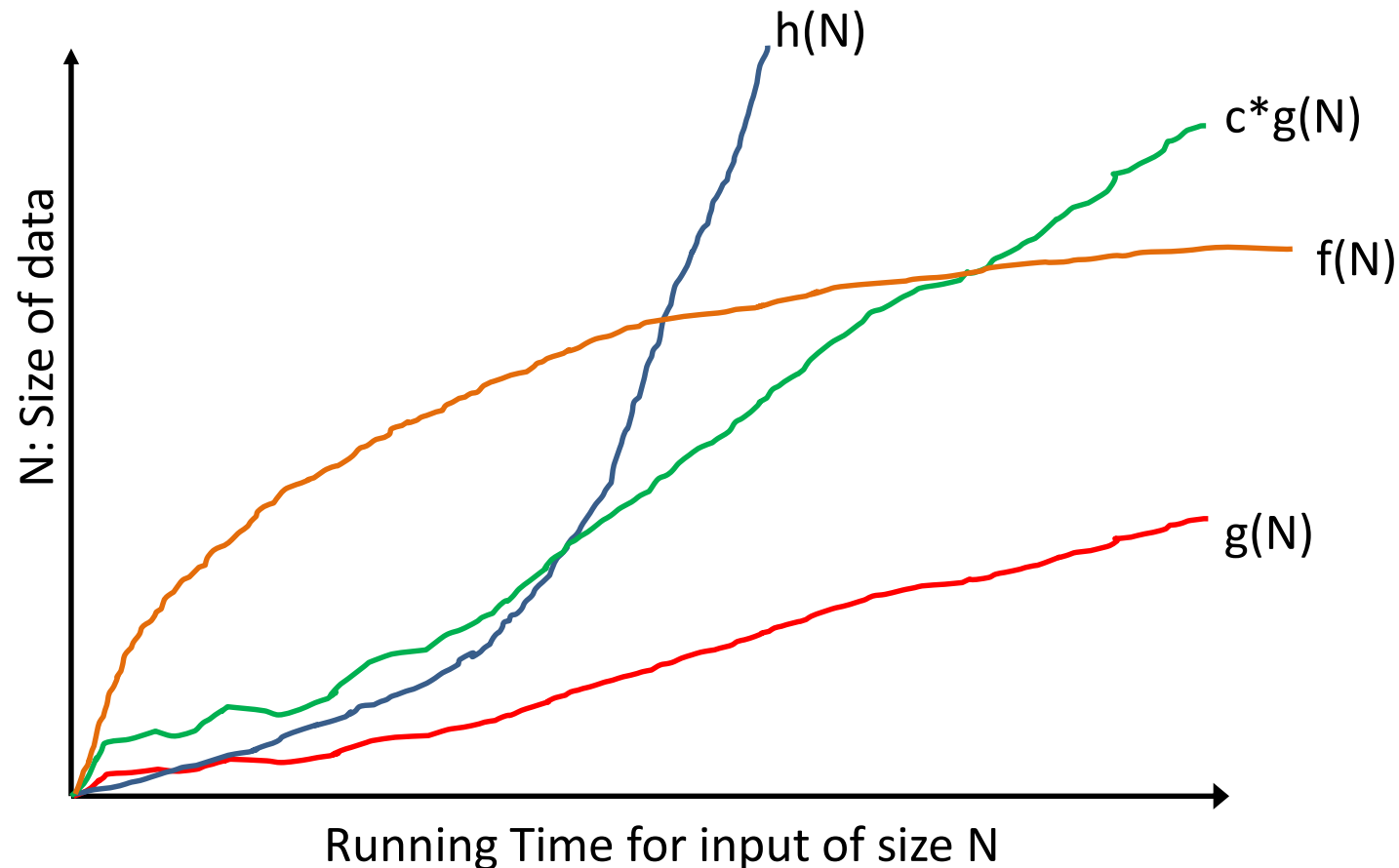


Analysis of Algorithms: Methods and Examples

CS 350 – Algorithms and Complexity
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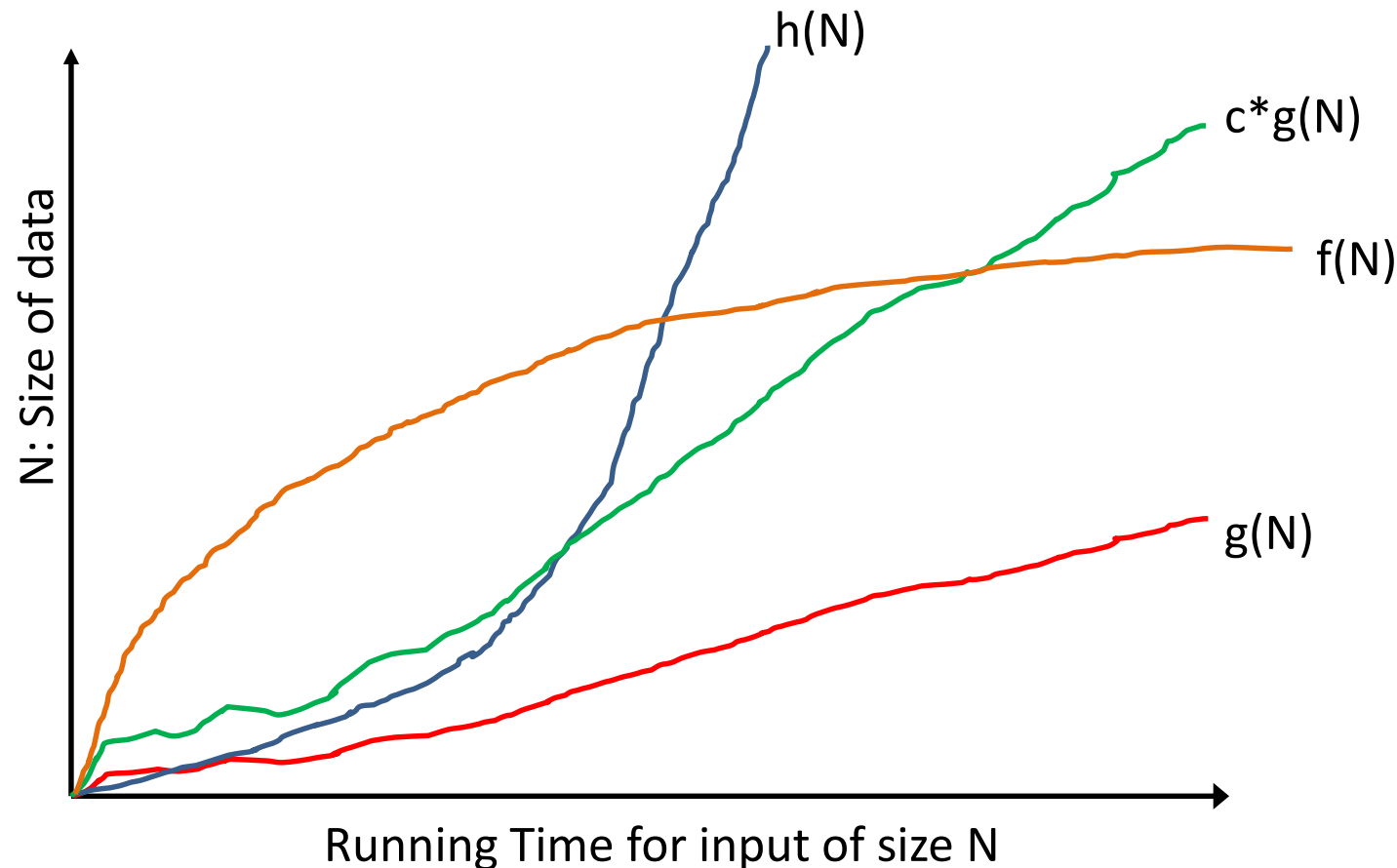
Why Asymptotic Behavior Matters

- Asymptotic behavior: The behavior of a function as the input approaches infinity.



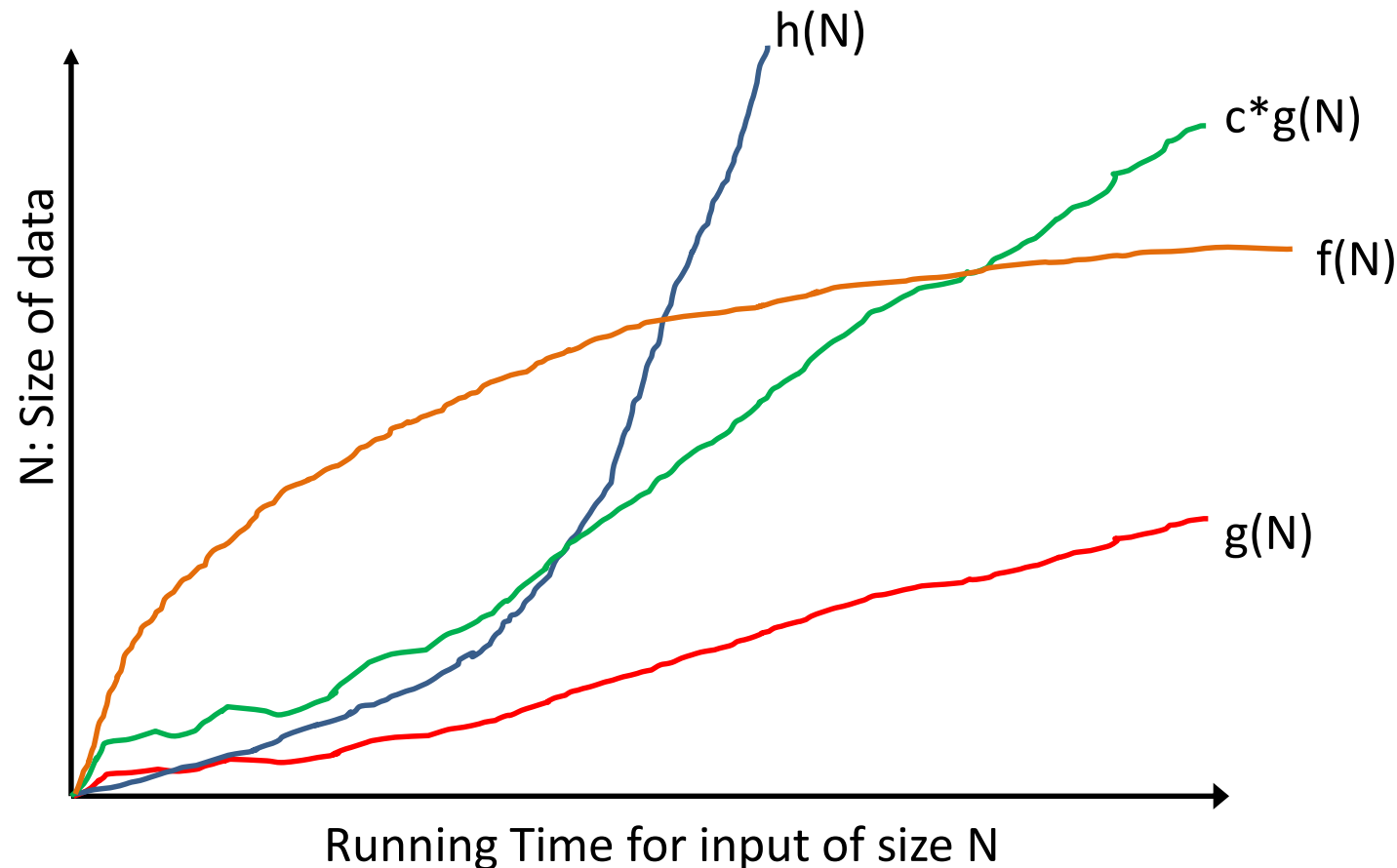
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- Which of these functions is smallest asymptotically?



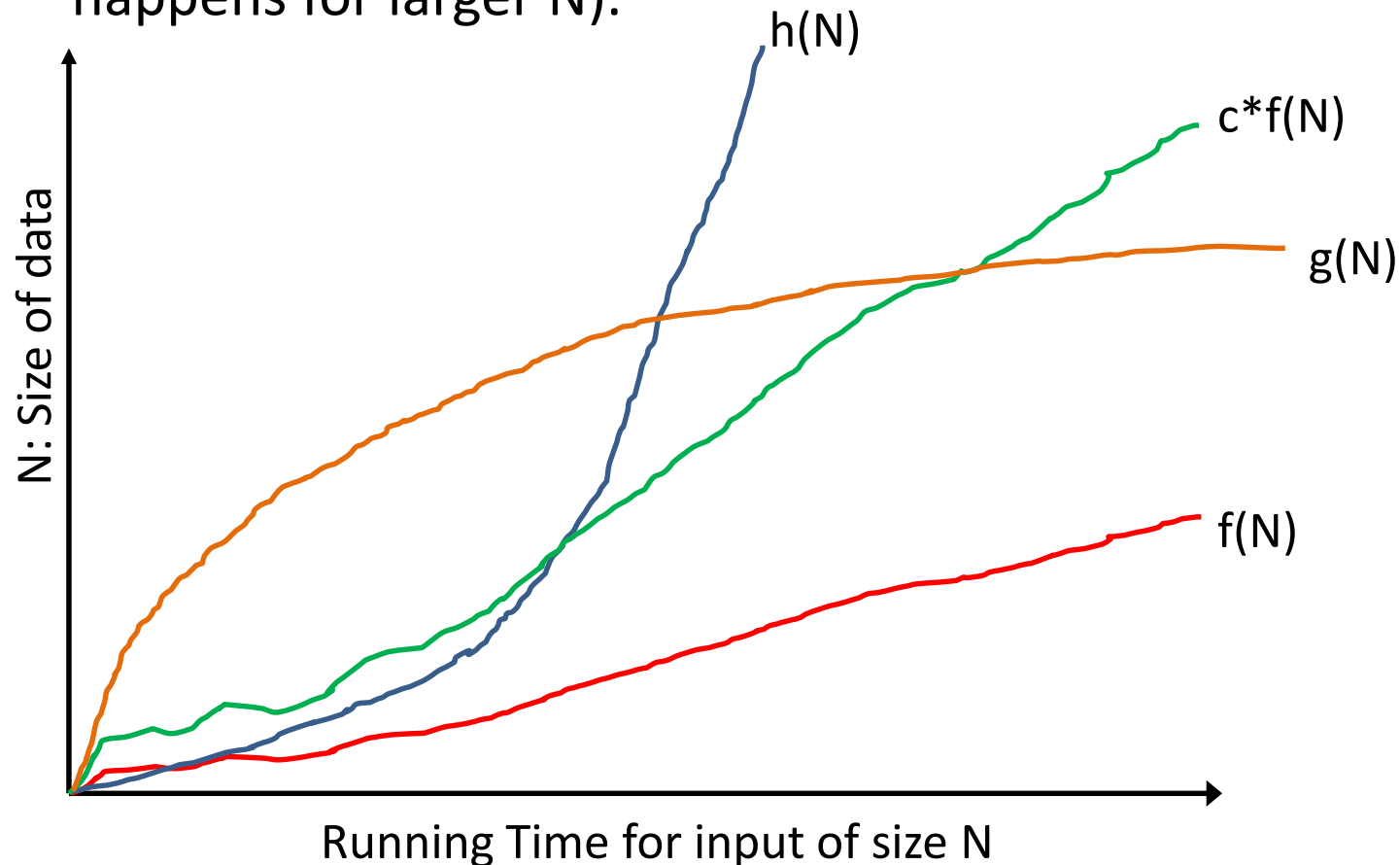
Why Asymptotic Behavior Matters

- Which of these functions is smallest asymptotically?
 - $f(N)$ seems to grow very slowly after a while.



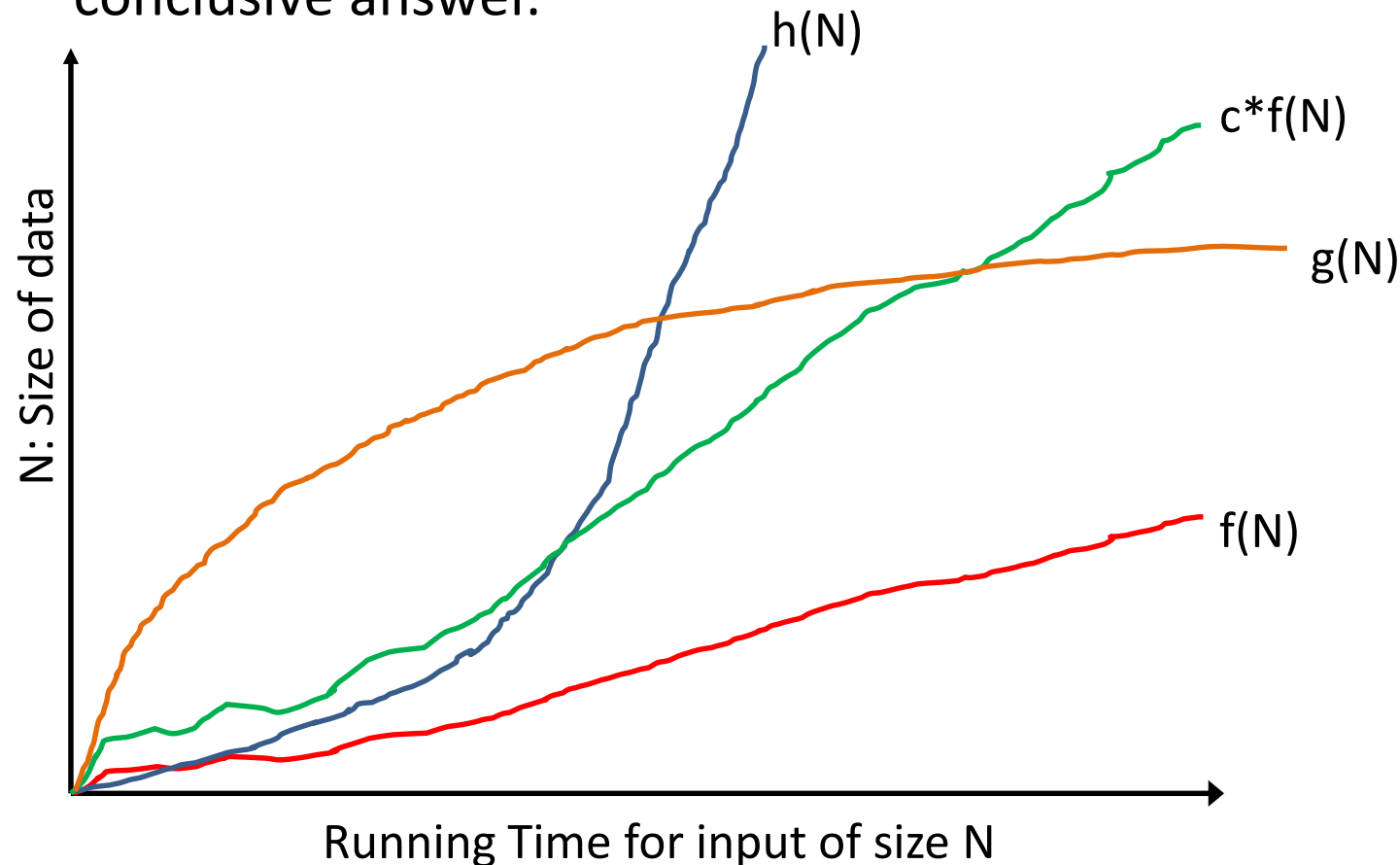
Why Asymptotic Behavior Matters

- Which of these functions is smallest asymptotically?
 - However, the picture is not conclusive (need to see what happens for larger N).



Why Asymptotic Behavior Matters

- Which of these functions is smallest asymptotically?
 - Proving that $f(N) = O(g(N))$ would provide a conclusive answer.



Using Limits for Comparing Order of Growth

- $\lim_{n \rightarrow \infty} \frac{t(n)}{g(n)} = 0$, implies that $t(n)$ has a smaller order of growth than $g(n)$
- $\lim_{n \rightarrow \infty} \frac{t(n)}{g(n)} = c, c > 0$, implies that $t(n)$ has the same order of growth as $g(n)$
- $\lim_{n \rightarrow \infty} \frac{t(n)}{g(n)} = \infty$, implies that $t(n)$ has a larger order of growth than $g(n)$
- Note that the first two cases mean that: $t(n) \in O(g(n))$
- Note that the last two cases mean that: $t(n) \in \Omega(g(n))$
- Note that the second case means that: $t(n) \in \Theta(g(n))$

Using Limits - Comments

- The previous formulas relating limits to big-Oh notation show once again that big-Oh notation ignores:
 - constants
 - behavior for small values of N .
- How do we see that?

Using Limits - Comments

- The previous formulas relating limits to big-Oh notation show once again that big-Oh notation ignores:
 - constants
 - behavior for small values of N .
- How do we see that?
 - In the previous formulas, it is sufficient that the limit is equal to a constant. **The value of the constant does not matter.**
 - In the previous formulas, only **the limit at infinity** matters. This means that we can ignore behavior up to any finite value, if we need to.

Using Limits: An Example

- Show that $\frac{n^5 + 3n^4 + 2n^3 + n^2 + n + 12}{5n^3 + n + 3} = \Theta(???)$.

Using Limits: An Example

- Show that $\frac{n^5+3n^4+2n^3+n^2+n+12}{5n^3+n+3} = \Theta(n^2)$.

- Let $f(n) = \frac{n^5+3n^4+2n^3+n^2+n+12}{5n^3+n+3}$

- Let $g(n) = n^2$.

$$\begin{aligned}\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} &= \lim_{n \rightarrow \infty} \left(\frac{n^5 + 3n^4 + 2n^3 + n^2 + n + 12}{5n^3 + n + 3} \cdot \frac{1}{n^2} \right) \\ &= \lim_{n \rightarrow \infty} \left(\frac{n^5 + 3n^4 + 2n^3 + n^2 + n + 12}{5n^5 + n^3 + 3n^2} \right) = \frac{1}{5}\end{aligned}$$

Using Limits: An Example

- Show that $\frac{n^5+3n^4+2n^3+n^2+n+12}{5n^3+n+3} = \Theta(n^2)$.
- Let $f(n) = \frac{n^5+3n^4+2n^3+n^2+n+12}{5n^3+n+3}$
- Let $g(n) = n^2$.
- In the previous slide, we showed that $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \frac{1}{5}$
- Therefore, $f(n) = \Theta(g(n))$.

Big-Oh Hierarchy

- $1 = O(\log(N))$
- $\log(N) = O(N)$
- $N = O(N^2)$
- If $c \geq d \geq 0$, then $N^d = O(N^c)$.

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- For any d , if $c > 1$, $N^d = O(c^N)$.
 - 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 $\log(N) = O(N^2)$.

Using Substitutions

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

$$f(x) = O(g(x)) \Rightarrow f(h(x)) = O(g(h(x))).$$

- How do we use that?
- For example, prove that $\log(\sqrt{N}) = O(\sqrt{N})$.

Using Substitutions

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

$$g(x) = O(f(x)) \Rightarrow g(h(x)) = O(f(h(x))).$$

- How do we use that?
- For example, prove that $\log(\sqrt{N}) = O(\sqrt{N})$.
- Use $h(x) = \sqrt{N}$. We get:

$$\log(N) = O(N) \Rightarrow \log(\sqrt{N}) = O(\sqrt{N})$$

Big-Oh Notation: Example Problem

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

Big-Oh Notation: Example Problem

- Is $N = O(\sin(N) N^2)$?
- Answer: no!
- Why? $\sin(N)$ fluctuates forever between -1 and 1.
- As a result, $\sin(N) N^2$ fluctuates forever between negative and positive values.
- Therefore, for every possible $c_0 > 0$ and N_0 , we can always find an $N > N_0$ such that:

$$N > c_0 \sin(N) N^2$$