

**Module #6:
Orders of Growth**

Rosen 5th ed., §2.2
~22 slides, ~1 lecture

Orders of Growth (§1.8)

- For functions over numbers, we often need to know a rough measure of *how fast a function grows*.
- If $f(x)$ is *faster growing* than $g(x)$, then $f(x)$ always eventually becomes larger than $g(x)$ *in the limit* (for large enough values of x).
- Useful in engineering for showing that one design *scales* better or worse than another.

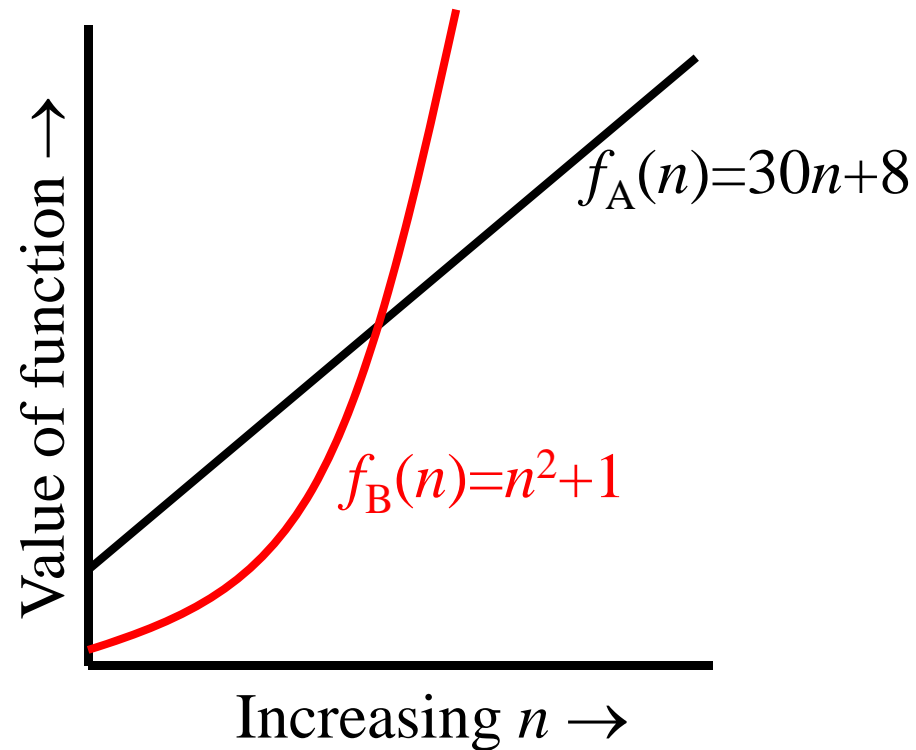
Orders of Growth - Motivation

- Suppose you are designing a web site to process user data (*e.g.*, financial records).
- Suppose database program A takes $f_A(n)=30n+8$ microseconds to process any n records, while program B takes $f_B(n)=n^2+1$ microseconds to process the n records.
- Which program do you choose, knowing you'll want to support millions of users?



Visualizing Orders of Growth

- On a graph, as you go to the right, a faster growing function eventually becomes larger...



Concept of order of growth

- We say $f_A(n)=30n+8$ is *order n* , or $O(n)$. It is, at most, roughly *proportional* to n .
- $f_B(n)=n^2+1$ is *order n^2* , or $O(n^2)$. It is roughly proportional to n^2 .
- Any $O(n^2)$ function is faster-growing than any $O(n)$ function.
- For large numbers of user records, the $O(n^2)$ function will always take more time.

Definition: $O(g)$, at most order g

Let g be any function $\mathbf{R} \rightarrow \mathbf{R}$.

- Define “*at most order g* ”, written $O(g)$, to be: $\{f: \mathbf{R} \rightarrow \mathbf{R} \mid \exists c, k: \forall x > k: f(x) \leq cg(x)\}$.
 - “Beyond some point k , function f is at most a constant c times g (*i.e.*, proportional to g).”
- “ f is *at most order g* ”, or “ f is $O(g)$ ”, or “ $f = O(g)$ ” all just mean that $f \in O(g)$.
- Sometimes the phrase “at most” is omitted.

Points about the definition

- Note that f is $O(g)$ so long as *any* values of c and k exist that satisfy the definition.
- But: The particular c, k , values that make the statement true are *not* unique: **Any larger value of c and/or k will also work.**
- You are **not** required to find the smallest c and k values that work. (Indeed, in some cases, there may be no smallest values!)

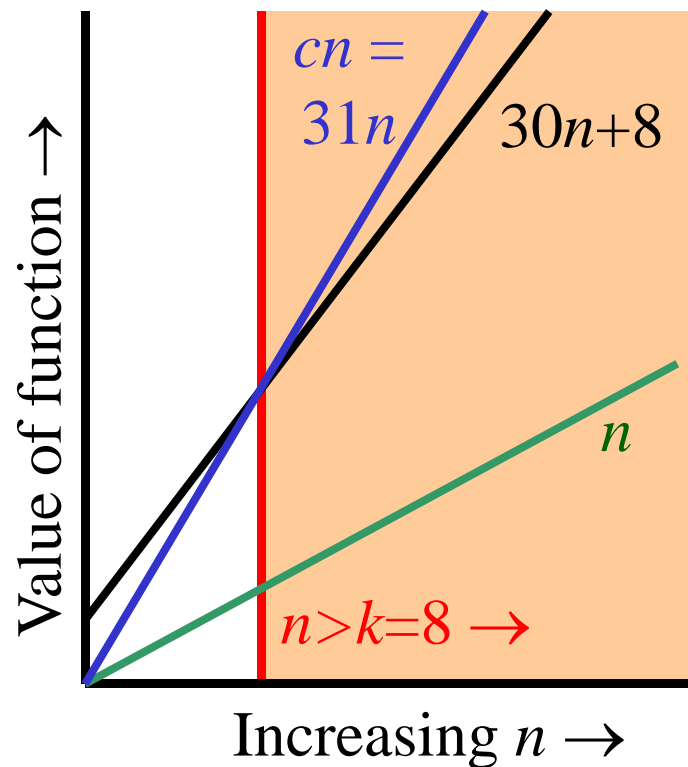
However, you should **prove** that the values you choose do work.

“Big-O” Proof Examples

- Show that $30n+8$ is $O(n)$.
 - Show $\exists c,k: \forall n>k: 30n+8 \leq cn$.
 - Let $c=31, k=8$. Assume $n>k=8$. Then $cn = 31n = 30n + n > 30n+8$, so $30n+8 < cn$.
- Show that n^2+1 is $O(n^2)$.
 - Show $\exists c,k: \forall n>k: n^2+1 \leq cn^2$.
 - Let $c=2, k=1$. Assume $n>1$. Then $cn^2 = 2n^2 = n^2+n^2 > n^2+1$, or $n^2+1 < cn^2$.

Big-O example, graphically

- Note $30n+8$ isn't less than n *anywhere* ($n>0$).
- It isn't even less than $31n$ *everywhere*.
- But it *is* less than $31n$ everywhere to the right of $n=8$.



$$30n+8 \in O(n)$$

Useful Facts about Big O

- Big O, as a relation, is transitive:
 $f \in O(g) \wedge g \in O(h) \rightarrow f \in O(h)$
- O with constant multiples, roots, and logs...
 $\forall f$ (in $\omega(1)$) & constants $a, b \in \mathbf{R}$, with $b \geq 0$,
 af , f^{1-b} , and $(\log_b f)^a$ are all $O(f)$.
- Sums of functions:
If $g \in O(f)$ and $h \in O(f)$, then $g+h \in O(f)$.

More Big-O facts

- $\forall c > 0, O(cf) = O(f+c) = O(f-c) = O(f)$
- $f_1 \in O(g_1) \wedge f_2 \in O(g_2) \rightarrow$
 - $f_1 f_2 \in O(g_1 g_2)$
 - $f_1 + f_2 \in O(g_1 + g_2)$
 - $= O(\max(g_1, g_2))$
 - $= O(g_1)$ if $g_2 \in O(g_1)$ (Very useful!)

Orders of Growth (§1.8) - So Far

- For any $g:\mathbf{R}\rightarrow\mathbf{R}$, “*at most order g*”,
 $O(g) \equiv \{f:\mathbf{R}\rightarrow\mathbf{R} \mid \exists c,k \forall x>k |f(x)| \leq |cg(x)|\}$.
 - Often, one deals only with positive functions and can ignore absolute value symbols.
- “ $f \in O(g)$ ” often written “ f is $O(g)$ ” or “ $f = O(g)$ ”.
 - The latter form is an instance of a more general convention...

Order-of-Growth Expressions

- “ $O(f)$ ” when used as a term in an arithmetic expression means: “some function f such that $f \in O(f)$ ”.
- *E.g.:* “ $x^2 + O(x)$ ” means “ x^2 plus some function that is $O(x)$ ”.
- Formally, you can think of any such expression as denoting a set of functions:
“ $x^2 + O(x)$ ” $:= \{g \mid \exists f \in O(x): g(x) = x^2 + f(x)\}$

Order of Growth Equations

- Suppose E_1 and E_2 are order-of-growth expressions corresponding to the sets of functions S and T , respectively.
- Then the “equation” $E_1=E_2$ really means
$$\forall f \in S, \exists g \in T : f=g$$
or simply $S \subseteq T$.
- Example: $x^2 + O(x) = O(x^2)$ means
$$\forall f \in O(x) : \exists g \in O(x^2) : x^2 + f(x) = g(x)$$

Useful Facts about Big O

- $\forall f, g$ & constants $a, b \in \mathbf{R}$, with $b \geq 0$,
 - $af = O(f)$; (e.g. $3x^2 = O(x^2)$)
 - $f + O(f) = O(f)$; (e.g. $x^2 + x = O(x^2)$)
- Also, if $f = \Omega(1)$ (at least order 1), then:
 - $|f|^{1-b} = O(f)$; (e.g. $x^{-1} = O(x)$)
 - $(\log_b |f|)^a = O(f)$. (e.g. $\log x = O(x)$)
 - $g = O(fg)$ (e.g. $x = O(x \log x)$)
 - $fg \neq O(g)$ (e.g. $x \log x \neq O(x)$)
 - $a = O(f)$ (e.g. $3 = O(x)$)

Definition: $\Theta(g)$, *exactly order g*

- If $f \in O(g)$ and $g \in O(f)$ then we say “*g and f are of the same order*” or “*f is (exactly) order g*” and write $f \in \Theta(g)$.
- Another equivalent definition:
$$\Theta(g) \equiv \{f: \mathbf{R} \rightarrow \mathbf{R} \mid \exists c_1 c_2 k \forall x > k: |c_1 g(x)| \leq |f(x)| \leq |c_2 g(x)| \}$$
- “Everywhere beyond some point k , $f(x)$ lies in between two multiples of $g(x)$.”

Rules for Θ

- Mostly like rules for $O(\)$, except:
- $\forall f, g > 0$ & constants $a, b \in \mathbf{R}$, with $b > 0$,
 $af \in \Theta(f)$, but \leftarrow Same as with O .
 $f \notin \Theta(fg)$ unless $g = \Theta(1)$ \leftarrow Unlike O .
 $|f|^{1-b} \notin \Theta(f)$, and \leftarrow Unlike with O .
 $(\log_b |f|)^c \notin \Theta(f)$. \leftarrow Unlike with O .
- The functions in the latter two cases we say are *strictly of lower order* than $\Theta(f)$.

Θ example

- Determine whether: $\left(\sum_{i=1}^n i \right) \in \Theta(n^2)$?
- Quick solution:

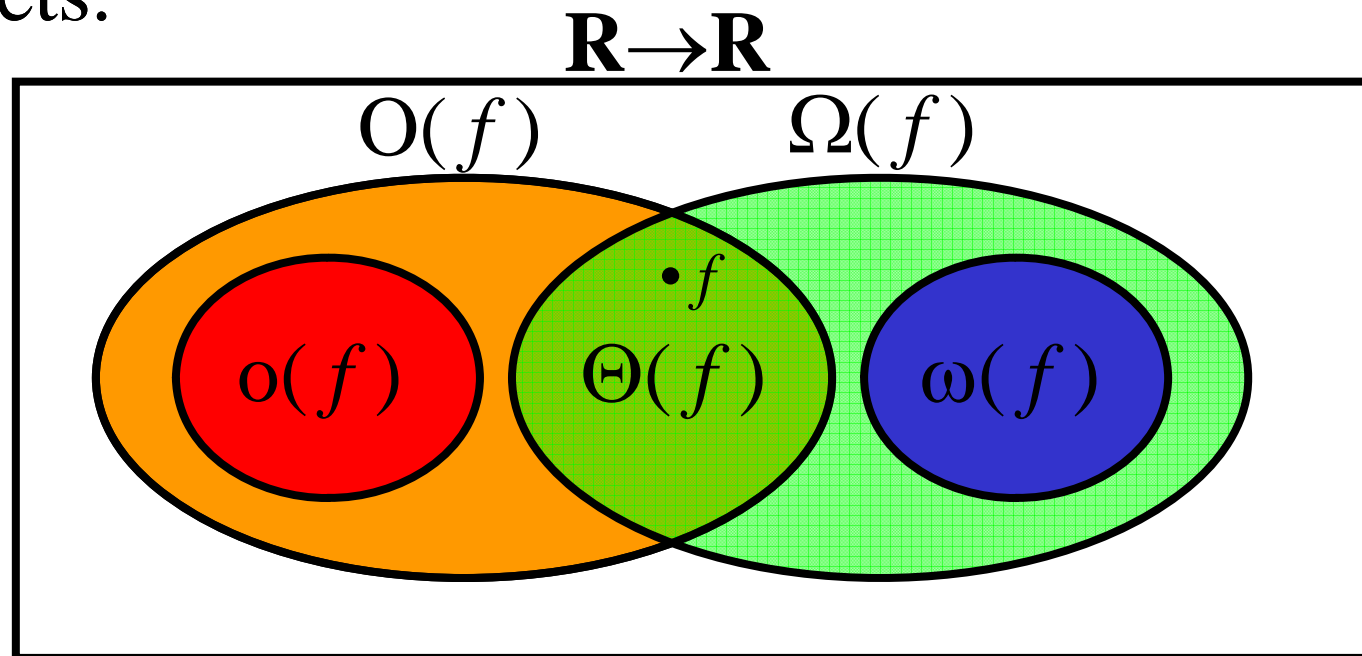
$$\begin{aligned}
 \left(\sum_{i=1}^n i \right) &= n(n-1)/2 \\
 &= n \Theta(n) / 2 \\
 &= n \Theta(n) \\
 &= \Theta(n^2)
 \end{aligned}$$

Other Order-of-Growth Relations

- $\Omega(g) = \{f \mid g \in O(f)\}$
“The functions that are *at least order g*.”
- $o(g) = \{f \mid \forall c > 0 \exists k \forall x > k : |f(x)| < |cg(x)|\}$
“The functions that are *strictly lower order than g*.” $o(g) \subset O(g) - \Theta(g)$.
- $\omega(g) = \{f \mid \forall c > 0 \exists k \forall x > k : |cg(x)| < |f(x)|\}$
“The functions that are *strictly higher order than g*.” $\omega(g) \subset \Omega(g) - \Theta(g)$.

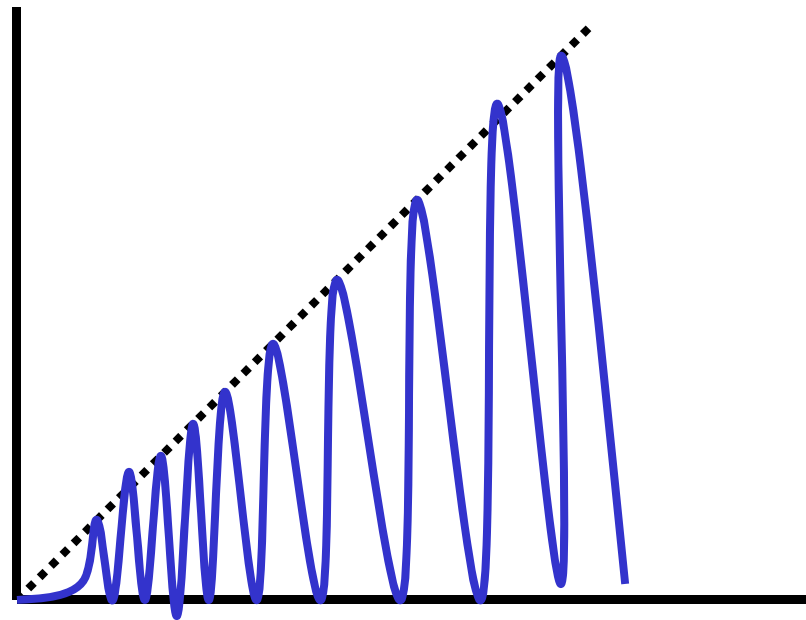
Relations Between the Relations

- Subset relations between order-of-growth sets.



Why $o(f) \subset O(x) - \Theta(x)$

- A function that is $O(x)$, but neither $o(x)$ nor $\Theta(x)$:



Strict Ordering of Functions

- Temporarily let's write $f \prec g$ to mean $f \in o(g)$,
 $f \sim g$ to mean $f \in \Theta(g)$
- Note that $f \prec g \Leftrightarrow \lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = 0$.
- Let $k > 1$. Then the following are true:
 $1 \prec \log \log n \prec \log n \sim \log_k n \prec \log^k n$
 $\prec n^{1/k} \prec n \prec n \log n \prec n^k \prec k^n \prec n! \prec n^n \dots$

Review: Orders of Growth (§1.8)

Definitions of order-of-growth sets, $\forall g:\mathbf{R}\rightarrow\mathbf{R}$

- $O(g) \equiv \{f \mid \exists c>0 \exists k \forall x>k |f(x)| < |cg(x)|\}$
- $o(g) \equiv \{f \mid \forall c>0 \exists k \forall x>k |f(x)| < |cg(x)|\}$
- $\Omega(g) \equiv \{f \mid g \in O(f)\}$
- $\omega(g) \equiv \{f \mid g \in o(f)\}$
- $\Theta(g) \equiv O(g) \cap \Omega(g)$