

# CS 3110

## Lecture 25: Amortized Analysis

Prof. Clarkson

Fall 2014

Today's music:

"Money, Money, Money" by ABBA

"Mo Money Mo Problems" by The Notorious B.I.G.

"Material Girl" by Madonna

# Review

**Current topic:** Reasoning about performance

- Efficiency
- Big Oh
- Recurrences

**Today:**

- Alternative notions of efficiency
- Amortized analysis
  - Efficiency of data abstractions, not just individual functions

# Question #1

How much of PS6 have you finished?

- A. None
- B. About 25%
- C. About 50%
- D. About 75%
- E. I'm done!!!

# Question #2

Do you think you will submit to the tournament?

- A. Yes
- B. No

# **Review: What is "efficiency"?**

**Final attempt:** An algorithm is efficient if its worst-case running time is  $O(N^d)$  for some constant  $d$ .

## Review:

# Running times of some algorithms

- $O(1)$ : access an element of an array (of length  $n$ )
- $O(\log n)$ : binary search through sorted array of length  $n$
- $O(n)$ : maximum element of list of length  $n$
- $O(n \log n)$ : mergesort a list of length  $n$
- $O(n^2)$ : bubblesort an array of length  $n$
- $O(n^3)$ : matrix multiplication of  $n$ -by- $n$  matrices
- $O(2^n)$ : enumerate all integers of bit length  $n$

...some of these are not obvious, require proof

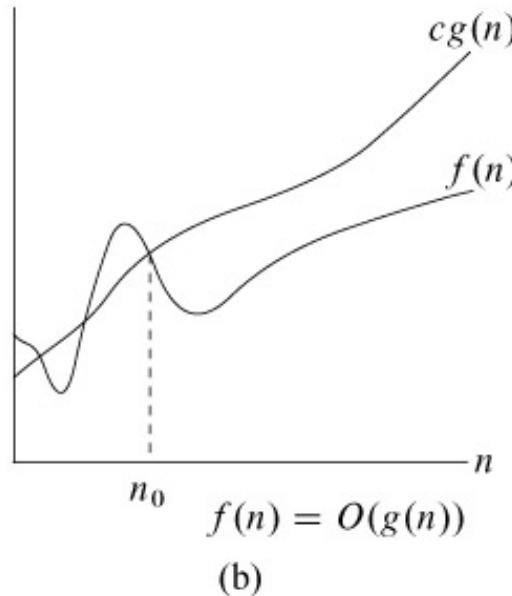
# Names of running times

- $O(1)$ : constant
- $O(\log n)$ : logarithmic
- $O(n)$ : linear
- $O(n \log n)$ : linearithmic
- $O(n^2)$ : quadratic
- $O(n^3)$ : cubic
- $O(2^n)$ : exponential

# Asymptotic bounds

## Big Oh:

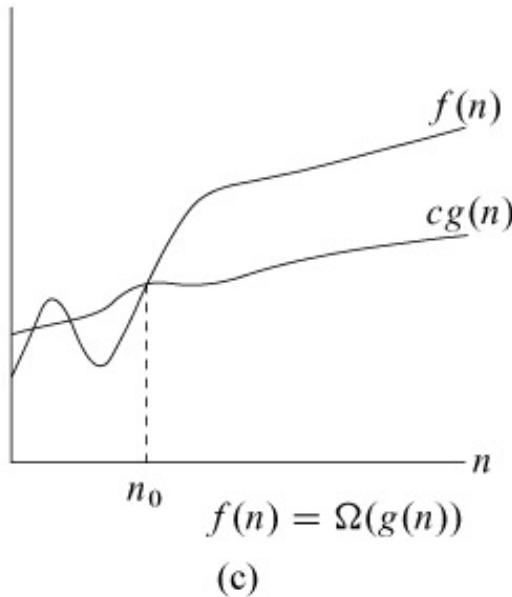
- *asymptotic upper bound*
- $O(g) = \{f \mid \text{exists } c > 0, n_0 > 0, \text{forall } n \geq n_0, \text{abs}(f(n)) \leq c * \text{abs}(g(n))\}$
- intuitions:  $f \leq g$ ,  $f$  is at least as efficient as  $g$



# Asymptotic bounds

## Big Omega

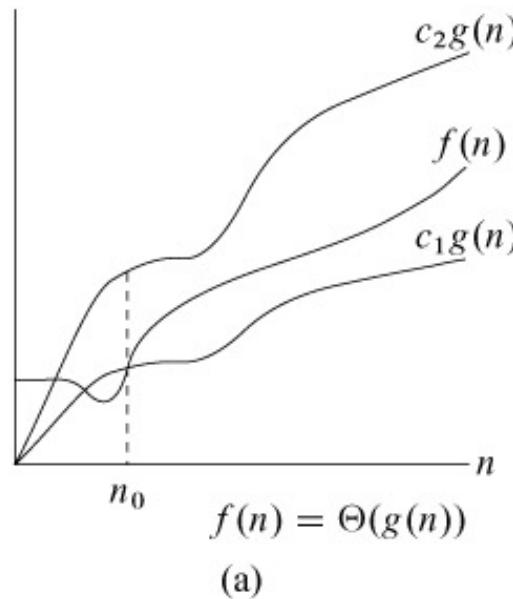
- *asymptotic lower bound*
- $\Omega(g) = \{f \mid \text{exists } c > 0, n_0 > 0, \text{forall } n \geq n_0, \text{abs}(f(n)) \geq c * \text{abs}(g(n))\}$
- intuitions:  $f \geq g$ ,  $f$  is at most as efficient as  $g$



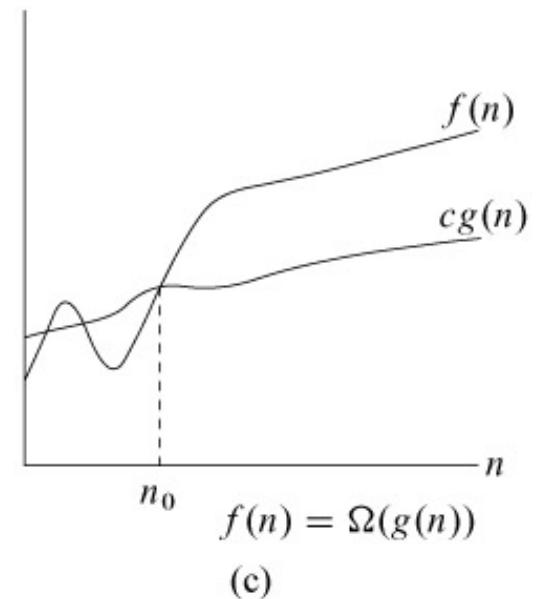
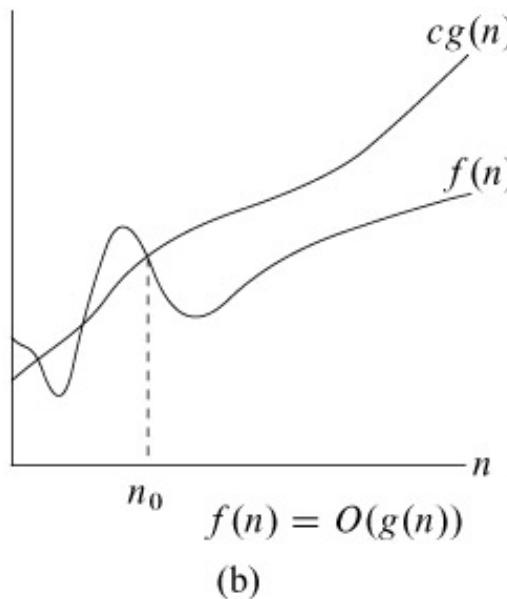
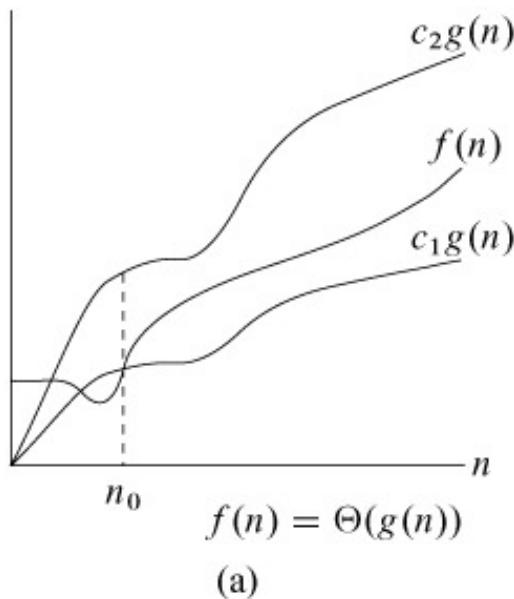
# Asymptotic bounds

## Big Theta

- *asymptotic tight bound*
- $\Theta(g) = O(g) \cap \Omega(g)$
- $\Theta(g) = \{f \mid \text{exists } c_1 > 0, c_2 > 0, n_0 > 0, \text{forall } n \geq n_0,$   
 $c_1 * \text{abs}(g(n)) \leq \text{abs}(f(n)) \leq c_2 * \text{abs}(g(n))\}$
- intuitions:  $f = g$ ,  $f$  is just as efficient as  $g$
- beware: some people write  $O(g)$  when they really mean  $\Theta(g)$



# Asymptotic bounds



# Alternative notions of efficiency

- Expected-case running time
  - Instead of worst case
  - Useful for randomized algorithms
  - Maybe less useful for deterministic algorithms
    - Unless you really do know something about probability distribution of inputs
    - All inputs are probably not equally likely
- Space
  - How much memory is used? Cache space? Disk space?
- Other resources
  - Power, network bandwidth, ...
- Efficiency of an entire data abstraction...

# Stacks with multipop

```
module type STACK = sig
  type 'a t
  exception Empty

  val empty : 'a t
  val is_empty : 'a t -> bool
  val push : 'a -> 'a t -> 'a t
  val peek : 'a t -> 'a
  val pop : 'a t -> 'a t
  val multipop : int -> 'a t -> 'a t
end
```

# Stacks with multipop

```
module Stack : STACK = struct
  type 'a t = 'a list
  exception Empty

  let empty = []
  let is_empty s = s = []
  let push x s = x :: s
  ...
```

# Stacks with multipop

```
module Stack : STACK = struct
  type 'a t = 'a list
  exception Empty

  let empty = []          (* O(1) *)
  let is_empty s = s = [] (* O(1) *)
  let push x s = x :: s  (* O(1) *)
  ...
  ...
```

# Stacks with multipop

```
module Stack : STACK = struct  
  ...  
  let peek = function  
    | []      -> raise Empty  
    | x :: xs -> x  
  
  let pop = function  
    | []       -> raise Empty  
    | x :: xs -> xs  
  ...
```

# Stacks with multipop

```
module Stack : STACK = struct  
  ...  
  let peek = function (* O(1) *)  
    | []    -> raise Empty  
    | x :: xs -> x  
  
  let pop = function (* O(1) *)  
    | []     -> raise Empty  
    | x :: xs -> xs  
  ...
```

# Stacks with multipop

```
module Stack : STACK = struct  
  ...  
  let multipop k s =  
    let rec repeat m f x =  
      if m=0 then x  
      else repeat (m-1) f (f x)  
    in repeat k pop s  
end
```

# Stacks with multipop

```
module Stack : STACK = struct  
  ...  
  let multipop k s =  
    let rec repeat m f x =  
      if m=0 then x  
      else repeat (m-1) f (f x)  
    in repeat k pop s  
    (* imprecise bound: O(n),  
     * where n=length s*)  
end
```

# Question #3

- Start with an initially empty stack
  - Do a sequence of STACK operations
  - Suppose maximum length stack ever reaches is  $n$
  - Suppose (coincidentally) that the sequence of operations is of length  $n$
  - **What is worst-case running time of entire sequence?**
- A.  $O(1)$
  - B.  $O(n)$
  - C.  $O(n \log n)$
  - D.  $O(n^2)$
  - E.  $O(2^n)$

# Question #3

- Start with an initially empty stack
- Do a sequence of STACK operations
- Suppose maximum length stack ever reaches is  $n$
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- **What is worst-case running time of entire sequence?**

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Why?

- $n$  operations
  - each is  $O(n)$
  - $n * O(n) = O(n^2)$
- ...that's correct but pessimistic

# Improved analysis of efficiency

- Consider the average cost of each operation in the sequence, still in the worst case
  - average = arithmetic mean =  $T(n)/n$ 
    - where  $T(n)$  is total worst-case cost of  $n$  operations
  - average  $\neq$  expected value of random variable

# Improved analysis of efficiency

- **Fact:** each value pushed onto stack can be popped off at most once
  - In a sequence of  $n$  operations, can't be more than  $n$  calls to **push**
  - So can't be more than  $n$  calls to **pop**, including calls **multipop** makes to **pop**
  - Each of those calls to **push** and **pop** is  $O(1)$
- So worst-case running time of entire sequence is  $T(n) = n * O(1) = O(n)$
- And average worst-case running time of each operation in sequence is  $T(n)/n = O(n)/n = O(1)$

# A monetary analysis

- Real cost:
  - **push**: \$1
  - **pop**: \$1
  - **multipop**:  $\$min(k, \text{length } s)$
- Let's engage in some "creative accounting"
- Billed cost:
  - **push**: \$2
  - **pop**: \$0
  - **multipop**: \$0
- **Fact:** we can use **billed cost** to pay the **real cost** of any sequence of operations

# A monetary analysis

Sequence	Real cost	Billed cost
push	1	2
push	1	2
pop	1	0
push	1	2
push	1	2
multipop 2	2	0
push	1	2
multipop 3	2	0
<b>TOTAL</b>	<b>10</b>	<b>10</b>

# A monetary analysis

- Cost of **push**:
  - \$2 billed
  - use \$1 of that to pay the real cost
  - save an extra \$1 in that element's "bank account"
- Cost of **pop**:
  - \$0 billed
  - use the saved \$1 in that element's account to pay the real cost
- Cost of **multipop**:
  - (see **pop**)
- So cost of any operation is  $O(1)$ 
  - Because 2 and 0 are both  $O(1)$
- These costs are called *amortized costs*

# A monetary analysis

- Amortized cost of **push**:
  - \$2 billed
  - use \$1 of that to pay the real cost
  - save an extra \$1 in that element's "bank account"
- Amortized cost of **pop**:
  - \$0 billed
  - use the saved \$1 in that element's account to pay the real cost
- Amortized cost of **multi pop**:
  - (see **pop**)
- So amortized cost of any operation is  $O(1)$ 
  - Because 2 and 0 are both  $O(1)$
- These costs are called *amortized costs*

# Amortized analysis of efficiency

- Amortize: put aside money at intervals for gradual payment of debt [Webster's 1964]
  - L. "mort-" as in "death"
- Pay extra money for some operations as a *credit*
- Use that credit to pay higher cost of some later operations
- a.k.a. *banker's method* and *accounting method*
- Invented by Sleator and Tarjan (1985)

# Robert Tarjan



b. 1948

**Turing Award Winner (1986)  
with Prof. John Hopcroft**

*For fundamental achievements in  
the design and analysis of  
algorithms and data structures.*

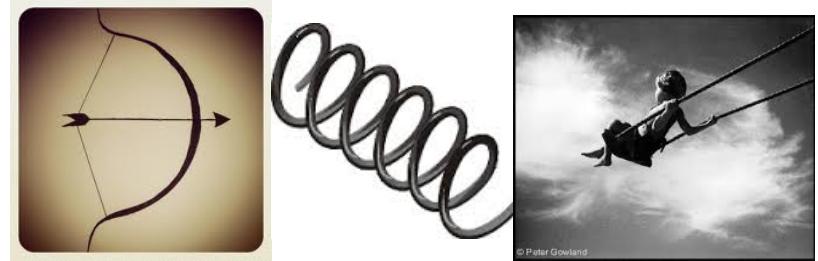
Cornell CS faculty 1972-1973

# Another kind of amortized analysis

- Banker's method required tracking credit from sequence of operations
- Possibly better idea:
  - determine amount of credit available just from state of data structure, not from its history
  - i.e., "let's ignore history"
- Leads to *physicist's method* a.k.a. *potential method*

# Physicist's method

- Potential energy: stored energy of position possessed by an object
  - drawn bow
  - stretched spring
  - child on playground at height of swing
- Suppose we have function  $U(d)$  giving us the "potential energy" stored in a data structure
- We'll use that stored energy to pay for expensive operations



# Physicist's method

- Suppose operation changes data structure from  $d_0$  to  $d_1$
- Define amortized cost of operation to be  
 $= \text{realcost}(\text{op}) + U(d_1) - U(d_0)$
- Amortized cost of sequence of two operations  
 $= \text{realcost}(\text{op}_1) + U(d_1) - U(d_0)$   
 $\quad + \text{realcost}(\text{op}_2) + U(d_2) - U(d_1)$   
 $= \text{realcost}(\text{op}_1) + \text{realcost}(\text{op}_2) + U(d_2) - U(d_0)$
- Amortized cost of sequence of  $n$  operations  
 $= [\sum_{i=1..n} (\text{realcost}(\text{op}_i))] + U(d_n) - U(d_0)$
- *Telescoping sum:* intermediate potentials cancel out; we can ignore them in analysis

# A physical analysis

Potential of stack is length of list:  $U(s) = \text{length}(s)$

Sequence	Real cost	$U(s)$
---	---	0
push	1	1
push	1	2
pop	1	1
push	1	2
push	1	3
multipop 2	2	1
push	1	2
multipop 3	2	0
<b>TOTAL</b>	<b>10</b>	---

# A physical analysis

- Amortized cost of **push**:
  - real cost is 1
  - change in potential is 1
    - because  $U(\mathbf{x} : : \mathbf{s}) - U(\mathbf{s}) = 1$
  - so amortized cost is  $2 = O(1)$

# A physical analysis

- Amortized cost of `pop`:
  - real cost is 1
  - change in potential is  $-1$ 
    - because  $U(\mathbf{s}) - U(\mathbf{x} : : \mathbf{s}) = -1$
  - so amortized cost is  $0 = O(1)$

# A physical analysis

- Amortized cost of `multipop`:
  - real cost is  $\min(k, \text{length}(s))$ . let that be  $k'$ .
  - change in potential is  $-k'$
  - so amortized cost is  $0 = O(1)$
- So amortized cost of any operation is  $O(1)$

# Recall from Lec13: Hash tables

- If load factor gets too high, make the array bigger, thus reducing load factor
  - OCaml **Hashtbl** and **java.util.HashMap**: if load factor > 2.0 then double array size, bringing load factor back to around 1.0
  - Rehash elements into new buckets
  - Efficiency:
    - **insert**: O(1)
    - **find & remove**: O(2), which is O(1)
    - rehashing: arguably still constant time; **will return to this later in course**
- If load factor gets too small (hence memory is being wasted), could shrink the array, thus increasing load factor
  - Neither OCaml nor Java do this

# Hash tables: physicist's method

- Simplifying assumptions:
  - no **remove** operation
  - ignore cost of all operations until load factor reaches 1 for the first time
- Potential:  $U(h) = 4(n - m)$ 
  - where  $n$  is number of elements in  $h$
  - and  $m$  is number of buckets in  $h$
  - Causes potential to increase as load factor ( $=n/m$ ) grows
  - When load factor is 1, it holds that  $m=n$ , so  $U(h) = 0$ 
    - no extra credit stored up immediately after resize
  - When load factor is 2, it holds that  $m=n/2$ , so  $U(h) = 2n$ 
    - enough extra credit stored up to pay to rehash and insert each element just when we need to resize

# Hash tables: physicist's method

- Amortized cost of `insert` (including resize)
  - Let  $n$  be # elements and  $m$  be # buckets before `insert`
  - If no resize is triggered:
    - Cost of 1 each to hash and insert element
    - Change in potential =  $4(n+1-m) - 4(n - m) = 4n + 4 - 4m$   
 $- 4n + 4m = 4$
    - Amortized cost =  $1 + 1 + 4 = 6 = O(1)$

# Hash tables: physicist's method

- Amortized cost of **insert** (including resize)
  - If resize is triggered:
    - Then  $n+1 = 2m$
    - Cost of  $2(n+1)$  to hash and insert  $n+1$  elements
    - Change in potential =  $4(n+1 - 2m) - 4(n - m) = 4n + 4 - 8m - 4n + 4m = 4 - 4m = 4 - 2(2m) = 4 - 2(n+1) = 4 - 2n - 2$
    - Amortized cost =  $2(n + 1) + 4 - 2n - 2 = 2n + 2 + 4 - 2n - 2 = 4 = O(1)$
  - Either way, amortized cost of  $O(1)$

# Hash tables: physicist's method

- Suppose we did have **remove** operation
  - Cost of remove itself is 1 to hash
  - Plus expected worst-case time of at most 2 to delete element from bucket
    - because load factor is at most 2
  - Potential:  $U(h) = \max(4(n - m), 0)$ 
    - No "negative potential" or "negative credit": always pay for expensive operations in advance, otherwise might end a sequence without ever paying off debt
  - Analysis of insert proceeds as before
- Conclusion: resizing hash tables have amortized expected worst-case running time that is constant!
  - Notes have a similar analysis for dynamic arrays using banker's method

Please hold still for 1 more minute

## **WRAP-UP FOR TODAY**

# Upcoming events

- Clarkson office hours cancelled today; extra hour Wednesday 3-4 pm
- **PS6 due on Thursday**, no late passes

*This is money.*

**THIS IS 3110**