

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

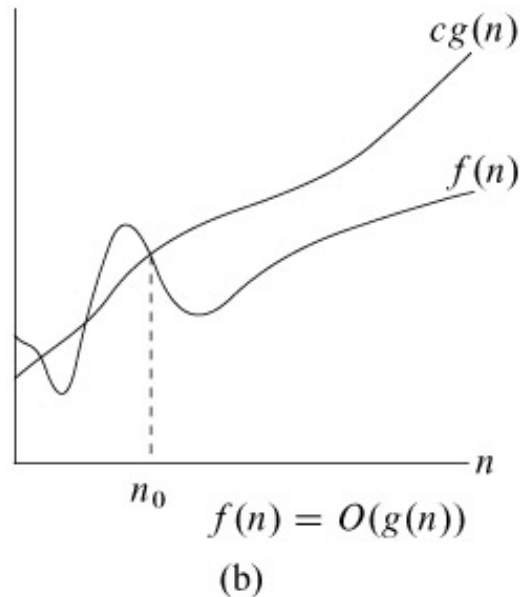
Names of running times

- **$O(1)$** : 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:

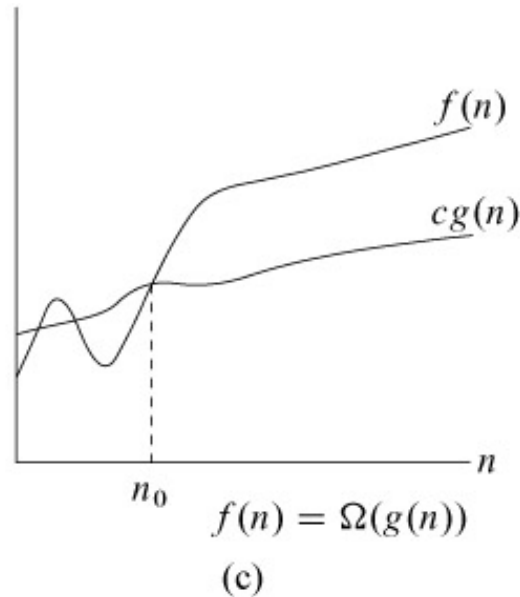
- *asymptotic upper bound*
- $O(g) = \{f \mid \text{exists } c > 0, n_0 > 0, \text{forall } n \geq n_0, \text{abs}(f(n)) \leq c * \text{abs}(g(n))\}$
- intuitions: $f \leq g$, f is at least as efficient as g



Asymptotic bounds

Big Omega

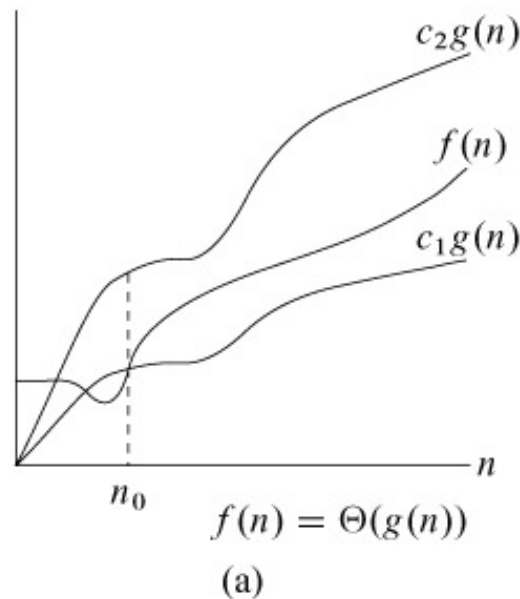
- asymptotic lower bound
- $\Omega(g) = \{f \mid \text{exists } c > 0, n_0 > 0, \text{ for all } n \geq n_0, \text{abs}(f(n)) \geq c * \text{abs}(g(n))\}$
- intuitions: $f \geq g$, f is at most as efficient as g



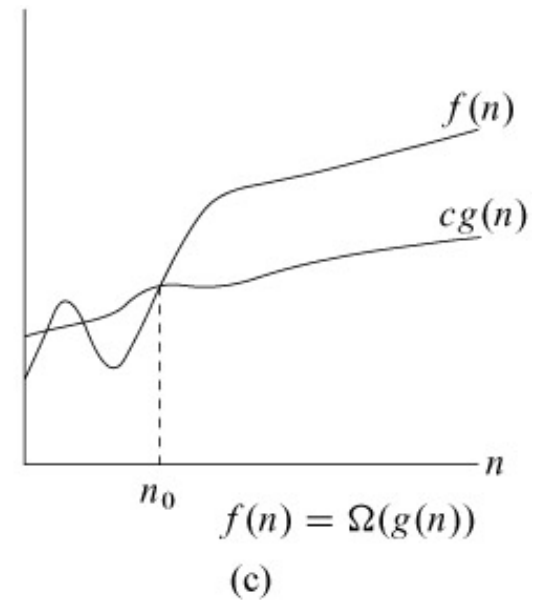
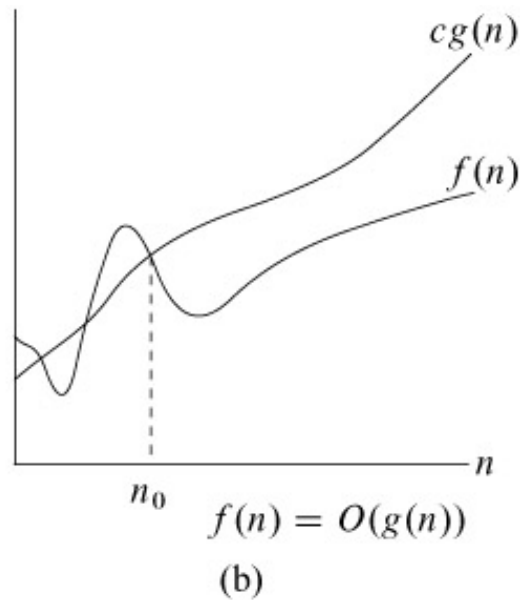
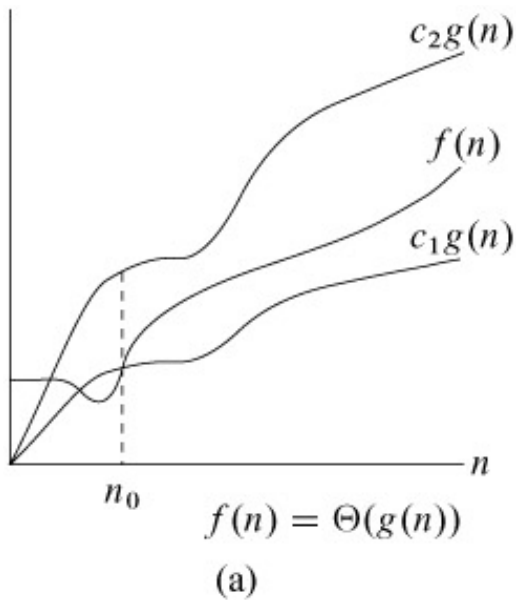
Asymptotic bounds

Big Theta

- asymptotic tight bound
- $\Theta(g) = O(g) \cap \Omega(g)$
- $\Theta(g) = \{f \mid \text{exists } c_1 > 0, c_2 > 0, n_0 > 0, \text{ for all } n \geq n_0, \\ c_1 * \text{abs}(g(n)) \leq \text{abs}(f(n)) \leq c_2 * \text{abs}(g(n))\}$
- 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
- 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. $O(2^n)$

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
 - `multi-pop`: $\$ \min(k, \text{length } s)$
- Let's engage in some "creative accounting"
- Billed cost:
 - `push`: \$2
 - `pop`: \$0
 - `multi-pop`: \$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
TOTAL	10	10

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 **multipop**:
 - (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*

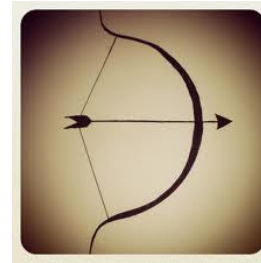
Physicist's method

- 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
 $= \text{realcost}(\text{op}_1) + U(d_1) - U(d_0)$
 $+ \text{realcost}(\text{op}_2) + U(d_2) - U(d_1)$
 $= \text{realcost}(\text{op}_1) + \text{realcost}(\text{op}_2) + U(d_2) - U(d_0)$
- 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
TOTAL	10	---

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