3-2 Amortized Analysis

Hengfeng Wei

hfwei@nju.edu.cn

October 08, 2018





Robert Tarjan



John Hopcroft

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

— Turing Award, 1986

2 / 50

© 1985 Society for Industrial and Applied Mathematics

SIAM J. ALG. DISC. METH Vol. 6, No. 2, April 1985

AMORTIZED COMPUTATIONAL COMPLEXITY*

ROBERT ENDRE TARJAN†

Abstract. A powerful technique in the complexity analysis of data structures is amortization, or averaging over time. Amortized running time is a realistic but robust complexity measure for which we can obtain surprisingly tight upper and lower bounds on a variety of algorithms. By following the principle of designing algorithms whose amortized complexity is low, we obtain "self-adjusting" data structures that are simple, flexible and efficient. This paper surveys recent work by several researchers on amortized complexity.

"Amortized Computational Complexity", 1985

3 / 50

Amortized analysis is

an algorithm analysis technique for
analyzing a sequence of operations
irrespective of the input to show that
the average cost per operation is small, even though
a single operation within the sequence might be expensive.

By averaging the cost per operation over a worst-case sequence,
amortized analysis can yield a time complexity that is
more robust than average-case analysis, since
its probabilistic assumptions on inputs may be false,
and more realistic than worst-case analysis, since it may be
impossible for every operation to take the worst-case time,
as occurs often in manipulation of data structures.

5 / 50





Dynamic Tables

- (I) Summation Method
- (II) Accounting Method
- (III) Potential Method



- (I) Summation Method
- (II) Accounting Method
- (III) Potential Method



Dynamic Tables

"Move-to-Front" List Splay Tree

The Summation Method

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

7 / 50

The Summation Method

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$\forall i, \ \hat{c_i} = \frac{\left(\sum\limits_{i=1}^n c_i\right)}{n}$$

7 / 50

```
o_i: o_1 o_2 o_3 o_4 o_5 o_6 o_7 o_8 o_9 o_{10}
c_i: 1 2 3 1 5 1 1 9 1
```

On any sequence of n INSERTs on an initially empty array.

$$o_i: o_1 \quad o_2 \quad o_3 \quad o_4 \quad o_5 \quad o_6 \quad o_7 \quad o_8 \quad o_9 \quad o_{10}$$

 $c_i: 1 \quad 2 \quad 3 \quad 1 \quad 5 \quad 1 \quad 1 \quad 1 \quad 9 \quad 1$

$$c_i = \begin{cases} (i-1) + 1 = i & \text{if } i-1 \text{ is an exact power of 2} \\ 1 & \text{o.w.} \end{cases}$$

8 / 50

On any sequence of n INSERTS on an initially empty array.

$$o_i$$
: o_1 o_2 o_3 o_4 o_5 o_6 o_7 o_8 o_9 o_{10}
 c_i : 1 2 3 1 5 1 1 9 1

$$c_i = \begin{cases} (i-1) + 1 = i & \text{if } i-1 \text{ is an exact power of } 2\\ 1 & \text{o.w.} \end{cases}$$

$$\sum_{i=1}^{n} c_i \le n + \sum_{j=0}^{\lceil \lg n \rceil - 1} 2^j = n + (2^{\lceil \lg n \rceil} - 1) \le n + 2n = 3n$$

8 / 50

$$c_i = \begin{cases} (i-1) + 1 = i & \text{if } i-1 \text{ is an exact power of 2} \\ 1 & \text{o.w.} \end{cases}$$

$$\sum_{i=1}^{n} c_i \le n + \sum_{j=0}^{\lceil \lg n \rceil - 1} 2^j = n + (2^{\lceil \lg n \rceil} - 1) \le n + 2n = 3n$$





The Summation Method

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

The Summation Method

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$\forall i, \ \hat{c_i} = \frac{\left(\sum\limits_{i=1}^n c_i\right)}{n}$$

```
o_i: o_1 o_2 o_3 o_4 o_5 o_6 o_7 o_8 o_9 o_{10}
c_i: 1 2 3 1 5 1 1 9 1
```

$$o_i: o_1 \quad o_2 \quad o_3 \quad o_4 \quad o_5 \quad o_6 \quad o_7 \quad o_8 \quad o_9 \quad o_{10}$$

 $c_i: 1 \quad 2 \quad 3 \quad 1 \quad 5 \quad 1 \quad 1 \quad 1 \quad 9 \quad 1$

$$c_i = \begin{cases} (i-1) + 1 = i & \text{if } i-1 \text{ is an exact power of 2} \\ 1 & \text{o.w.} \end{cases}$$

$$c_i = \begin{cases} (i-1) + 1 = i & \text{if } i-1 \text{ is an exact power of } 2\\ 1 & \text{o.w.} \end{cases}$$

$$\sum_{i=1}^{n} c_i \le n + \sum_{j=0}^{\lceil \lg n \rceil - 1} 2^j = n + (2^{\lceil \lg n \rceil} - 1) \le n + 2n = 3n$$

$$c_i = \begin{cases} (i-1) + 1 = i & \text{if } i-1 \text{ is an exact power of 2} \\ 1 & \text{o.w.} \end{cases}$$

$$\sum_{i=1}^{n} c_i \le n + \sum_{j=0}^{\lceil \lg n \rceil - 1} 2^j = n + (2^{\lceil \lg n \rceil} - 1) \le n + 2n = 3n$$

$$\forall i, \ \hat{c_i} = 3$$



$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$a_1, a_2, \ldots, a_n$$

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$a_1, a_2, \ldots, a_n$$

$$\hat{c_i} = c_i + a_i, \ a_i > = < 0$$

 $\mathsf{Amortized}\ \mathsf{Cost}\ =\ \mathsf{Actual}\ \mathsf{Cost}\ +\ \mathsf{Accounting}\ \mathsf{Cost}$

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$a_1, a_2, \ldots, a_n$$

$$\hat{c_i} = c_i + a_i, \ a_i > = < 0$$

Amortized Cost = Actual Cost + Accounting Cost

$$\forall n, \ \sum_{i=1}^{n} c_i \le \sum_{i=1}^{n} \hat{c_i}$$

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$a_1, a_2, \ldots, a_n$$

$$|\hat{c_i} = c_i + a_i, \ a_i > = < 0$$

Amortized Cost = Actual Cost + Accounting Cost

$$\forall n, \sum_{i=1}^{n} c_i \leq \sum_{i=1}^{n} \hat{c}_i \implies \boxed{\forall n, \sum_{i=1}^{n} a_i \geq 0}$$

$$o_1, o_2, \ldots, o_n$$

$$c_1, c_2, \ldots, c_n$$

$$a_1, a_2, \ldots, a_n$$

$$|\hat{c_i} = c_i + a_i, \ a_i > = < 0$$

Amortized Cost = Actual Cost + Accounting Cost

$$\forall n, \sum_{i=1}^{n} c_i \leq \sum_{i=1}^{n} \hat{c}_i \implies \boxed{\frac{\forall n, \sum_{i=1}^{n} a_i \geq 0}{}}$$

Key Point: Put the accounting cost on specific objects.

The Accounting Method for Array Doubling

$$Q: \hat{c_i} = 3$$
 vs. $\hat{c_i} = 2$

The Accounting Method for Array Doubling

$$Q: \hat{c_i} = 3$$
 vs. $\hat{c_i} = 2$

$$\hat{c_i} = 3 = \underbrace{1}_{\text{insert}} + \underbrace{1}_{\text{move itself}} + \underbrace{1}_{\text{help move another}}$$

The Accounting Method for Array Doubling

$$Q: \hat{c_i} = 3$$
 vs. $\hat{c_i} = 2$

$$\hat{c_i} = 3 = \underbrace{1}_{\text{insert}} + \underbrace{1}_{\text{move itself}} + \underbrace{1}_{\text{help move another}}$$

	$\hat{c_i}$	c_i (actual cost)	a_i (accounting cost)
Insert (normal)	3	1	2
Insert (expansion)	3	1+t	-t+2

Simulating a queue Q using two stacks S_1, S_2 (Problem $\mathbb{E}3$)

```
procedure \operatorname{EnQ}(x)

\operatorname{Push}(S_1,x)

procedure \operatorname{DeQ}()

if S_2 = \emptyset then

while S_1 \neq \emptyset do

\operatorname{Push}(S_2,\operatorname{Pop}(S_1))

\operatorname{Pop}(S_2)
```

The Summation Method for Queue Simulation

$$\frac{\left(\sum_{i=1}^{n} c_i\right)}{n}$$

The Summation Method for Queue Simulation

$$\frac{\left(\sum\limits_{i=1}^{n}c_{i}\right)}{n}$$

The operation sequence is **NOT** known.

The Accounting Method for Queue Simulation

item: Push into
$$S_1$$
 Pop from S_1 Push into S_2 Pop from S_2 1 1 1

The Accounting Method for Queue Simulation

item: Push into
$$S_1$$
 Pop from S_1 Push into S_2 Pop from S_2
$$1 \qquad 1 \qquad 1 \qquad 1$$

$$\hat{c}_{\rm ENQ}=3$$

$$\hat{c}_{\rm DEQ}=1$$

The Accounting Method for Queue Simulation

$$\hat{c}_{\text{ENQ}} = 3$$

$$\hat{c}_{\mathrm{DEQ}} = 1$$

$$\sum_{i=1}^{n} a_i \ge 0$$

The Accounting Method for Queue Simulation

$$\hat{c}_{\text{ENQ}} = 3$$

$$\hat{c}_{\mathrm{DeQ}} = 1$$

$$\sum_{i=1}^{n} a_i \ge 0 \Longleftrightarrow \sum_{i=1}^{n} a_i = \#S_1 \times 2$$

The Accounting Method for Queue Simulation

$$\hat{c}_{\text{ENQ}} = 3$$
 $\hat{c}_{\text{DEQ}} = 1$

$$\#S_1 = t$$

	$\hat{c_i}$	c_i (actual cost)	a_i (accounting cost)
Enqueue	3	1	2
Dequeue $(S_2 = \emptyset)$	1	1	0
DEQUEUE $(S_2 \neq \emptyset)$	1	1+2t	-2t

"Move-to-Front" (MTF) List

What work are you proudest of?



What work are you proudest of?



Proudest? It's hard to choose.

What work are you proudest of?



Proudest? It's hard to choose.

I like the **self-adjusting search tree** data structure that Danny Sleator and I developed.

Self-Adjusting Binary Search Trees

DANIEL DOMINIC SLEATOR AND ROBERT ENDRE TARJAN

AT&T Bell Laboratories, Murray Hill, NJ

Abstract. The splay tree, a self-adjusting form of binary search tree, is developed and analyzed. The binary search tree is a data structure for representing tables and lists so that accessing, inserting, and deleting items is easy. On an n-node splay tree, all the standard search tree operations have an amortized time bound of O(log n) per operation, where by "amortized time" is meant the time per operation averaged over a worst-case sequence of operations. Thus splay trees are as efficient as balanced trees when total running time is the measure of interest. In addition, for sufficiently long access sequences, splay trees are as efficient, to within a constant factor, as static optimum search trees. The efficiency of splay trees comes not from an explicit structural constraint, as with balanced trees, but from applying a simple restructuring heuristic, called splaying, whenever the tree is accessed. Extensions of splaying give simplified forms of two other data structures: lexicographic or multidimensional search trees and link/ cut trees.

"Self-Adjusting Binary Search Trees - Splay Tree", JACM, 1985

32 / 50

Self-Adjusting Binary Search Trees



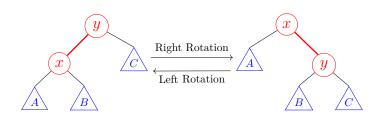
vs. Balanced Binary Search Trees

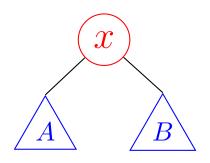
Splay(x):

Moving node x to the root of the tree by performing a sequence of rotations along the path from x to the root.

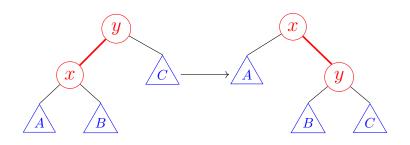
SPLAY(x):

Moving node x to the root of the tree by performing a sequence of rotations along the path from x to the root.

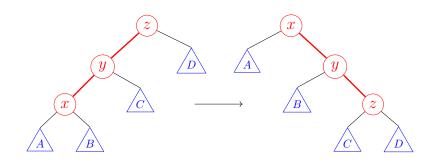




Case I: x is the root



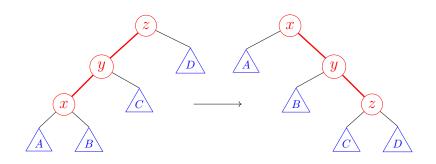
Case II: y = p(x) is the root



CASE III: zig-zig

$$y = p(x)$$
 $z = p(y)$

$$x = lc(y)$$
 $y = lc(z)$

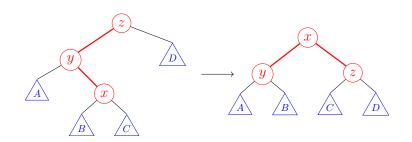


${\rm CASE~III:~zig\text{-}zig}$

$$y = p(x)$$
 $z = p(y)$

$$x = lc(y)$$
 $y = lc(z)$

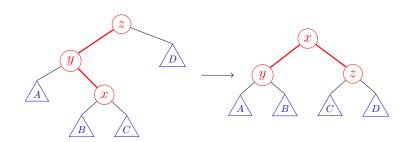
$$(1): y-z$$
 $(2): x-y$



$CASE\ IV: \ \hbox{zig-zag}$

$$y = p(x)$$
 $z = p(y)$

$$x = rc(y)$$
 $y = lc(z)$

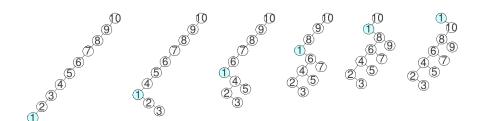


CASE IV: zig-zag

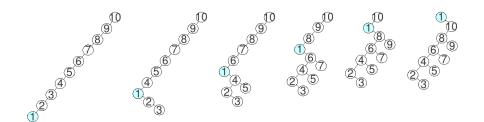
$$y = p(x)$$
 $z = p(y)$

$$x = rc(y)$$
 $y = lc(z)$

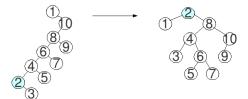
$$(1): x - y$$
 $(2): x - z$



Splay(1)



Splay(1)



Splay(2)

Amortized analysis of $\operatorname{SPLAY} % {\operatorname{SPLAY}}$

Amortized analysis of Splay

Amortized analysis of Splay

A splay tree T of $\emph{n}\text{-node}$ An arbitrary sequence of \emph{m} Splay operations

Theorem

$$\hat{c}_{\text{SPLAY}} = O(\log n).$$

$$\Phi_0$$
 Splay₁ Φ_1 Splay₂ Φ_2 \cdots $\underbrace{\Phi_{i-1}}_{\mathsf{the } i \mathsf{-th}} \underbrace{\mathsf{Splay}}_{\mathsf{i}} \underbrace{\mathsf{Ord}}_{\mathsf{i}} \underbrace{\mathsf{Splay}}_{\mathsf{i}} \Phi_m$

$$\hat{c}_{\text{SPLAY}_i} = c_{\text{SPLAY}_i} + (\Phi_{\text{SPLAY}_i} - \Phi_{\text{SPLAY}_{i-1}})$$

$$\Phi_0$$
 Splay₁ Φ_1 Splay₂ Φ_2 \cdots $\underbrace{\Phi_{i-1}}$ Splay_i Φ_i \cdots Splay_m Φ_m

$$\hat{c}_{\text{SPLAY}_i} = c_{\text{SPLAY}_i} + (\Phi_{\text{SPLAY}_i} - \Phi_{\text{SPLAY}_{i-1}})$$

How to define Φ ?

$$r(x) = \log s(x)$$

$$r(x) = \log s(x)$$

$$\Phi = \sum_{x \in T} r(x)$$

$$r(x) = \log s(x)$$

$$\Phi = \sum_{x \in T} r(x)$$

How to calculate $(\Phi_{SPLAY_i} - \Phi_{SPLAY_{i-1}})$ and c_{SPLAY_i} ?

$$\hat{c}_{\mathrm{SPLAY}_i} = c_{\mathrm{SPLAY}_i} + (\Phi_{\mathrm{SPLAY}_i} - \Phi_{\mathrm{SPLAY}_{i-1}})$$



 Φ_0 Splay₁ Φ_1 Splay₂ Φ_2 \cdots $\underbrace{\Phi_{i-1}$ Splay_i Φ_i \cdots Splay_m Φ_m

$$\Phi_0$$
 Splay₁ Φ_1 Splay₂ Φ_2 \cdots $\underbrace{\Phi_{i-1}}_{\text{the } i\text{-th Splay}} \underbrace{\Phi_i}_{\text{the } i\text{-th Splay}} \cdots$ Splay_m Φ_m

$$\Phi_{i-1} \xrightarrow{\text{SPLAY}_i} \Phi_i :$$

$$\Phi_{i-1} \triangleq \Phi_{0'} \text{ ITER}_1 \Phi_{1'} \cdots \underbrace{\Phi_{k-1} \text{ ITER}_k \Phi_k}_{\text{the k-th ITERATION}} \cdots \text{ ITER}_l \Phi_l \triangleq \Phi_i$$

$$\Phi_0$$
 Splay₁ Φ_1 Splay₂ Φ_2 · · · · Φ_{i-1} Splay_i Φ_i · · · Splay_m Φ_m

$$\Phi_{i-1} \xrightarrow{\text{SPLAY}_i} \Phi_i :$$

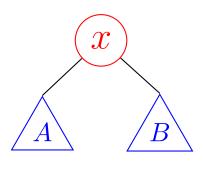
$$\Phi_{i-1} \triangleq \Phi_{0'} \text{ ITER}_1 \Phi_{1'} \cdots \underbrace{\Phi_{k-1} \text{ ITER}_k \Phi_k}_{\text{the k-th ITERATION}} \cdots \text{ ITER}_l \Phi_l \triangleq \Phi_i$$

$$\begin{split} \hat{c}_{\mathrm{SPLAY}_i} &= \sum_{1 \leq j \leq l} \hat{c}_{\mathrm{ITER}_j} \\ &= \sum_{1 \leq j \leq n} c_{\mathrm{ITER}_j} + (\Phi_{\mathrm{ITER}_j} - \Phi_{\mathrm{ITER}_{j-1}}) \end{split}$$

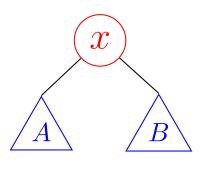
$$\hat{c}_{\text{ITER}_j} = c_{\text{ITER}_j} + (\Phi_{\text{ITER}_j} - \Phi_{\text{ITER}_{j-1}})$$

$$\hat{c}_{\text{ITER}_j} = c_{\text{ITER}_j} + (\Phi_{\text{ITER}_j} - \Phi_{\text{ITER}_{j-1}})$$

By Case Analysis.



Case I

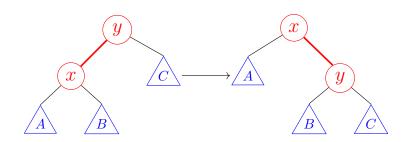


Case I

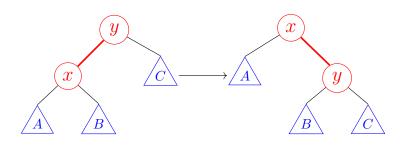
$$\hat{c}_{\text{ITER}_j} = c_{\text{ITER}_j} + (\Phi_{\text{ITER}_j} - \Phi_{\text{ITER}_{j-1}})$$

$$= 0 + 0$$

$$= 0$$



Case II

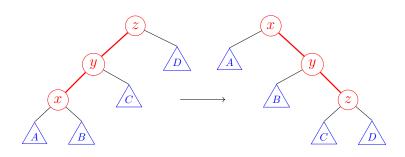


Case II

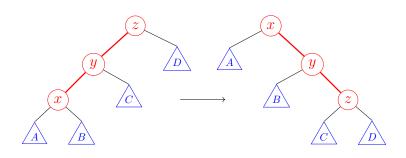
$$\hat{c}_{\text{ITER}_j} = c_{\text{ITER}_j} + (\Phi_{\text{ITER}_j} - \Phi_{\text{ITER}_{j-1}})$$

$$= 0 + 0$$

$$= 0$$



Case III

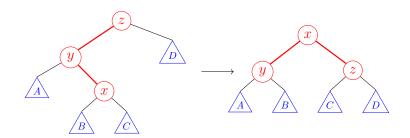


Case III

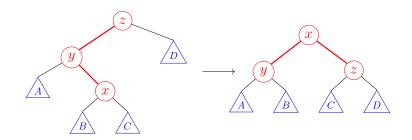
$$\hat{c}_{\text{ITER}_j} = c_{\text{ITER}_j} + (\Phi_{\text{ITER}_j} - \Phi_{\text{ITER}_{j-1}})$$

$$= 0 + 0$$

$$= 0$$



Case IV



Case IV

$$\hat{c}_{\text{ITER}_j} = c_{\text{ITER}_j} + (\Phi_{\text{ITER}_j} - \Phi_{\text{ITER}_{j-1}})$$

$$= 0 + 0$$

$$= 0$$





Office 302

Mailbox: H016

hfwei@nju.edu.cn