

# طراحی الگوریتم

۲۱ و ۲۳ مهر ۹۸

ملکی مجد

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Recursion and Backtracking	Ch.1 and Ch.2 JeffE
Dynamic Programming	Ch.3 JeffE and Ch.15 CLRS
Greedy Algorithms	Ch.4 JeffE and Ch.16 CLRS
Amortized Analysis	Ch.17 CLRS
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# Pseudocode of Huffman's algorithm

HUFFMAN( $C$ )

1  $n \leftarrow |C|$

2  $Q \leftarrow C$

3 **for**  $i \leftarrow 1$  **to**  $n - 1$

4     **do** allocate a new node  $z$

5          $left[z] \leftarrow x \leftarrow \text{EXTRACT-MIN}(Q)$

6          $right[z] \leftarrow y \leftarrow \text{EXTRACT-MIN}(Q)$

7          $f[z] \leftarrow f[x] + f[y]$

8         INSERT( $Q, z$ )

9 **return** EXTRACT-MIN( $Q$ )

▷ Return the root of the tree.

# Correctness of Huffman's algorithm

- Let  $C$  be an alphabet in which each character  $c \in C$  has frequency  $f[c]$ . Let  $x$  and  $y$  be two characters in  $C$  having the **lowest frequencies**. Then there exists an optimal prefix code for  $C$  in which the codewords for  $x$  and  $y$  have the **same length and differ only in the last bit**.
- Let  $C$  be a given alphabet with frequency  $f[c]$  defined for each character  $c \in C$ . Let  $x$  and  $y$  be two characters in  $C$  with **minimum frequency**. Let  $C_2$  be the alphabet  $C$  with characters  $x, y$  removed and (new) character  $z$  added, so that  $C_2 = C - \{x, y\} \cup \{z\}$ ; define  $f$  for  $C_2$  as for  $C$ , except that  $f[z] = f[x] + f[y]$ . Let  $T_2$  be any tree representing an optimal prefix code for the alphabet  $C_2$ . Then the tree  $T$ , obtained from  $T_2$  by replacing the leaf node for  $z$  with an internal node having  $x$  and  $y$  as children, represents an optimal prefix code for the alphabet  $C$ .

# Correctness of Huffman's algorithm

- اثبات لم ها در کلاس توضیح داده شده است.

# Amortized Analysis

- In an *amortized analysis*, the time required to perform a sequence of data-structure operations is averaged over all the operations performed.
- the average cost of an operation is small
  - even though a single operation within the sequence might be expensive.
- Amortized analysis differs from average-case analysis in that probability is not involved
  - an amortized analysis guarantees the average performance of each operation in the worst case

## three most common techniques used in amortized analysis.

- aggregate analysis
- the accounting method
- the potential method

• یادآوری:

- هزینه های نسبت داده شده در فرایند تحلیل سرشکن، برای تحلیل استفاده می شود. در کُد برنامه نویسی، از این هزینه ها استفاده نمی کنیم.

# Aggregate analysis

- we show that for all  $n$ , a sequence of  $n$  operations takes worst-case time  $T(n)$  in total. In the worst case, the average cost, or **amortized cost**, per operation is therefore  $T(n)/n$ .
  - all operations have the same amortized cost.

# The accounting method

- When there is more than one type of operation
  - each type of operation may have a different amortized cost
- we assign differing charges to different operations, with some operations charged **more or less than they actually cost**.
  - The amount we charge an operation is called its ***amortized cost***.
- When an operation's amortized cost exceeds its actual cost, the difference is assigned to specific objects in the data structure as prepaid ***credit***.
  - Credit can be used later on to help pay for operations whose amortized cost is less than their actual cost.

## The accounting method (2)

- This method is very different from aggregate analysis, in which all operations have the same amortized cost
- The total credit stored in the data structure is the difference between the total amortized cost and the total actual cost
  - Must be greater than zero
- the total amortized cost of a sequence of operations must be an upper bound on the total actual cost of the sequence

$$\sum_{i=1}^n \hat{c}_i \geq \sum_{i=1}^n c_i$$

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# The potential method

- Is like the accounting method in that we **determine the amortized cost of each operation and may overcharge operations** early on to compensate for undercharges later
- maintains the credit as the “potential energy” of the data structure **as a whole instead of associating the credit with individual objects** within the data structure
- For each  $i = 1, 2, \dots, n$ , we let  $c_i$  be the actual cost of the  $i$ th operation
- $D_i$  be the data structure that results after applying the  $i$ th operation to data structure  $D_{i-1}$ .
- A **potential function** maps each data structure  $D_i$  to a real number  $\Phi(D_i)$ ,
- **amortized cost**  $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$ .

## The potential method (2)

- the total amortized cost of the  $n$  operations is

$$\begin{aligned}\sum_{i=1}^n \hat{c}_i &= \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1})) \\ &= \sum_{i=1}^n c_i + \Phi(D_n) - \Phi(D_0) .\end{aligned}$$

- require that  $\Phi(D_i) \geq \Phi(D_0)$  for all  $i$  ,
  - guarantee, as in the accounting method, that we pay in advance
  - define  $(D_0)$  to be 0 and then show that  $(D_i) \geq 0$  for all  $i$
- $((D_i) \geq 0 \text{ for all } i) \Rightarrow$  the total amortized cost of a sequence of  $n$  INCREMENT operations is an upper bound on the total actual cost

## Example1 : stack

- **Stack operations**

PUSH( $S, x$ ) pushes object  $x$  onto stack  $S$ .  $O(1)$

POP( $S$ ) pops the top of stack  $S$  and returns the popped object.  $O(1)$

MULTIPOP( $S, k$ ) // *pops several objects at once  $O(\min(s, k))$*

**while** not STACK-EMPTY( $S$ ) and  $k \neq 0$

**do** POP( $S$ )

$k \leftarrow k - 1$

## Example1 : stack

- aggregate analysis
  - For any value of  $n$ , any sequence of  $n$  PUSH, POP, and MULTIPOP operations takes a total of  $O(n)$  time. The average cost of an operation is  $O(n)/n = O(1)$ .
  - all three stack operations have an amortized cost of  $O(1)$ .
- The accounting method

actual costs:

PUSH	1 ,
POP	1 ,
MULTIPOP	$\min(k, s)$ ,

amortized costs:

PUSH	2 ,
POP	0 ,
MULTIPOP	0 .

## Example1 : stack (The potential method)

- define the potential function on a stack to be the number of objects in the stack

$$\begin{aligned}\Phi(D_i) &\geq 0 \\ &= \Phi(D_0) .\end{aligned}$$

$$\Phi(D_i) - \Phi(D_{i-1}) = (s + 1) - s = 1$$

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &= 1 + 1 = 2 .\end{aligned}$$

## Example1 : stack (The potential method 2)

- the amortized cost of the MULTIPOP operation is

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &= k' - k' \\ &= 0 .\end{aligned}$$

- The amortized cost of each of the three operations is  $O(1)$

## Example2: a binary counter

- a k-bit binary counter
  - counts up from 0 by means of the single operation INCREMENT.
- use an array  $A[0..k-1]$  of bits, where  $\text{length}[A] = k$ , as the counter.
  - A binary number  $x$  that is stored in the counter has its lowest-order bit in  $A[0]$  and its highest-order bit in  $A[k-1]$
- Initially,  $x = 0$ , and thus  $A[i] = 0$  for  $i = 0, 1, \dots, k-1$ .
  - To add 1 (modulo  $2^k$ ) to the value in the counter, we use the following procedure.
- INCREMENT( $A$ )
  - $i \leftarrow 0$
  - while  $i < \text{length}[A]$  and  $A[i] = 1$ 
    - do  $A[i] \leftarrow 0$
    - $i \leftarrow i + 1$
  - if  $i < \text{length}[A]$ 
    - then  $A[i] \leftarrow 1$

# incrementing a binary counter

Counter value	A[7]	A[6]	A[5]	A[4]	A[3]	A[2]	A[1]	A[0]	Total cost
0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	1	1
2	0	0	0	0	0	0	1	0	3
3	0	0	0	0	0	0	1	1	4
4	0	0	0	0	0	1	0	0	7
5	0	0	0	0	0	1	0	1	8
6	0	0	0	0	0	1	1	0	10
7	0	0	0	0	0	1	1	1	11
8	0	0	0	0	1	0	0	0	15
9	0	0	0	0	1	0	0	1	16
10	0	0	0	0	1	0	1	0	18
11	0	0	0	0	1	0	1	1	19
12	0	0	0	0	1	1	0	0	22
13	0	0	0	0	1	1	0	1	23
14	0	0	0	0	1	1	1	0	25
15	0	0	0	0	1	1	1	1	26
16	0	0	0	1	0	0	0	0	31

## aggregate analysis

- The total number of flips in the sequence is

$$\sum_{i=0}^{\lfloor \lg n \rfloor} \left\lfloor \frac{n}{2^i} \right\rfloor < n \sum_{i=0}^{\infty} \frac{1}{2^i} = 2n$$

- The worst-case time for a sequence of  $n$  INCREMENT operations on an initially zero counter is therefore  $O(n)$ . The average cost of each operation, and therefore the amortized cost per operation, is  $O(n)/n = O(1)$ .

# The accounting method

- charge an amortized cost of 2 dollars to set a bit to 1
  - When a bit is set, we use **1 dollar** (out of the 2 dollars charged) to pay for the actual setting of the bit, and we place the **other dollar** on the bit as **credit to be used later** when we flip the bit back to 0.
- The number of 1's in the counter is never negative, and thus the amount of credit is always nonnegative
- for  $n$  INCREMENT operations, the total amortized cost is  $O(n)$ , which bounds the total actual cost.

# The potential method

- define the potential of the counter after the  $i$ th INCREMENT operation to be  $b_i$ , the number of 1's in the counter after the  $i$ th operation
- The actual cost of the operation (the  $i$ th INCREMENT operation resets  $t_i$  bits)
  - at most  $t_i + 1$
  - If  $b_i = 0$ ,  $b_{i-1} = t_i = k$ .
  - If  $b_i > 0$ , then  $b_i = b_{i-1} - t_i + 1$

## The potential method(2)

- In either case,  $b_i \leq b_{i-1} - t_i + 1$ , and the potential difference is

$$\begin{aligned}\Phi(D_i) - \Phi(D_{i-1}) &\leq (b_{i-1} - t_i + 1) - b_{i-1} \\ &= 1 - t_i .\end{aligned}$$

- The amortized cost is therefore

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &\leq (t_i + 1) + (1 - t_i) \\ &= 2 .\end{aligned}$$

analyze the counter even when it does not start at zero

- There are initially  $b_0$  1's, and after  $n$  INCREMENT operations there are  $b_n$  1's, where  $0 \leq b_0, b_n \leq k$ .

$$\sum_{i=1}^n c_i = \sum_{i=1}^n \hat{c}_i - \Phi(D_n) + \Phi(D_0) .$$

$$\begin{aligned} \sum_{i=1}^n c_i &\leq \sum_{i=1}^n 2 - b_n + b_0 \\ &= 2n - b_n + b_0 . \end{aligned}$$

analyze the counter even when it does not start at zero

- since  $b_0 \leq k$ , as long as  $k = O(n)$ , the total actual cost is  $O(n)$
- if we execute at least  $n = (k)$  INCREMENT operations, the total actual cost is  $O(n)$ , no matter what initial value the counter contains

$$\begin{aligned}\sum_{i=1}^n c_i &\leq \sum_{i=1}^n 2 - b_n + b_0 \\ &= 2n - b_n + b_0 .\end{aligned}$$