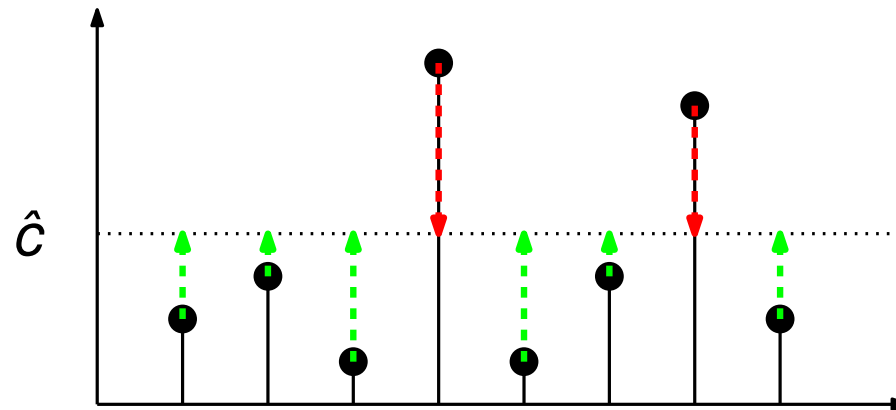




ICS 621: Analysis of Algorithms

Prof. Nodari Sitchinava



Amortized Analysis

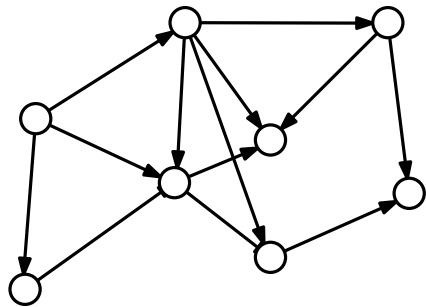
Last time: Simple analysis

```
procedure TRIPLELOOP( $A, n$ )  
  for  $k = 1$  to  $10$  do  
    for  $i = 1$  to  $n$  do  
      for  $j = n$  downto  $i$  do  
         $A[i] = A[j] + 1$ 
```

```
function FOO( $n$ )  
  if then  $n < 2$   
    return  $1$   
  else  
    for do  $i = 1$  to  $n$   
       $x = x + 1$   
    for do  $i = 1$  to  $2$   
       $x = x + \text{FOO}(n/4)$   
  return  $x$ 
```

Today: More complex analysis

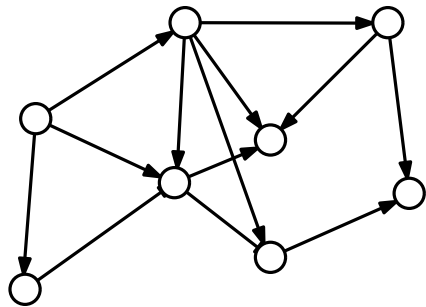
```
procedure PROCESSGRAPH( $G$ )  
  for each vertex  $u \in G.V$  do  
     $u.color = \text{WHITE}$   
     $u.\pi = \text{NIL}$   
 $time = 0$   
  for each vertex  $u \in G.V$  do  
    if  $u.color == \text{WHITE}$  then  
      VISIT( $G, u$ )
```



```
procedure VISIT( $G, u$ )  
   $time = time + 1$   
   $u.d = time$   
   $u.color = \text{GRAY}$   
  for each  $v \in G.Adj[u]$  do  
    if  $v.color == \text{WHITE}$  then  
       $v.\pi = u$   
      VISIT( $G, v$ )  
  
   $u.color = \text{BLACK}$   
   $time = time + 1$   
   $u.f = time$ 
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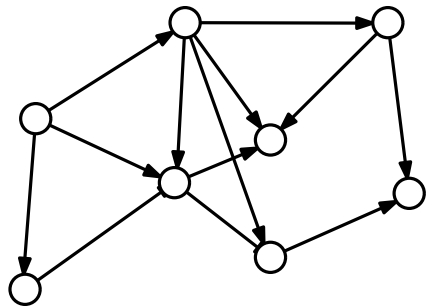


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$$T(n) \leq (n-1) \cdot T(n-1) + O(1) \leq \Theta(n^2)$$

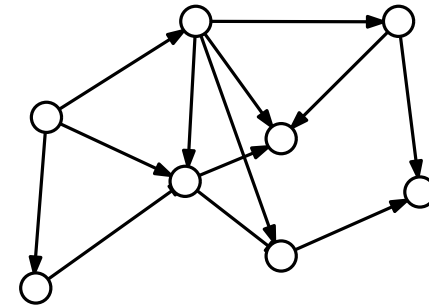
More precise analysis

of vertices

$$\sum_{i=1}^n (1 + d_i) = n + \sum_{i=1}^n d_i = n + m$$

out-degree of
the i -th vertex

of edges



Amortized analysis

If i -th operation takes time ("costs") c_i time steps, then total cost of an algorithm with n operations:

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$$\hat{c} = \frac{T(n)}{n}$$

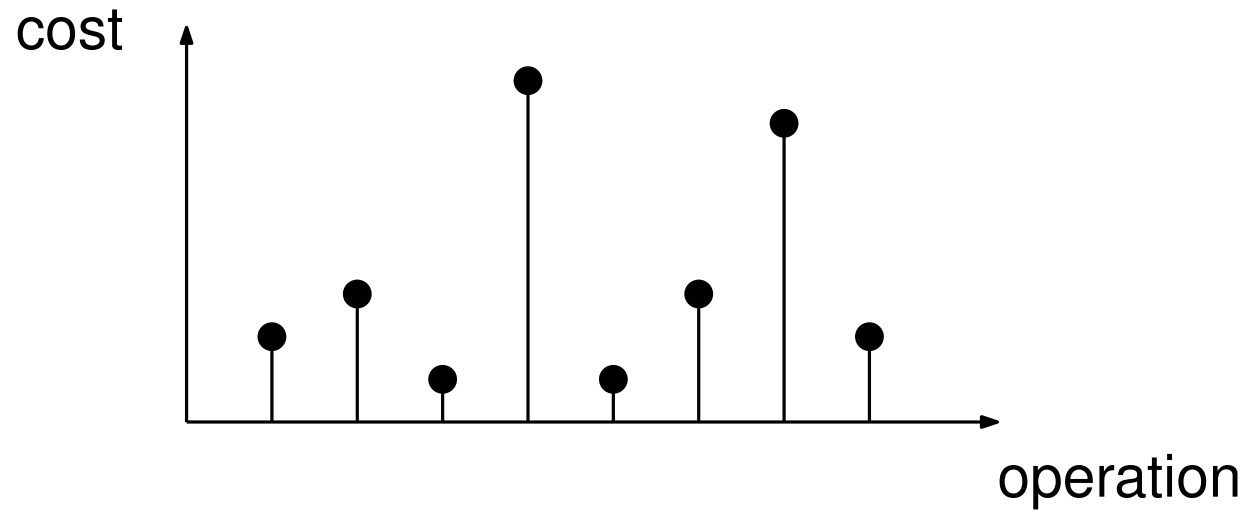
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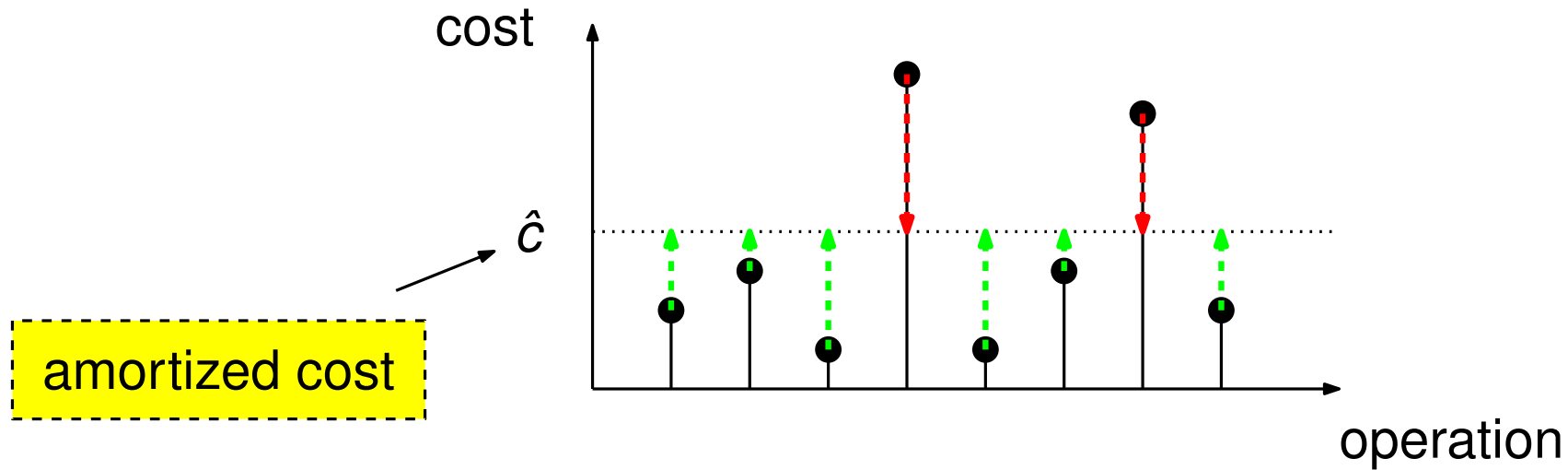
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Examples

$$\hat{c} = \Theta(\log n)$$

$$T(n) \leq \sum_{i=1}^n \hat{c} = \Theta(n \log n)$$

worst-case

Amortized cost \neq average cost = expected cost

Amortized cost is
worst-case upper bound

Expected cost is for
randomized algorithms

Examples

$$\hat{c} = \Theta(\log n)$$

$$T(n) \leq \sum_{i=1}^n \hat{c} = \Theta(n \log n)$$

worst-case

Quicksort

$$T(n) = \Theta(n \log n)$$

expected

Exist inputs for which
 $T(n) = \Theta(n^2)$

Amortized analysis: Aggregate analysis

$$\hat{c} = \frac{1}{n} \cdot \sum_{i=1}^n c_i$$

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Example: Binary Counter

B = array of k bits

000000

Operation: INCREMENT(B, k)

Cost: number of bits flipped

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procedure INCREMENT(B, k)

$i = 0$

while $i < k$ **and** $B[i] == 1$ **do**

$B[i] = 0$

$i = i + 1$

if $i < k$ **then**

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000001  ↘ 1
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} 2

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procedure $\text{INCREMENT}(B, k)$

$$T(n) \leq \sum_{i=1}^n k = nk = O(n \log n)$$

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$$\hat{C} = \frac{T(n)}{n} = 2$$

More precisely

When analyzing algorithms, we want an upper bound \hat{C} on the total cost:

$$T(n) = \sum_{i=0}^n c_i \leq \hat{C}$$

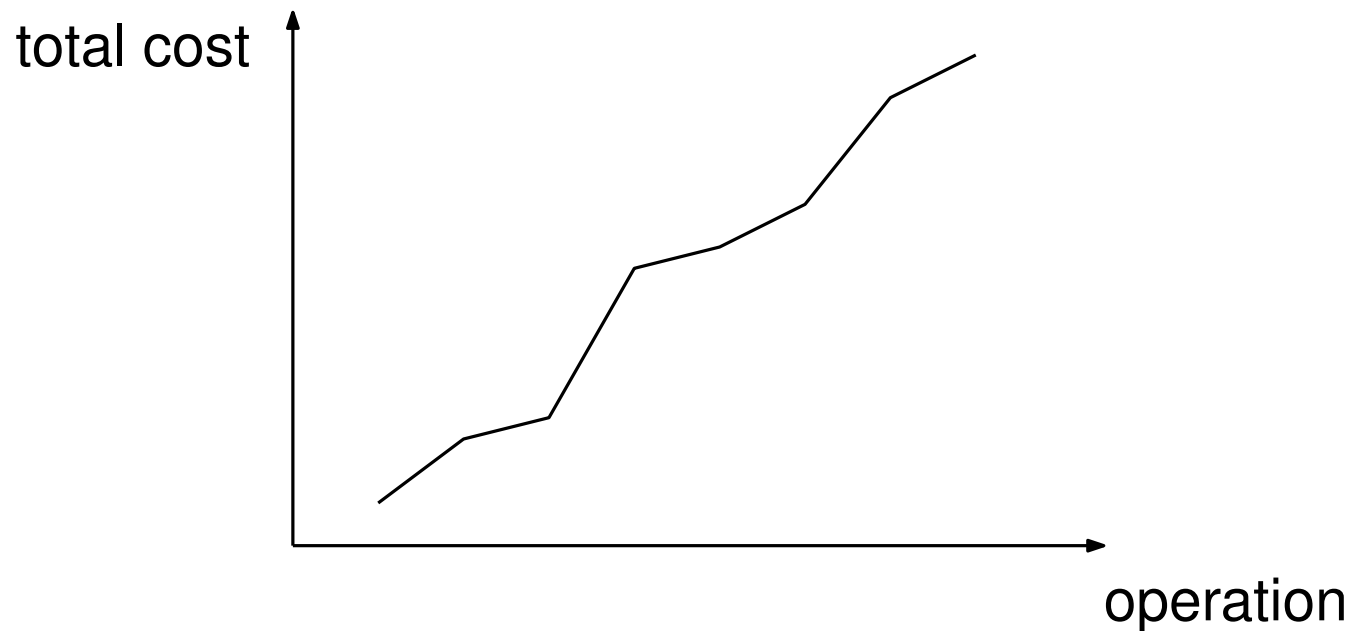
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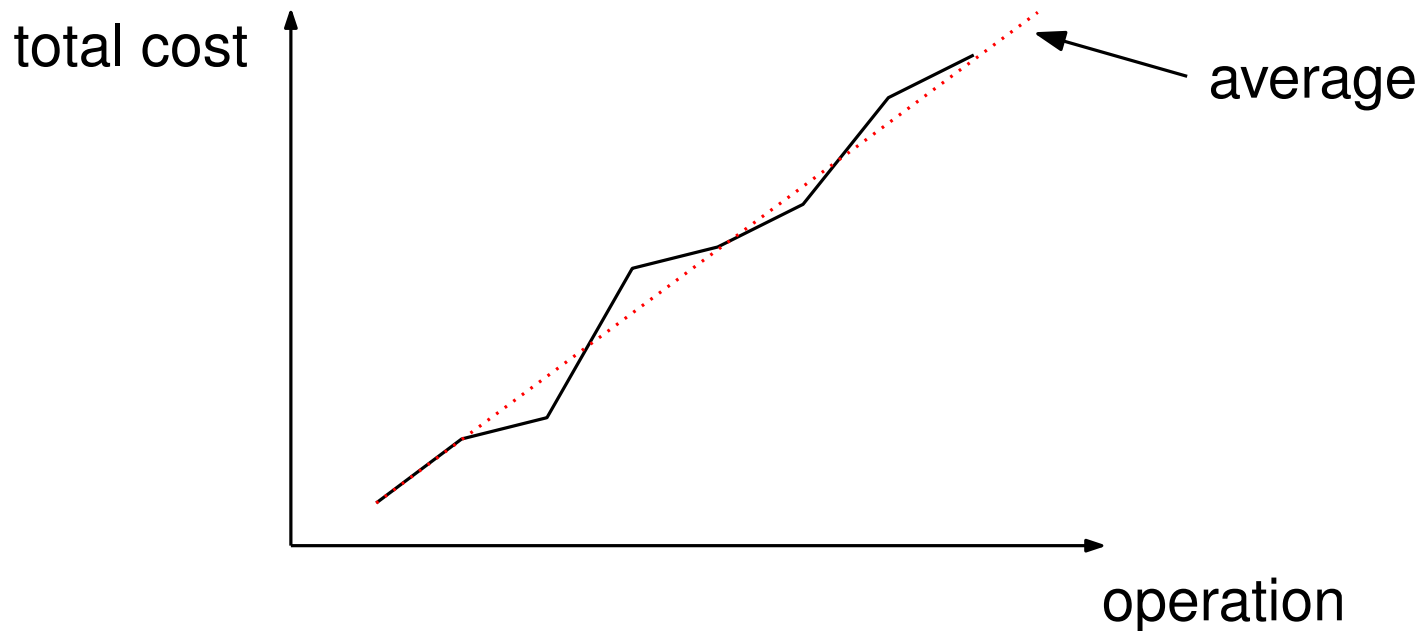


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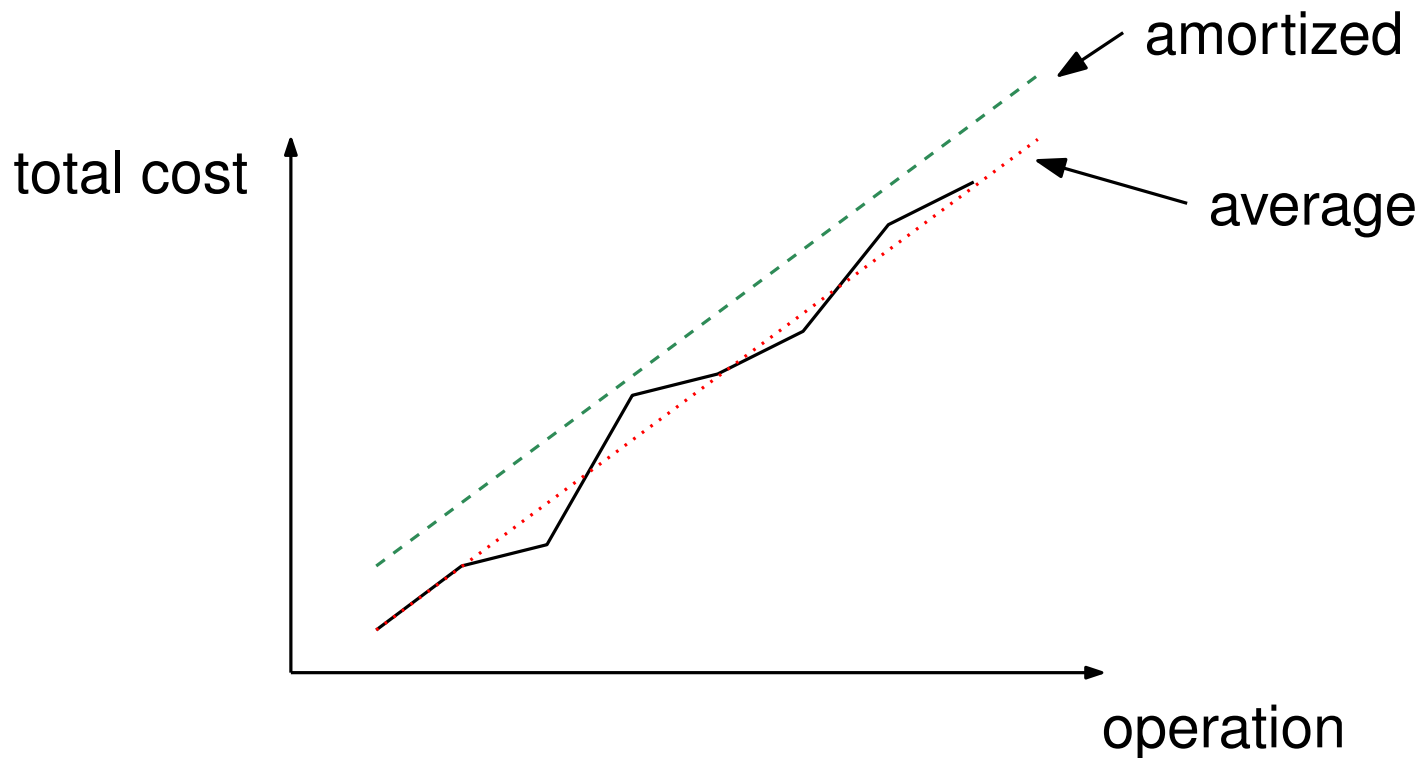


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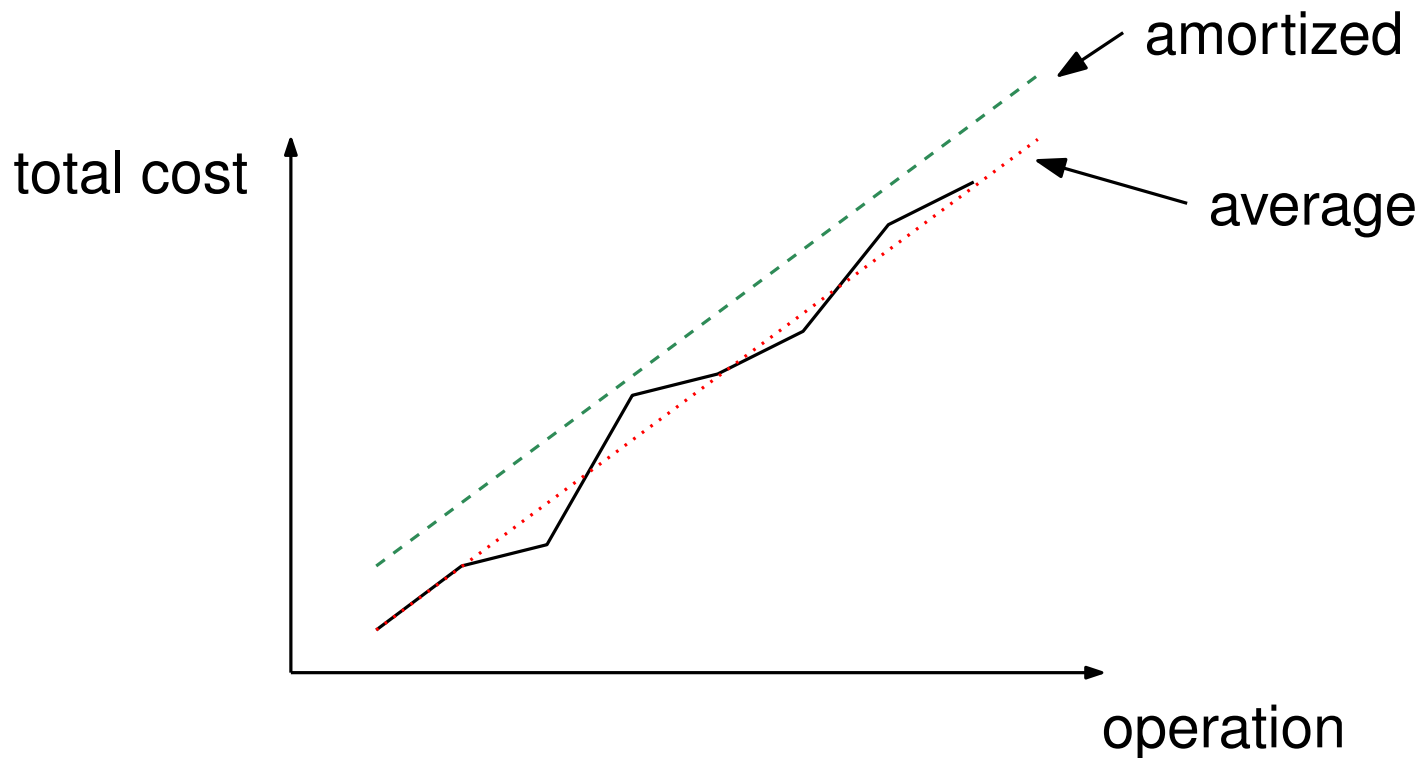


More precisely

When analyzing algorithms, we want an upper bound \hat{C} on the total cost:

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Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

"Charge" \hat{c}_i
per operation

Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

"Charge" \hat{c}_i
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Pay c_i for the
actual cost of
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Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

"Charge" \hat{c}_i
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Pay c_i for the
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Save $(\hat{c}_i - c_i)$



Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

"Charge" \hat{c}_i
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Save $(\hat{c}_i - c_i)$



Pay for more expensive
operations using savings

Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

"Charge" $\$ \hat{c}_i$
per operation

Pay $\$ c_i$ for the
actual cost of
operation

Save $\$(\hat{c}_i - c_i)$



Pay for more expensive
operations using savings

$$\sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i \quad \Rightarrow \quad \sum_{i=1}^n \hat{c}_i - \sum_{i=0}^n c_i = \sum_{i=1}^n (\hat{c}_i - c_i) \geq 0$$

Amortized analysis: Accounting method

$$T(n) = \sum_{i=0}^n c_i \leq \sum_{i=1}^n \hat{c}_i = \hat{C}$$

"Charge" \hat{c}_i
per operation

Pay c_i for the
actual cost of
operation

Save $(\hat{c}_i - c_i)$



Pay for more expensive
operations using savings

Prove that $\sum_{i=0}^n (\hat{c}_i - c_i) \geq 0$

i.e., the balance in the bank is always non-negative

Example: Binary Counter



Charge \hat{c}_i

Actual Cost c_i

000000

000001

000010

000011

000100


000101

000110

000111

001000

Example: Binary Counter

	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	 \$1
000001			
000010			
000011			
000100			
000101			
000110			
000111			
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000			
	\$2	\$1	\$1
000001			
	\$2	\$2	\$1
000010			
000011			
000100			
000101			
000110			
000111			
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	\$1
000001	\$2	\$2	\$1
000010	\$2	\$1	\$2
000011			
000100			
000101			
000110			
000111			
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	\$1
000001	\$2	\$2	\$1
000010	\$2	\$1	\$2
000011	\$2	\$3	\$1
000100			
000101			
000110			
000111			
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	\$1
000001	\$2	\$2	\$1
000010	\$2	\$1	\$2
000011	\$2	\$3	\$1
000100	\$2	\$1	\$2
000101			
000110			
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Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
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000100	\$2	\$1	\$2
000101	\$2	\$2	\$2
000110			
000111			
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
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Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	\$1
000001	\$2	\$2	\$1
000010	\$2	\$1	\$2
000011	\$2	\$3	\$1
000100	\$2	\$1	\$2
000101	\$2	\$2	\$2
000110	\$2	\$1	\$3
000111	\$2	\$4	\$1
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	\$1
000001	\$2	\$2	\$1
000010	\$2	\$1	\$2
000011	\$2	\$3	\$1
000100	\$2	\$1	\$2
000101	\$2	\$2	\$2
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000111	\$2	\$4	\$1
001000			

Example: Binary Counter



	Charge \hat{c}_i	Actual Cost c_i	
000000	\$2	\$1	\$1
000001	\$2	\$2	\$1
000010	\$2	\$1	\$2
000011	\$2	<div style="background-color: yellow; border: 1px solid black; padding: 5px; text-align: center;"> Keep savings on the data structure </div>	\$1
000100	\$2		\$2
000101	\$2	\$2	\$2
000110	\$2	\$1	\$3
000111	\$2	\$4	\$1
001000			

Amortized analysis: Potential method

Potential function:

$$\Phi : D \rightarrow \mathbb{R}^+ \cup 0$$

data structure

number

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E.g.: $\Phi(D_i) = \text{number of 1's in } D_i$

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Potential function:

$$\Phi : D \rightarrow \mathbb{R}^+ \cup 0$$

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E.g.: $\Phi(D_i) = \text{number of 1's in } D_i$

Amortized cost:

$$\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1}) = c_i + \Delta\Phi_i$$

Potential method: justification

$$\sum_{i=1}^n \hat{c}_i = \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1}))$$

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$$\sum_{i=1}^n \hat{c}_i = \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1}))$$

$$= \sum_{i=1}^n c_i + \Phi(D_1) - \Phi(D_0) + \Phi(D_2) - \Phi(D_1) + \cdots + \Phi(D_{n-1}) - \Phi(D_{n-2}) + \\ \Phi(D_n) - \Phi(D_{n-1})$$

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

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

$$\Rightarrow \sum_{i=1}^n \hat{c}_i - \sum_{i=1}^n c_i = \Phi(D_n) - \Phi(D_0) \geq 0$$

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

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Satisfied if:

- $\Phi(D_0) = 0$, and
- $\Phi(D_n) \geq 0$ for all $n > 0$

Potential method: justification

Potential function:

$$\Phi : D \rightarrow \mathbb{R}^+ \cup 0$$

$$\sum_{i=1}^n \hat{c}_i = \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1}))$$

$$= \sum_{i=1}^n c_i + \cancel{\Phi(D_1)} - \Phi(D_0) + \cancel{\Phi(D_2)} - \cancel{\Phi(D_1)} + \cdots + \cancel{\Phi(D_{n-1})} - \cancel{\Phi(D_{n-2})} + \Phi(D_n) - \cancel{\Phi(D_{n-1})}$$

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

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Example: Binary counter

$\Phi(D_i) = \#$ of 1's in the counter after the i -th INCREMENT(B, k)

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$\Phi(D_i) = \#$ of 1's in the counter after the i -th INCREMENT(B, k)

$$\hat{c}_i = c_i + \Delta\Phi_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$$

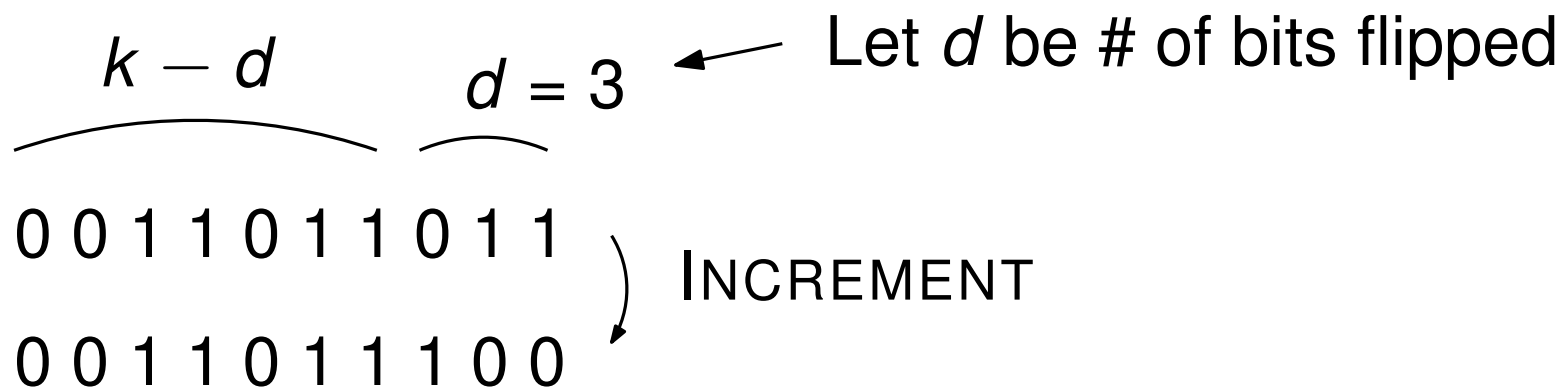
0 0 1 1 0 1 1 0 1 1
0 0 1 1 0 1 1 1 0 0

) INCREMENT
↓

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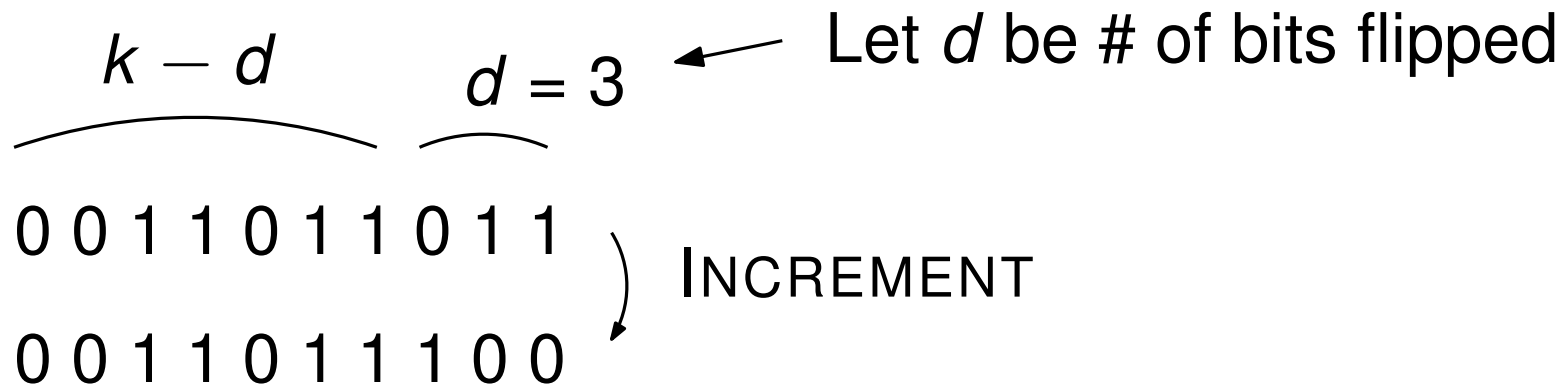
$$d' = 4$$

Let d' be # of 1's in $k - d$ MSBs

Example: Binary counter

$\Phi(D_i) = \#$ of 1's in the counter after the i -th INCREMENT(B, k)

$$\hat{c}_i = c_i + \Delta\Phi_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$$



$$\hat{c}_i = d + \Phi(D_i) - \Phi(D_{i-1})$$

$d' = 4$

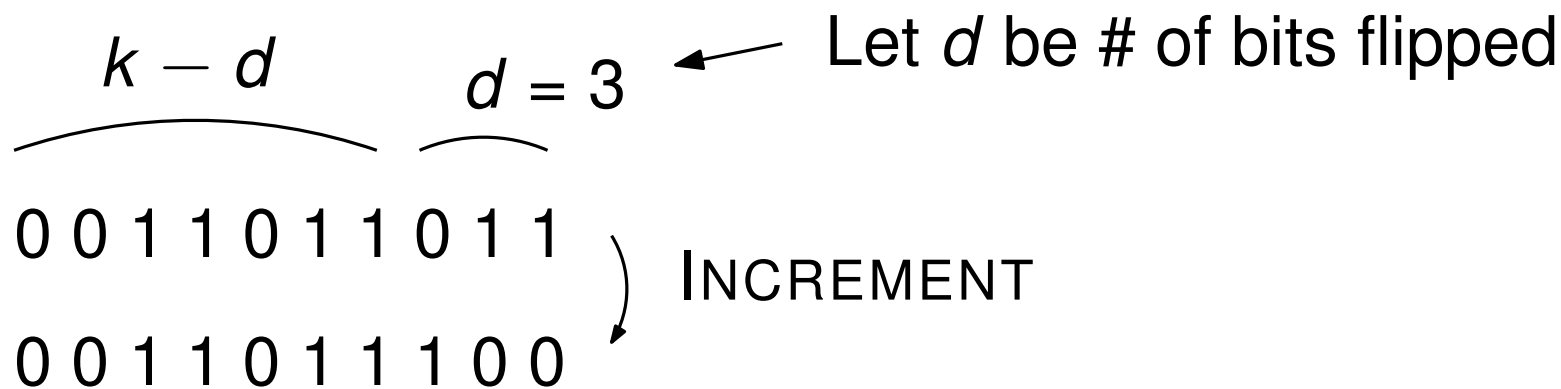
↑

Let d' be # of 1's in $k - d$ MSBs

Example: Binary counter

$\Phi(D_i) = \# \text{ of } 1\text{'s in the counter after the } i\text{-th INCREMENT}(B, k)$

$$\hat{c}_i = c_i + \Delta\Phi_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$$



$d' = 4$

↑

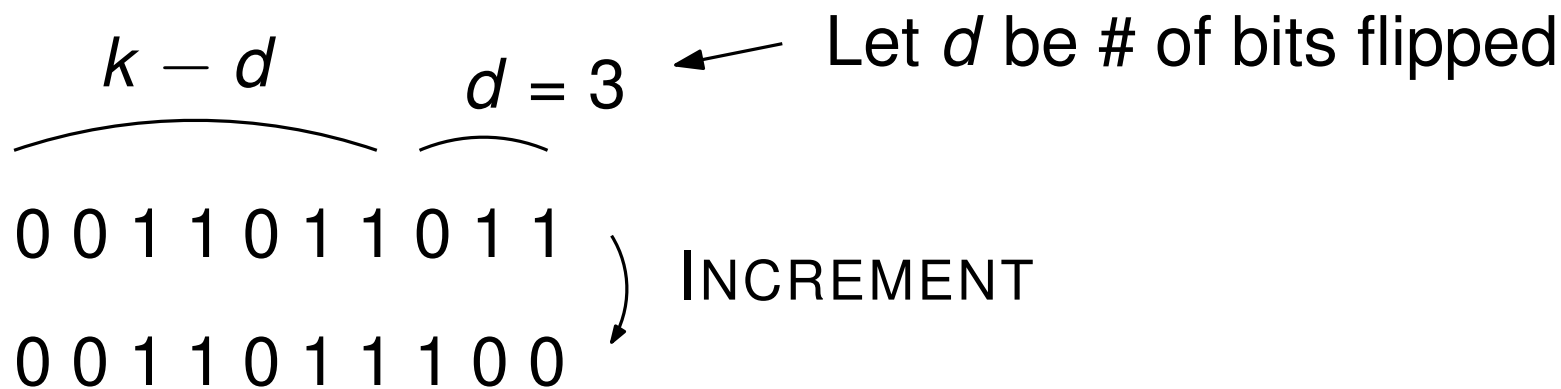
$$\begin{aligned} \hat{c}_i &= d + \Phi(D_i) - \Phi(D_{i-1}) \\ &= d + (d' + 1) - (d' + (d - 1)) \end{aligned}$$

Let d' be # of 1's in $k - d$ MSBs

Example: Binary counter

$\Phi(D_i) = \#$ of 1's in the counter after the i -th INCREMENT(B, k)

$$\hat{c}_i = c_i + \Delta\Phi_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$$



$d' = 4$

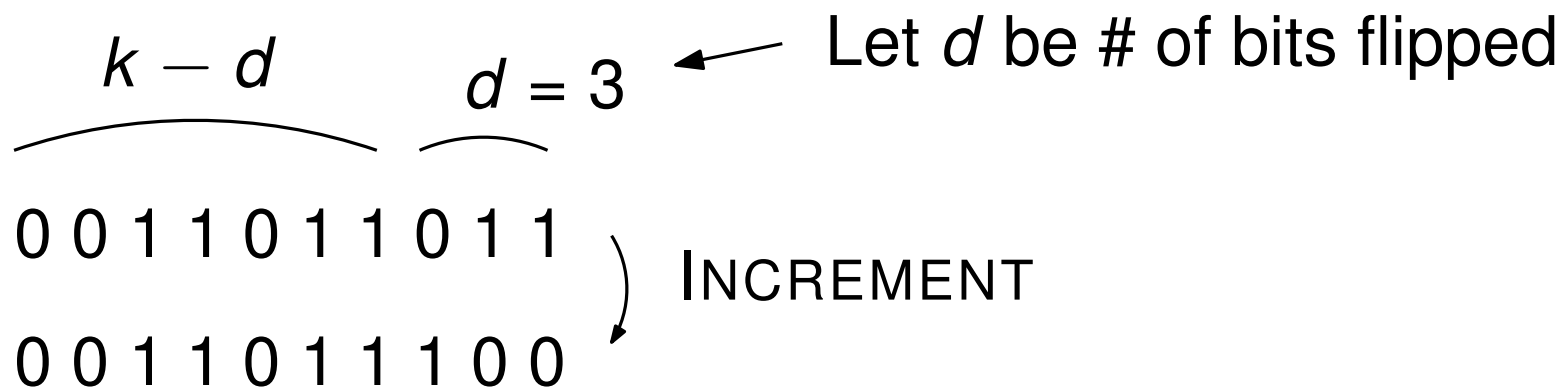
$$\begin{aligned} \hat{c}_i &= d + \Phi(D_i) - \Phi(D_{i-1}) \\ &= d + \cancel{(d'+1)} - \cancel{(d'+(d-1))} \end{aligned}$$

Let d' be # of 1's in $k - d$ MSBs

Example: Binary counter

$\Phi(D_i) = \#$ of 1's in the counter after the i -th INCREMENT(B, k)

$$\hat{c}_i = c_i + \Delta\Phi_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$$



$d' = 4$

Let d' be # of 1's in $k - d$ MSBs

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

Amortized analysis: Summary

- Aggregate method
- Accounting method
- Potential method

$$\hat{c} = \frac{1}{n} \cdot \sum_{i=1}^n c_i$$



Potential function:

- $\Phi : D \rightarrow \mathbb{R}^+ \cup \{0\}$
- $\Phi(D_0) = 0$, and
- $\Phi(D_n) \geq 0$ for all $n > 0$

Keep savings
on the data structure