

$$T(n) = T(n-1) + T(n-2) + 1 \quad || \quad F_n = F_{n-1} + F_{n-2}$$

? HINT

both upper bounded and lower bound

$$T(n) \approx F_{n-1} + 2F_n + F_{n+1} - 1$$

ind step

~~ind hyp~~

$$T(n) \approx F_{n-2} + 2F_{n-1} + F_n - 1$$

$$T(n) \approx F_{n-1} + 2F_n + F_{n+1} - 1$$

~~concl~~

$$T(n+1) = T(n) + T(n-1)$$

$$+ 1$$

$$\approx F_n + 2F_{n+1} + F_{n+2} - 1$$

$$= F_n + 2F_{n+1} + F_{n+2} - 1$$

Amortized Analysis

Fibonacci Heaps

thanks MIT slides

thanks “Amortized Analysis” by Rebecca Fiebrink

thanks Jay Aslam’s notes

Objectives

- Amortized Analysis
 - potential method
- Fibonacci Heaps
 - construction
 - operations

running time analysis

- typical: Algorithm uses data-structure and operations
 - structures: table, array, hash, heap, list, stack
 - operations: insert, delete, search, min, max, push, pop
- measure running time by analyzing
 - the sequence of operations,
 - their frequency
 - each operation running time (computation cost)

Running Time Analysis

- determine the c = costliest/longest iteration
 - usually an outer loop of n iterations
 - overall n^* (longest cost per iteration) = n^*c
- That's not very accurate!
 - not all iterations have the longest cost
 - perhaps some average technique can work, but how to prove?
- “compensate” : show that for every costly iteration, there must be other “cheap” iterations

Example : binary counter

log n bits

bit 5	bit 4	bit 3	bit 2	bit 1	bit 0
0	0	0	0	0	0
0	0	0	0	0	1
0	0	0	0	1	0
0	0	0	0	1	1
0	0	0	1	0	0
0	0	0	1	0	1
0	0	0	1	1	0
0	0	0	1	1	1
0	0	1	0	0	0

cost (bits changed)
N/A
1
2
1
3
1
2
1
4

n ops (+1)
 naive R.T
 (worst case)
 each op $\leq \log n$
 total
 $O(n \cdot \log n)$

● each row is a binary representation of the counter

- increasing by one
- right side: cost = how many bits require changes

$1 + 2 + 1 + 3 + 1 + 2 + 1 + 4 + \dots$
 $= n + \frac{n}{2} + \frac{n}{4} + \frac{n}{8} + \dots$
 $n(1 + \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots)$

● naive running time to increment from 0 to n :

- each row may cost up to $O(\log n)$
- n iterations/increments would be $O(n \cdot \log n)$

$\geq 2n = \Theta(n)$

Example : binary counter

bit 5	bit 4	bit 3	bit 2	bit 1	bit 0	cost (bits changed)
0	0	0	0	0	0	N/A
0	0	0	0	0	1	1
0	0	0	0	1	0	2
0	0	0	0	1	1	1
0	0	0	1	0	0	3
0	0	0	1	0	1	1
0	0	0	1	1	0	2
0	0	0	1	1	1	1
0	0	1	0	0	0	4

→ true cost

- true cost for n iterations: $1+2+1+3+1+2+1+4+\dots = 2n = O(n)$
- reason: the iteration cost very rarely is $O(\log n)$
 - $O(\log n)$ means changing a good number of bits
 - for one iteration of cost $O(\log n)$, there must be many "cheap" iterations

binary counter amortization

- Aggregation method: consider all n iterations
 - bit 0 changes every iteration \Rightarrow cost n
 - bit 1 changes for half of iterations \Rightarrow cost $n/2$
 - bit 2 changes quarter of iterations \Rightarrow cost $n/4$
 - bit 3 changes $1/8$ of iterations \Rightarrow cost $n/8$
 - ... etc

bit 5	bit 4	bit 3	bit 2	bit 1	bit 0
0	0	0	0	0	0
0	0	0	0	0	1
0	0	0	0	1	0
0	0	0	0	1	1
0	0	0	1	0	0
0	0	0	1	0	1
0	0	0	1	1	0
0	0	0	1	1	1
0	0	1	0	0	0

- total cost : add up the cost per bit

- $n + n/2 + n/4 + n/8 + \dots = 2n$

- Aggregation method impractical, only works on toy examples like this

for pedagogy only.

Amortized Analysis

- c_i = true cost of i -th operation/iteration
- \hat{c}_i = amortized cost of i -th operation/iteration
 - we have to come up with d_i
- the cumulative amortized cost can't be smaller than the true cumulative cost, up to any iteration k

$$\forall k : \sum_{i=1:k} c_i \leq \sum_{i=1:k} \hat{c}_i$$

Accounting Method

- assign the di amortized cost
- if overcharge some operation ($d_i > c_i$) use the excess as "prepaid credit",
- use the prepaid credit later for an expensive operation

Potential method

- associate a potential function Φ with datastructure T
 - $\Phi(T_i)$ = “potential” (or risk for cost) associated with datastructure after i -th operation
 - typically a measure of complexity/risk/size of the datastructure
- require $\hat{c}_i \geq c_i + \phi(T_i) - \phi(T_{i-1})$ for all i
- \hat{c}_i = amortized cost (up to us to define)
- c_i = true cost for operation i
- Φ = potential function
- T_i = datastructure after i th operation

Accounting Method for binary counter

$$\hat{c}_i = c_i + \Delta\phi$$

bit 5	bit 4	bit 3	bit 2	bit 1	bit 0	true cost (c_i)	amortized cost (\hat{c}_i)
0	0	0	0	0	0	N/A	N/A
0	0	0	0	0	1	1	2
0	0	0	0	1	0	2	2
0	0	0	0	1	1	1	2
0	0	0	1	0	0	3	2
0	0	0	1	0	1	1	2
0	0	0	1	1	0	2	2
0	0	0	1	1	1	1	2
0	0	1	0	0	0	4	2

savings $\sum_{i=1:k} c_i - \sum_{i=1:k} \hat{c}_i = \phi(\tau_i) - \phi(\tau_{i-1})$

Δ potential

+1
0
+1
-1
+1
0
+1
-2

cumulative cost up to $k \leq$ amortized cost up to k

- assign amortized cost of $d_i=2$ for each operation
- verify the amortized condition

$$\forall k : \sum_{i=1:k} c_i \leq \sum_{i=1:k} \hat{c}_i$$

Accounting Method for binary counter

bit 5	bit 4	bit 3	bit 2	bit 1	bit 0	true cost (c_i)	amortized cost \hat{c}_i	cum true cost $\sum c_i$	cum amortized cost $\sum_{i=1:k} \hat{c}_i$
0	0	0	0	0	0	N/A	N/A	N/A	N/A
0	0	0	0	0	1	1	2	1	2
0	0	0	0	1	0	2	2	3	4
0	0	0	0	1	1	1	2	4	6
0	0	0	1	0	0	3	2	7	8
0	0	0	1	0	1	1	2	8	10
0	0	0	1	1	0	2	2	10	12
0	0	0	1	1	1	1	2	11	14
0	0	1	0	0	0	4	2	15	16

- assign amortized cost of $d_i=2$ for each operation *constraint*
- verify the amortized condition

$$\forall k : \sum_{i=1:k} c_i \leq \sum_{i=1:k} \hat{c}_i$$

Potential method for binary count

- define the potential $\Phi(T_i) =$ the number of "1" bits

$i=1:k$ $T_i =$ binary counter after i iterations

- verify $\hat{c}_i \geq c_i + \phi(T_i) - \phi(T_{i-1})$ for each operation

telescope

- there is only one operation: "increment"

$\sum_{i=1}^k$

- $d_i=2$, amortized cost defined by us

$\sum_{i=1}^k \hat{c}_i \geq \sum_{i=1}^k c_i + \sum_{i=1}^k (\phi(T_i) - \phi(T_{i-1}))$

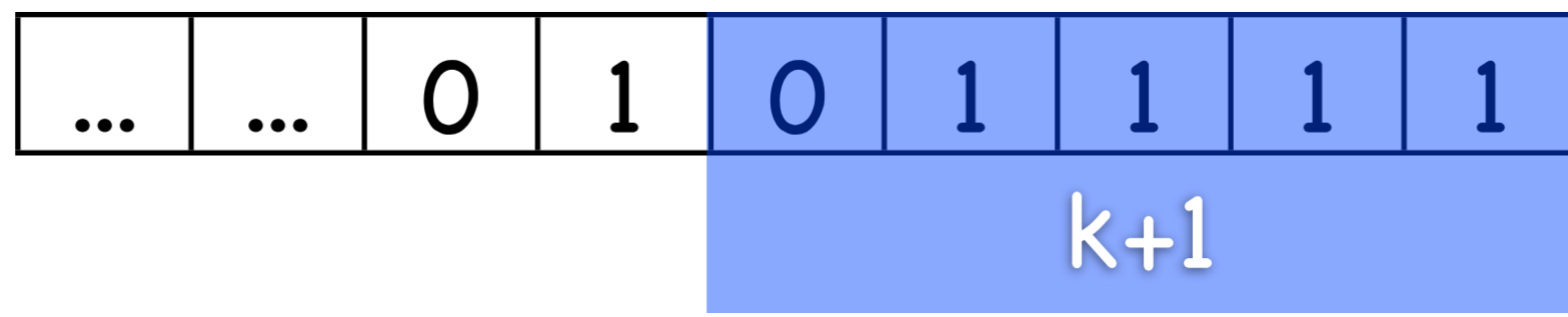
- before the operation i , at T_{i-1} , say there are k trailing 1 ones, before i -th increment

- $c_i =$ true cost = $k+1$ bit changes: k of "1" bits made "0" (from right to left up to the first "0"); plus the first "0" made "1"

$\phi(T_k) - \phi(T_0) \geq 0$

- $\Phi(T_i) - \Phi(T_{i-1}) =$ "1" gained - "1" lost = $1-k$

- equation becomes $2 \geq k+1 + 1-k$, it checks out! $d_i = 2$ is good



Stack operations - review

- stack is an array with LAST-IN-FIRST-OUT operations
- push(value): put the new value on the stack (at the top)
return
- pop(n): take the top n values, return the, delete them from stack
(or max stack if $\leq n$)
- naive analysis for n operations : $n * O(n) = O(n^2)$
- better: for pop() to extract many elements, many push() must have happened before, each push is $O(1)$

	z			
c	c		d	
b	b	b	b	b
a	a	a	a	a
	push(z)	pop(2)	push(d)	pop(1)

Accounting method for Stack

- account each $\text{push}(x)$ with \$2:
 - \$1 for the actual $\text{push}(x)$ operation, to add x to the stack
 - \$1 credit for the possible later $\text{pop}()$ operation that extracts x
- each $\text{pop}(k)$ also \$2, for any k
- so each operation is accounted with \$2,
- total running time for n operations is $2 \cdot n = O(n)$
- when $\text{pop}(k)$ is called, each one of the popped elements have stored \$1 to account for their extraction, $O(k)$ time

Potential method for Stack

- define the potential $\phi(\text{stack}) = \text{size}(\text{stack})$

design for stacks

– $\phi(T) = |T|$; $T = \text{the stack}$; $T_i = \text{stack after } i \text{ operations}$

- define the amortized costs: $d_{\text{push}}=2$; $d_{\text{pop}}=2$

- consider the true costs $c_{\text{push}}=1$; $c_{\text{pop}(k)}=k$

- prove that for each operation the potential satisfies the fundamental property (exercise)

$$\hat{c}_i \geq c_i + \phi(T_i) - \phi(T_{i-1})$$

universal

- which means the amortized cost $d=2$ is valid.

push:

$$2 \geq 1$$

$$+ 1$$

$$+ \Delta\phi$$

$$\text{pop}(k) \wedge c_i \geq k$$

$$\geq k$$

$$- k \Delta\phi$$

Amortized Analysis

Move to Front

Self-organizing lists

- ▶ List L of n elements
- ▶ The operation $\text{ACCESS}(x)$ costs

$\text{rank}_L(x)$ = distance of x from the head of L .

- ▶ L can be reordered by swapping adjacent elements at a cost of 1
choice: move accessed elem to front
- ▶ Goal: access to a sequence of n items with minimal cost

List access algorithms

- ▶ Off-line Algorithm: if the sequence of access S is known in advance, one can design an optimal algorithm to rearrange the list based on how often items are accessed
- ▶ On-line Algorithm: if the sequence is not known in advance, one can design an algorithm based on some heuristics.

Move-to-front algorithm

- ▶ Algorithm: After accessing x , move x to the head of L using swaps.

$$\text{cost} = 2 \cdot \text{rank}_L(x)$$

access
rank(x)

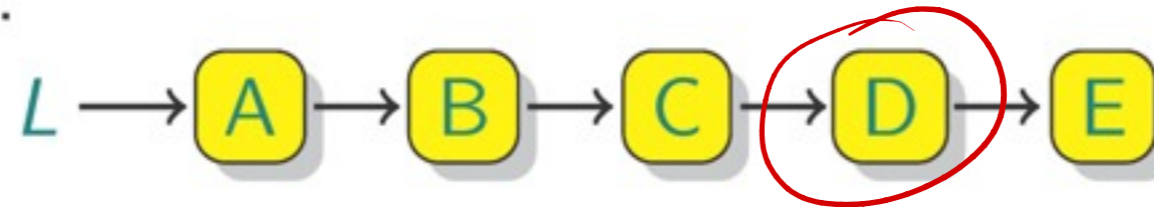
move to front
rank(x)

Move-to-front algorithm

- ▶ Algorithm: After accessing x , move x to the head of L using swaps.

$$\text{cost} = 2 \cdot \text{rank}_L(x)$$

- ▶ Access item D:

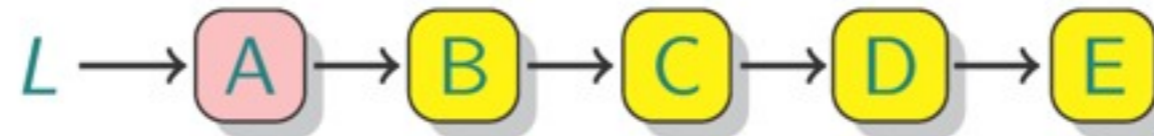


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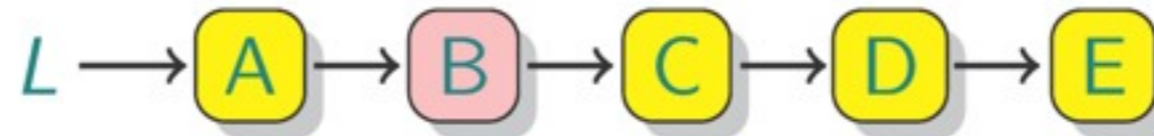


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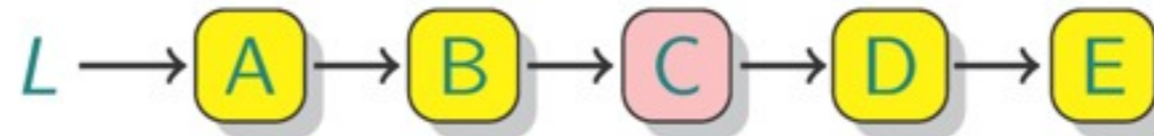


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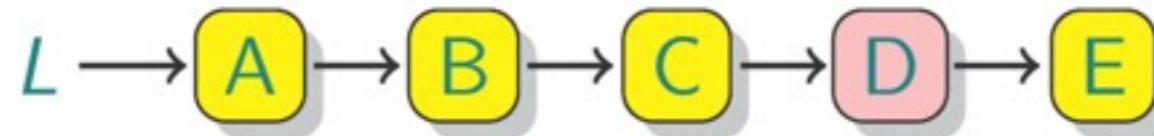


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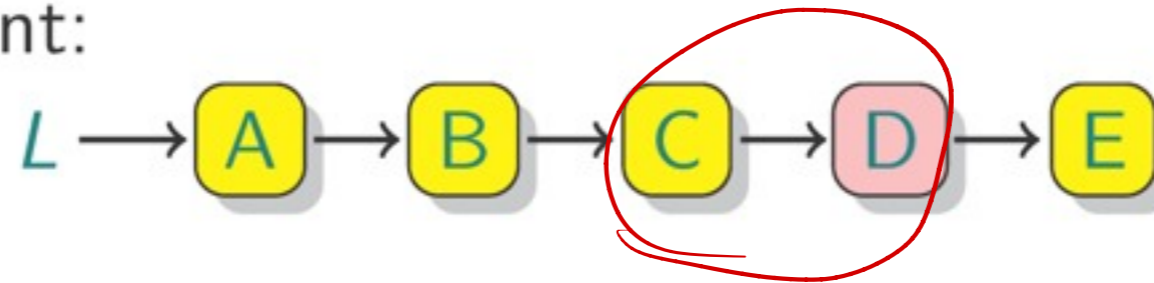


Move-to-front algorithm

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- ▶ Move D to front:

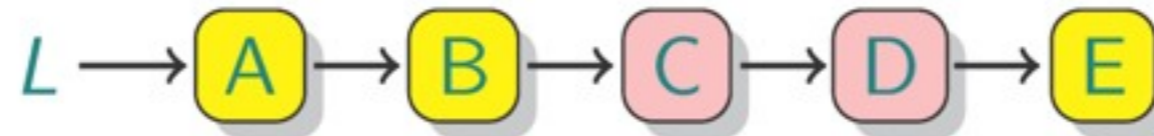


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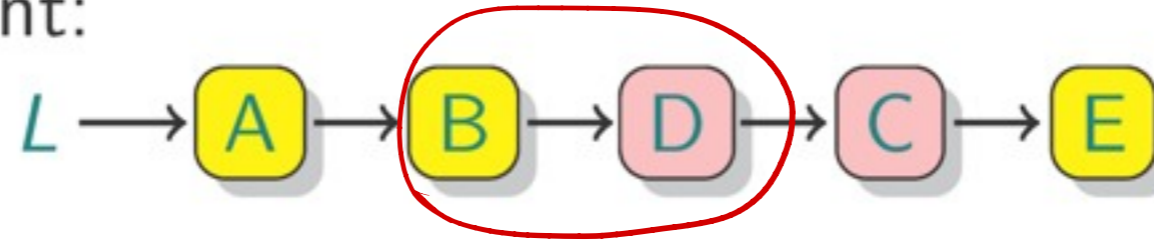


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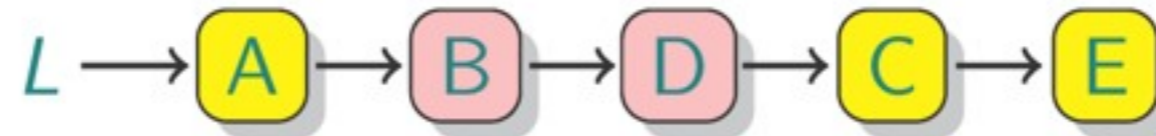


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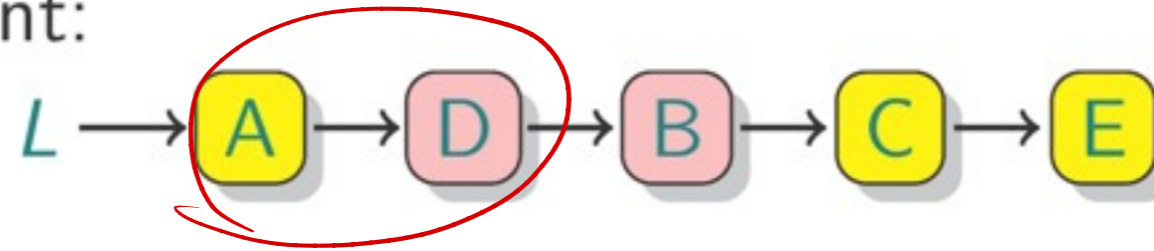


Move-to-front algorithm

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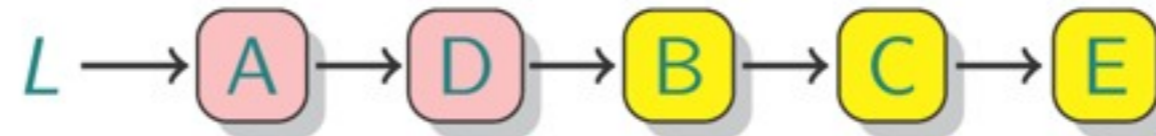


Move-to-front algorithm

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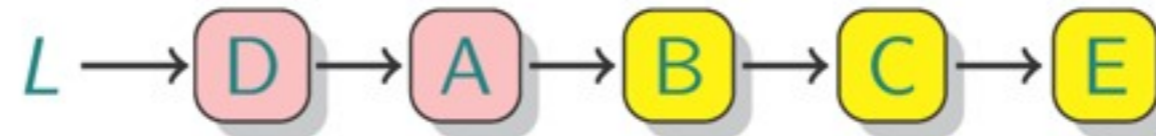


Move-to-front algorithm

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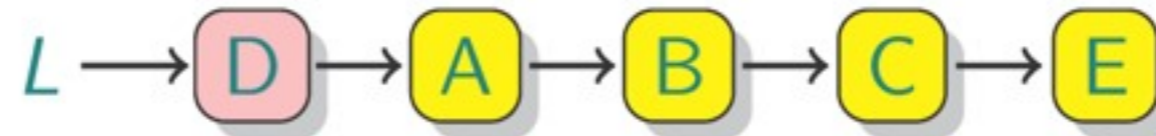


Move-to-front algorithm

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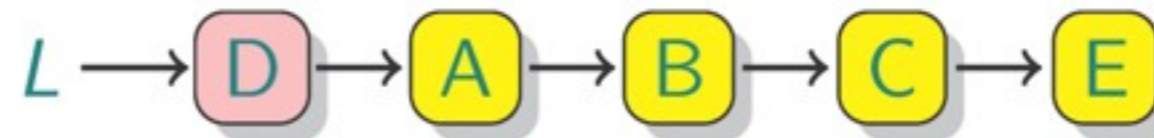


Move-to-front algorithm

- ▶ Algorithm: After accessing x , move x to the head of L using swaps.

$$\text{cost} = 2 \cdot \text{rank}_L(x)$$

- ▶ Move D to front:



- ▶ Heuristic: if x is accessed at time t , it is likely to be accessed again soon after time t .
- ▶ Cost: MTF always performs within a factor of 4 of the optimal algorithm.

Amortized analysis of MTF

Theorem: $C_{MTF}(S) \leq 4C_{OPT}(S)$

Proof: Let L_i be MTF's list after the i th access, and let L_i^* be OPT's list after the i th access. Let

$$c_i = \text{MTF's cost for the } i\text{th operation} \\ = 2 \cdot \text{rank}_{L_{i-1}}(x) \text{ if it accesses } x;$$

$$c_i^* = \text{OPT's cost for the } i\text{th operation} \\ = \text{rank}_{L_{i-1}^*}(x) + t_i,$$

lost → *after-access swaps*

where t_i is the number of swaps that OPT performs.

Potential function

Inversion $i < j$
 $A[i] > A[j]$

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\Phi(L_i) = 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}|$$

$$= 2 \cdot \# \text{ inversions}$$

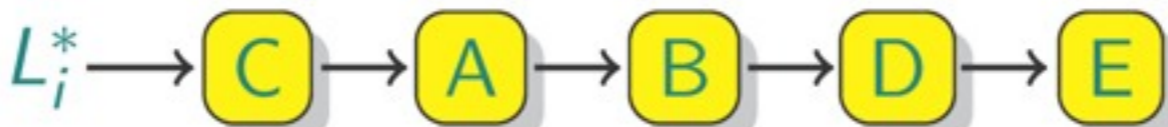
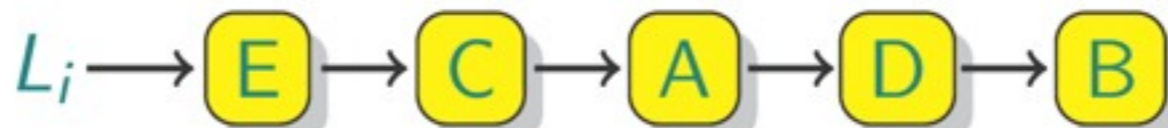
between

L_i → current MTF list

L_i^* → current OPT list

Example:

after i operations



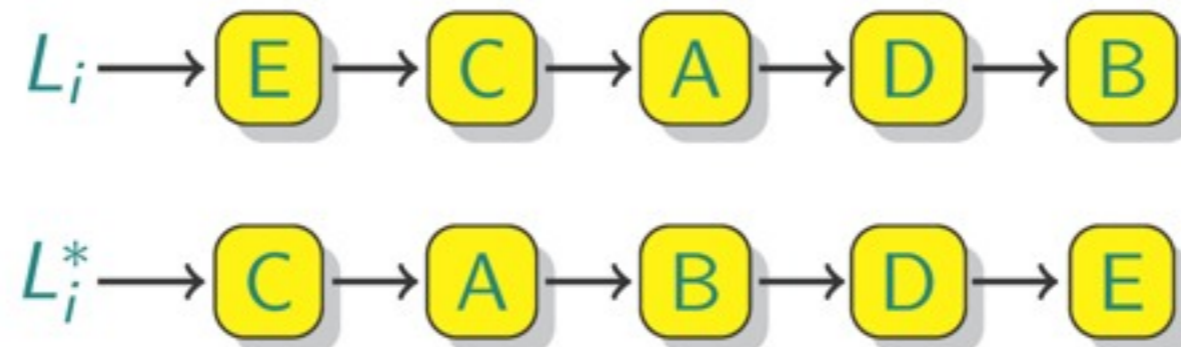
EC, EA, ED, EB, DB

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\begin{aligned}\Phi(L_i) &= 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}| \\ &= 2 \cdot \# \text{ inversions}\end{aligned}$$

Example:



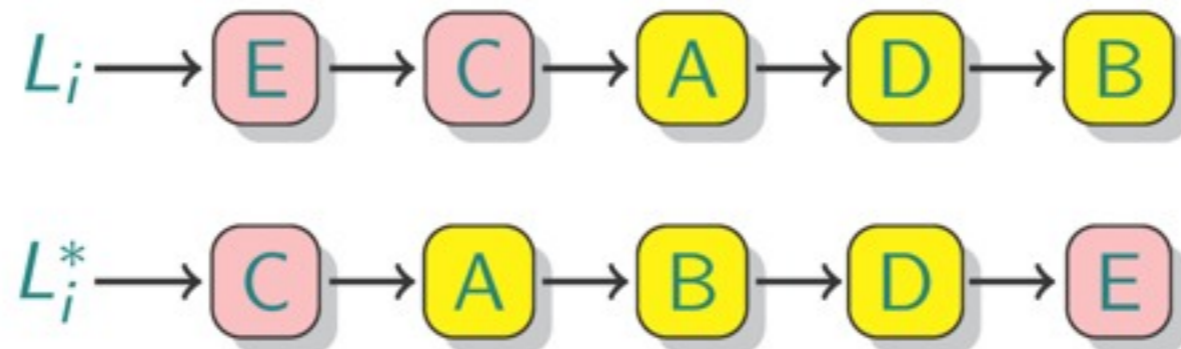
$$\Phi(L_i) = 2 \cdot |\{\dots\}|$$

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\begin{aligned}\Phi(L_i) &= 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}| \\ &= 2 \cdot \# \text{ inversions}\end{aligned}$$

Example:



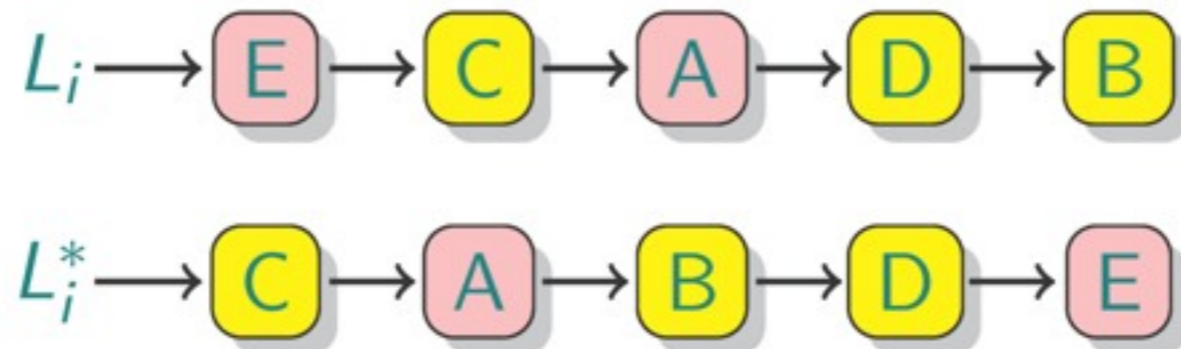
$$\Phi(L_i) = 2 \cdot |\{(E, C), \dots\}|$$

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\begin{aligned}\Phi(L_i) &= 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}| \\ &= 2 \cdot \# \text{ inversions}\end{aligned}$$

Example:



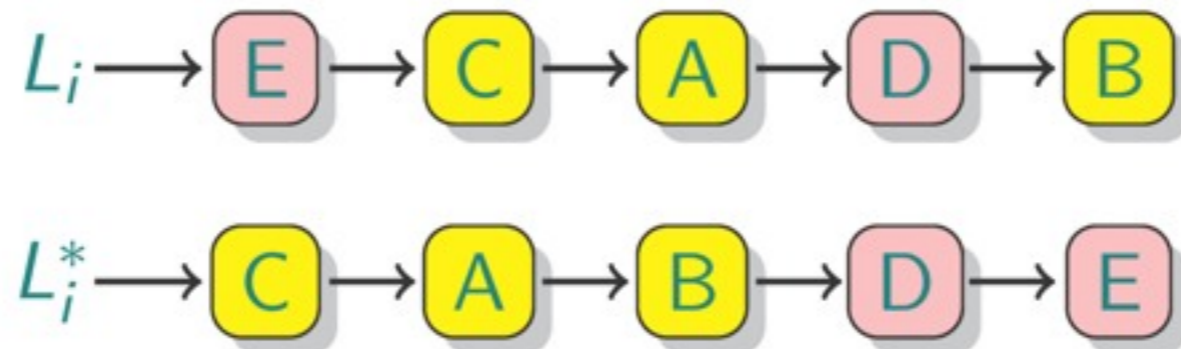
$$\Phi(L_i) = 2 \cdot |\{(E, C), (E, A), \dots\}|$$

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\begin{aligned}\Phi(L_i) &= 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}| \\ &= 2 \cdot \# \text{ inversions}\end{aligned}$$

Example:



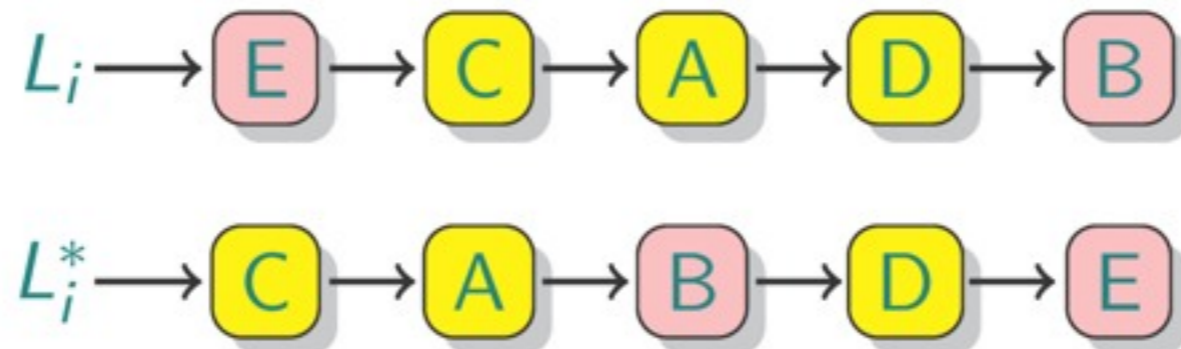
$$\Phi(L_i) = 2 \cdot |\{(E, C), (E, A), (E, D), \dots\}|$$

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\begin{aligned}\Phi(L_i) &= 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}| \\ &= 2 \cdot \# \text{ inversions}\end{aligned}$$

Example:



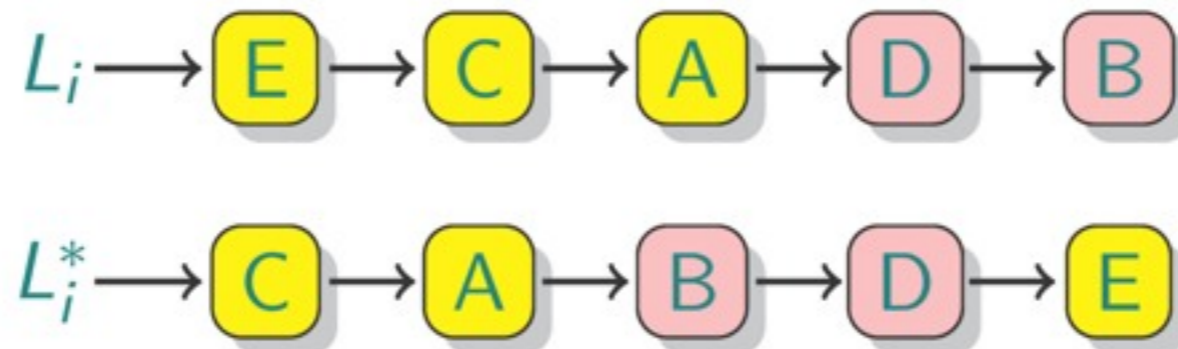
$$\Phi(L_i) = 2 \cdot |\{(E, C), (E, A), (E, D), (E, B), \dots\}|$$

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Example:



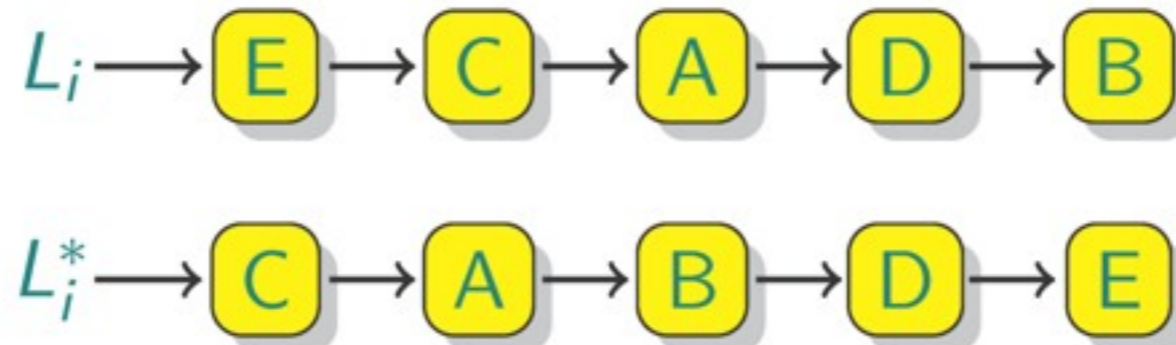
$$\Phi(L_i) = 2 \cdot |\{(\underline{E, C}), (\underline{E, A}), (\underline{E, D}), (\underline{E, B}), (\underline{D, B})\}| = 10$$

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

$$\begin{aligned}\Phi(L_i) &= 2 \cdot |\{(x, y) : x \prec_{L_i} y \text{ and } y \prec_{L_i^*} x\}| \\ &= 2 \cdot \# \text{ inversions}\end{aligned}$$

Example:



$$\Phi(L_i) = 2 \cdot |\{(E, C), (E, A), (E, D), (E, B), (D, B)\}| = 10$$

Potential function

Define the potential function $\Phi : L_i \rightarrow \mathcal{R}$ by

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Note that:

- ▶ $\Phi(L_i) \geq 0$ for $i = 0, 1, \dots$
- ▶ $\Phi(L_0) = 0$ if MTF and OPT start with the same list.

How much does Φ change from one swap?

- ▶ a swap creates/destroys 1 inversion
- ▶ $\Delta\Phi = \pm 2$

What happens on access?

Suppose that operation i access item x , and define

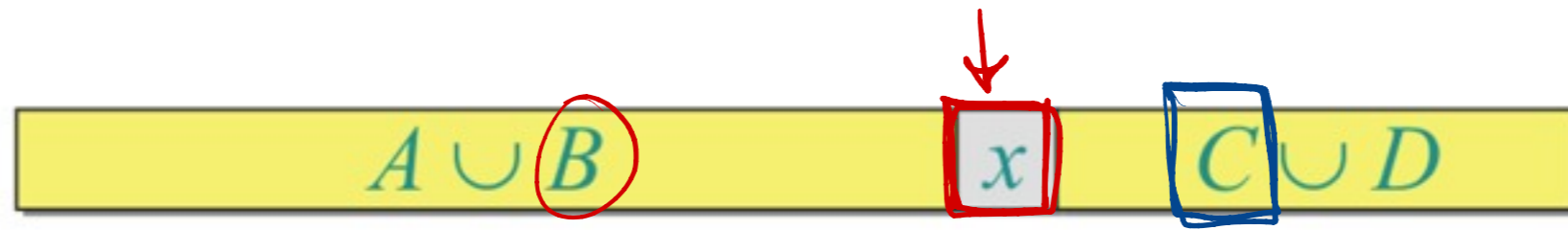
$$A = \{y \in L_{i-1} : y \prec_{L_{i-1}} x \text{ and } y \prec_{L_{i-1}^*} x\}, = \left. \begin{array}{l} \text{elem in front } (x) \\ \text{in both lists} \end{array} \right\}$$

$$B = \{y \in L_{i-1} : y \prec_{L_{i-1}} x \text{ and } y \succ_{L_{i-1}^*} x\}, = \left. \begin{array}{l} \text{elem in front } (x) \\ \text{in MTR} \end{array} \right\}$$

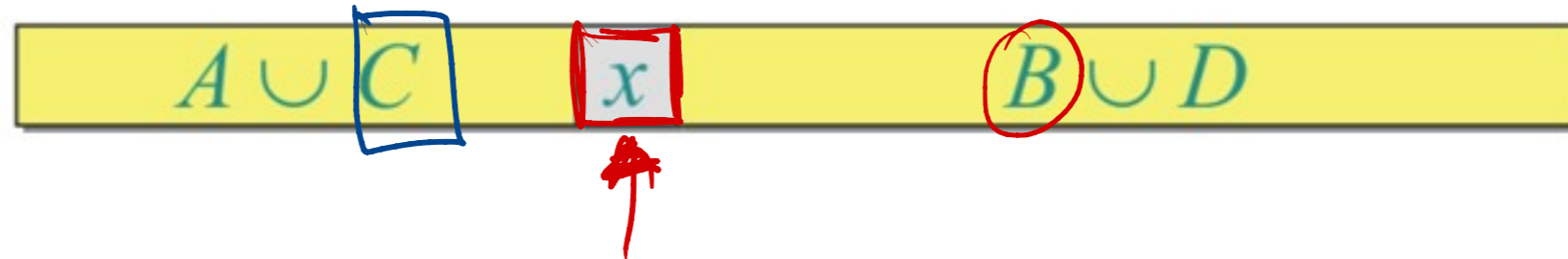
$$C = \{y \in L_{i-1} : y \succ_{L_{i-1}} x \text{ and } y \prec_{L_{i-1}^*} x\}, \quad \text{after } (x) \text{ in OPT}$$

$$D = \{y \in L_{i-1} : y \succ_{L_{i-1}} x \text{ and } y \succ_{L_{i-1}^*} x\}, = \left. \begin{array}{l} \text{elem after } (x) \\ \text{in both lists} \end{array} \right\}$$

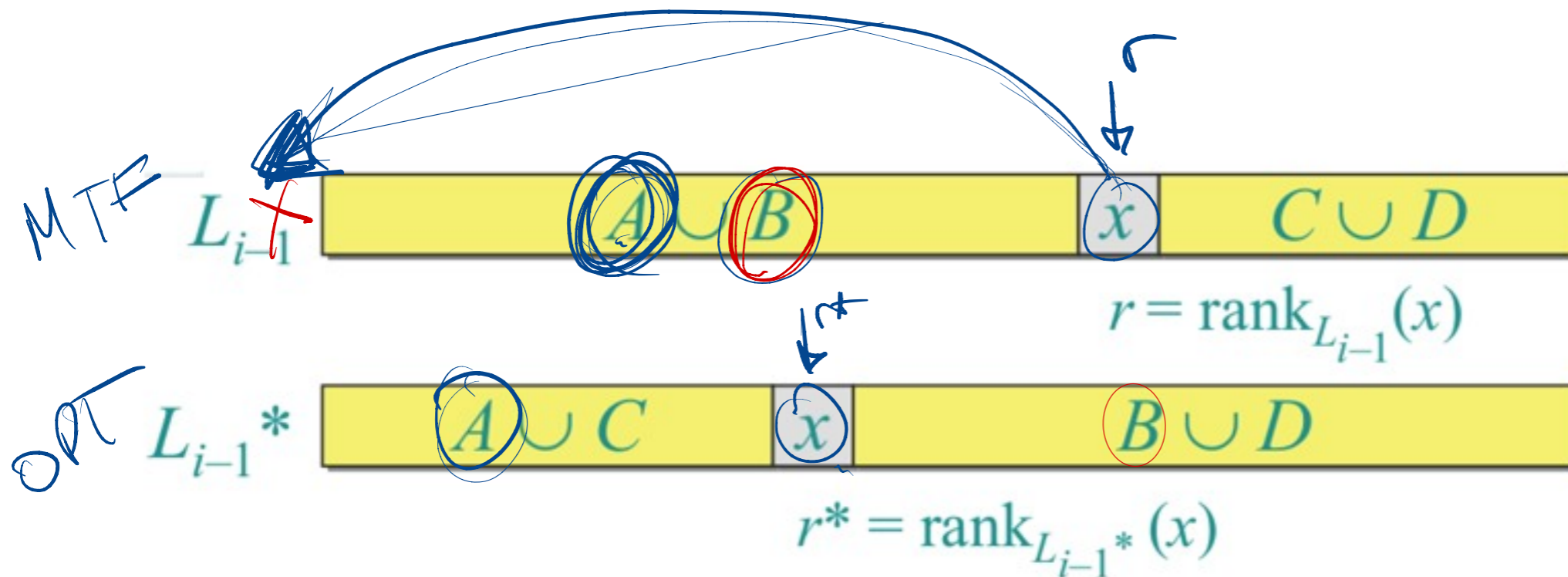
MTR = list after $i-1$ ops
 L_{i-1}



OPT $\leftarrow L_{i-1}^*$



What happens on access?



We have $r = |A| + |B| + 1$ and $r^* = |A| + |C| + 1$.

When MTF moves x to the front, it creates $|A|$ inversions and destroys $|B|$ inversions. Each swap by OPT creates ≤ 1 inversion.

Thus, we have

$$\Phi(L_i) - \Phi_{L_{i-1}} \leq 2(|A| - |B| + t_i).$$

MTF
OPT

#swaps OPT does after Access(i)

Amortized cost

The amortized cost for the i th operation of MTF with respect to Φ is

$$\hat{c}_i = c_i + \Phi(L_i) - \Phi(L_{i-1})$$

Amortized cost

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$$\hat{c}_i = c_i + \Phi(L_i) - \Phi(L_{i-1})$$
$$\leq 2r + 2(|A| - |B| + t_i) \rightarrow \text{upper bound of } \Delta\Phi$$

Amortized cost

The amortized cost for the i th operation of MTF with respect to Φ is

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(L_i) - \Phi(L_{i-1}) \\ &\leq 2r + 2(|A| - |B| + t_i) \\ &= 2r + 2(|A| - (r - 1 - |A|) + t_i) \\ &\quad (\text{since } r = |A| + |B| + 1)\end{aligned}$$

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Amortized cost

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(since $r = |A| + |B| + 1$)

$$= 2r + 4|A| - 2r + 2 + 2t_i$$

$$= \underline{4|A|} + \underline{2} + \underline{2t_i}$$

$$\leq 4(r^* + t_i)$$

(since $r^* = |A| + |C| + 1 \geq |A| + 1$)

$$\leq \underbrace{4(|A| + 1)}_{\leq r^*} + 4t_i$$

Amortized cost

The amortized cost for the i th operation of MTF with respect to Φ is

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(L_i) - \Phi(L_{i-1}) \\ &\leq 2r + 2(|A| - |B| + t_i) \\ &= 2r + 2(|A| - (r - 1 - |A|) + t_i) \\ &\quad (\text{since } r = |A| + |B| + 1) \\ &= 2r + 4|A| - 2r + 2 + 2t_i \\ &= 4|A| + 2 + 2t_i \\ &\leq 4(r^* + t_i) \rightarrow c^* = \overset{\text{ACC}}{r^*} + \overset{\text{SWAPS}}{t_i} \\ &\quad (\text{since } r^* = |A| + |C| + 1 \geq |A| + 1) \\ &= 4c_i^*\end{aligned}$$

WANT

$$\sum_{\text{MTF}} c_i \leq \sum_{\text{MTF-Amortized}} c_i \leq 4 \sum_{\text{OPT}} c_i^*$$

The grand finale

Thus, we have

$$C_{MTF}(S) = \sum_{i=1}^{|S|} c_i$$

The grand finale

Thus, we have

$$\begin{aligned} C_{MTF}(S) &= \sum_{i=1}^{|S|} c_i \\ &= \sum_{i=1}^{|S|} (\hat{c}_i + \Phi(L_{i-1}) - \Phi(L_i)) \end{aligned}$$

The grand finale

Thus, we have

$$\begin{aligned} C_{MTF}(S) &= \sum_{i=1}^{|S|} c_i \\ &= \sum_{i=1}^{|S|} (\hat{c}_i + \Phi(L_{i-1}) - \Phi(L_i)) \\ &\leq \left(\sum_{i=1}^{|S|} 4c_i^* \right) + \Phi(L_0) - \Phi(L_{|S|}) \end{aligned}$$

The grand finale

Thus, we have

$$\begin{aligned}C_{MTF}(S) &= \sum_{i=1}^{|S|} c_i \\&= \sum_{i=1}^{|S|} (\hat{c}_i + \Phi(L_{i-1}) - \Phi(L_i)) \\&\leq \left(\sum_{i=1}^{|S|} 4c_i^* \right) + \Phi(L_0) - \Phi(L_{|S|}) \\&\leq 4C_{OPT}(s) \\&\quad (\text{ since } \Phi(L_0) = 0 \text{ and } \Phi(L_{|S|}) \geq 0)\end{aligned}$$