

For advanced data structures like a Priority Queue

- ❖ We should care about the aggregate cost of a sequence of operations
- ❖ This might not be as bad as the per-operation worst cases suggest
- ❖ Amortized costs are a handy way to reason about aggregate costs

For advanced data structures like a Priority Queue

- ❖ We should care about the aggregate cost of a sequence of operations
- ❖ This might not be as bad as the per-operation worst cases suggest
- ❖ Amortized costs are a handy way to reason about aggregate costs

I've designed a data structure that supports push at amortized cost $O(1)$ and popmin at amortized cost $O(\log M)$, if the number of items never exceeds N .



Ex. sheet 6 q.6 asks you to think through why this is a sensible restriction

For any sequence of $m_1 \times$ push and $m_2 \times$ popmin, applied to an initially empty data structure,

$$\begin{array}{l} \text{aggregate} \\ \text{true} \\ \text{cost} \end{array} \leq \begin{array}{l} \text{aggregate} \\ \text{amortized} \\ \text{cost} \end{array} \leq m_1 O(1) + m_2 O(\log N) = O(m_1 + m_2 \log N)$$

SECTION 7.4

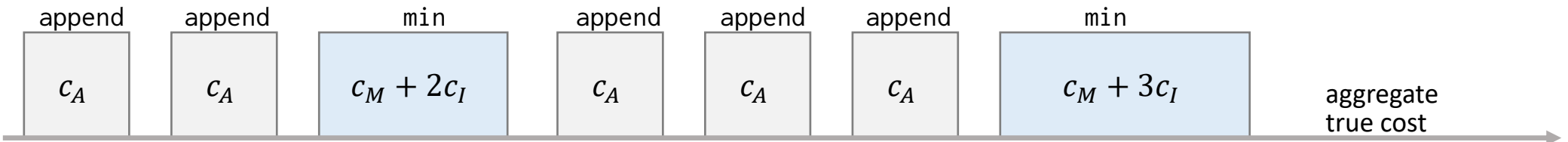
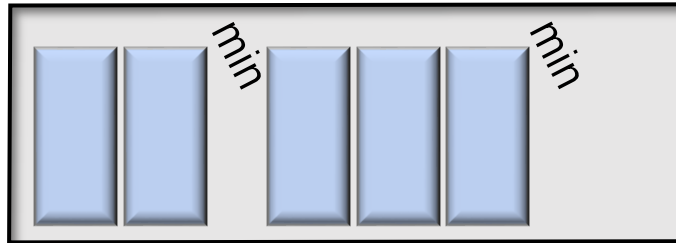
Potential functions

or, how on earth do we come up with
useful amortized costs?

```
class MinList<T>:
```

```
    def append(T value):
        # append a new value
```

```
    def T min():
        # caches the result, so we
        # only need to iterate over
        # newly-appended items
```

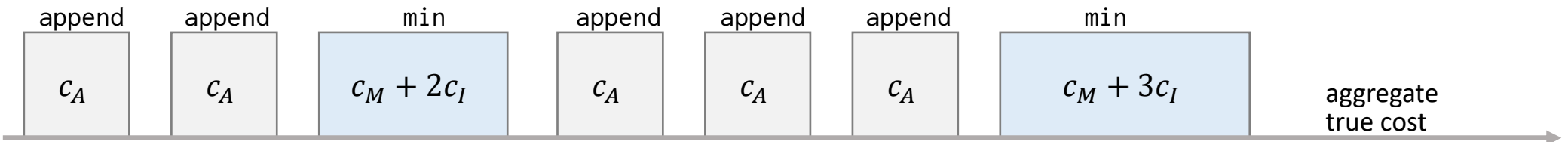
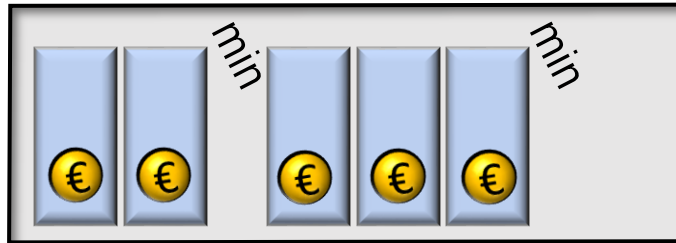


- ❖ Suppose we can store 'credit' in the data structure, and operations can either store or release credit
- ❖ Let the 'accounting' cost of an operation be: $\left(\begin{matrix} \text{accounting} \\ \text{cost} \end{matrix}\right) = \left(\begin{matrix} \text{true} \\ \text{cost} \end{matrix}\right) + \left(\begin{matrix} \text{credit} \\ \text{it stores} \end{matrix}\right) - \left(\begin{matrix} \text{credit} \\ \text{it releases} \end{matrix}\right)$
- ❖ Let's 'pay ahead' for the potentially-costly operations

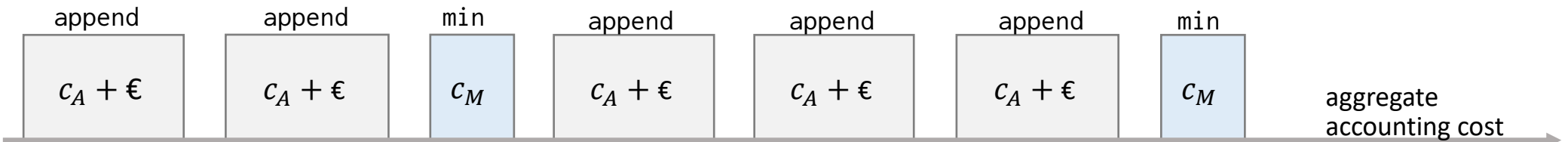
```
class MinList<T>:
```

```
def append(T value):
    # append a new value
```

```
def T min():
    # caches the result, so we
    # only need to iterate over
    # newly-appended items
```



$$acc. cost = c_M + 2c_I + 0 - 2c_I = c_M$$



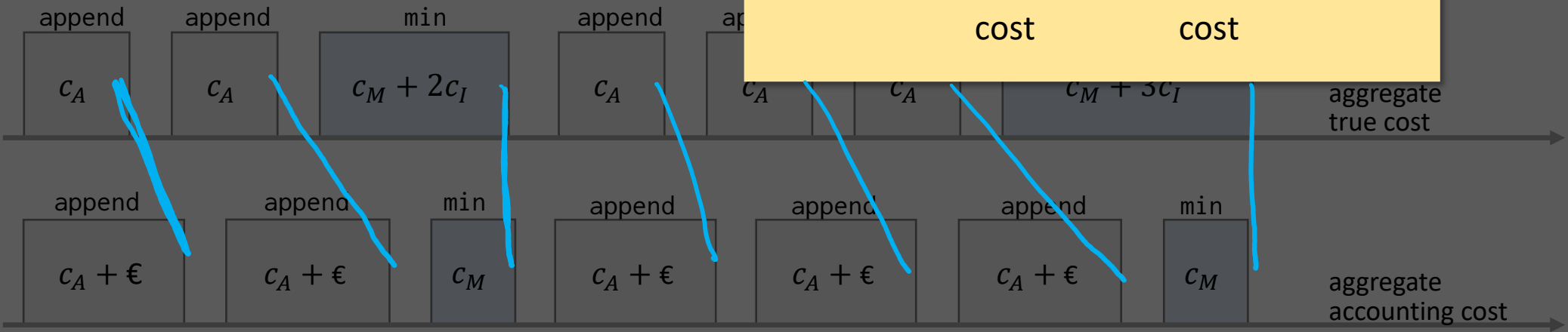
$$\text{Store } 1€ = c_I$$

- ❖ Suppose we can store 'credit' in the data structure, and operations can either store or release credit
- ❖ Let the 'accounting' cost of an operation be: $\left(\begin{smallmatrix} \text{accounting} \\ \text{cost} \end{smallmatrix} \right) = \left(\begin{smallmatrix} \text{true} \\ \text{cost} \end{smallmatrix} \right) + \left(\begin{smallmatrix} \text{credit} \\ \text{it stores} \end{smallmatrix} \right) - \left(\begin{smallmatrix} \text{credit} \\ \text{it releases} \end{smallmatrix} \right)$
- ❖ Let's 'pay ahead' for the potentially-costly operations

```
class MinList<T>:
    def append(T value):
        # append a new value
    def T min():
        # caches the result, so we
        # only need to iterate over
        # newly-appended items
```

These are valid amortized costs
 i.e. for any sequence of operations on an initially-empty data structure

$$\text{aggregate true cost} \leq \text{aggregate amortized cost}$$



Let Ω be the set of all states our data structure might be in.

A function $\Phi: \Omega \rightarrow \mathbb{R}$ is called a **potential function** if

$$\Phi(S) \geq 0 \quad \text{for all } S \in \Omega$$

$$\rightarrow \Phi(\text{empty}) = 0$$

state before *state after*

Φ = bank balance
= total amount of credit stored in the data structure.

For an operation $S_{\text{ante}} \rightarrow S_{\text{post}}$ with true cost c , define the **accounting cost** to be

$$c' = c + \Phi(S_{\text{post}}) - \Phi(S_{\text{ante}})$$

THE 'POTENTIAL' THEOREM: These are valid amortized costs.

PROOF: Consider an arbitrary sequence of operations, starting from empty: $\underbrace{S_0}_{\text{empty}} \xrightarrow{c_1} S_1 \xrightarrow{c_2} S_2 \rightarrow \dots \xrightarrow{c_m} S_m$

aggregate
accounting cost = $c'_1 + c'_2 + \dots + c'_m$

$$= -\Phi(S_0) + c_1 + \cancel{\Phi(S_1)} - \cancel{\Phi(S_1)} + c_2 + \cancel{\Phi(S_2)} \dots - \cancel{\Phi(S_{m-1})} + c_m + \Phi(S_m)$$

$$= -\cancel{\Phi(S_0)} + \underbrace{c_1 + \dots + c_m}_{\text{agg. true cost}} + \Phi(S_m)$$

\geq aggregate true cost

Example

Consider a dynamically-sized array to which we append items. It starts with capacity 1, and doubles its capacity whenever it becomes full.

Suppose the cost of writing an item is 1, and the cost of doubling capacity from m to $2m$ (and copying across the existing items) is κm . $O(m)$

Show that the amortized cost of append is $O(1)$.

$\Phi = \# \text{ items added since last doubling} \times 2\kappa$

We want the coins that we've stored so far (2ϵ) to pay for the doubling (cost 4κ). So we want $1\epsilon = 2\kappa$.

initially empty



$c=1$ append() $\text{am. cost } c' = c + \Delta\Phi = 1 + 2\kappa$



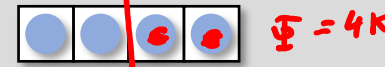
$c=2\kappa+1$ append(), requires doubling $c' = c + \Delta\Phi = \underline{1 + \kappa}$



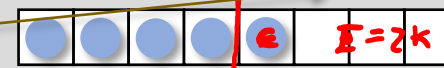
$c=2\kappa+1$ append(), requires doubling $c' = 1 + 2\kappa$



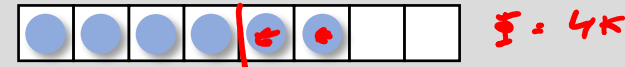
$c=1$ append() $c' = 2\kappa+1$



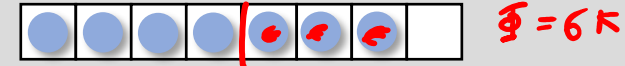
$c=4\kappa+1$ append(), requires doubling $c' = 4\kappa+1 - 2\kappa = 1 + 2\kappa$



$c=1$ append() $c' = 1 + 2\kappa$



$c=1$ append() $c' = 1 + 2\kappa$



Am cost is always $\leq 1 + 2\kappa = O(1)$

Example (sloppy style)

page 58

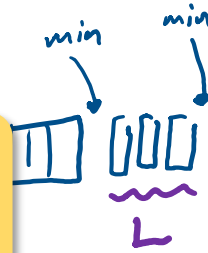
Consider a dynamically-sized array to which we append items. It starts with capacity 1, and doubles its capacity whenever it becomes full.

Suppose the cost of writing an item is $O(1)$, and the cost of doubling capacity from m to $2m$ (and copying across the existing items) is $O(m)$.

Show that the amortized cost of append is $O(1)$.

```
class MinList<T>:
    def append(T value):
        # append a new value
    def T min():
        # return the smallest
        # (without removing it)
```

QUESTION. What potential function might we use, to show that append and min both have amortized cost $O(1)$?



$L = \# \text{ items added since last min}$

$$\Phi = L$$

Stage 0

- Use a linked list
- min iterates over the entire list

Stage 1

- Use a linked list
- min caches its result, so that next time it only needs to iterate over newer values

Stage 2

- Use a linked list
- Store the current minimum, and update it on every append

Stage 3

- min caches its result, the same as Stage 1
- ... but we argue it's just as good as Stage 2



For one-shot algorithms such as sorting:

After we show that our algorithm is $O(n \log n)$, it's good manners to also show that the worst case is $\Omega(n \log n)$.

$\exists \kappa > 0$ such that, for all sufficiently large n ,
 $cost_n \leq \kappa n \log n$

$\exists \delta > 0$ and a sequence of example inputs with increasing n such that
 $cost_n \geq \delta n \log n$

i.e. design a family of example inputs of increasing size n where
 $cost_n = \Omega(n \log n)$



and, if we can't find matching O - Ω bounds, then maybe our O bound isn't as good as it could be.

For advanced data structures:

After we find big- O upper bounds for amortized costs, it's good manners to show matching worst-case performance.

$\exists \kappa > 0$ such that, for all sufficiently large N , and any operation-sequence s having $m_1 \times \text{push} + m_2 \times \text{popmin}$ such that $\# \text{items}$ is always $\leq N$,

$$\text{cost}_s \leq \kappa (m_1 + m_2 \log N)$$

Design a family of operation-sequences $s(N)$ having $m_1(N) \times \text{push} + m_2(N) \times \text{popmin}$ such that $\# \text{items}$ is always $\leq N$, and

$$\text{cost}_{s(N)} = \Omega(m_1(N) + m_2(N) \log N)$$

I've designed a data structure that supports push at amortized cost $O(1)$ and popmin at amortized cost $O(\log N)$, if the number of items never exceeds N .



for any seq of m_1 push
 m_2 popmin

agg. true cost
 \leq agg. am cost
 $= m_1 O(1) + m_2 O(\log N)$

and, if we can't find matching O - Ω bounds, then maybe our amortized costs aren't as good as they could be.

[See ex.sheet 6 q.7]

✓ Important question.

HAPPENINGNEXT



Fully Connected 2023

Schedule

Thu Mar 16 2023 at 10:00 am to 05:30 pm
UTC+00:00

Location

Computer Laboratory, University of
Cambridge | Cambridge, EN

Fully Connected
16.03.23 William Gates Building

Cambridge's flagship machine learning research event for undergraduates.

Logos: CUC@TS CU AI, Cambridge ML Collective, Microsoft, INVENIA LABS

Advertisement

Interested in ML research? Wondering what that ChatGPT thing is all about? Want a free lunch? Come to the CL!

About this Event

Interested in ML research but don't know where to start? Looking for a summer research project? Wondering what that ChatGPT thing is all about? Want a free lunch? Come along to **Fully Connected, Cambridge's flagship machine learning research event for undergraduates!**