

Lecture 24: Online Algorithms

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601.433/633 Introduction to Algorithms

Introduction

Class until now: difficulty was *computational power*

Today: difficulty is *lack of information*

Online:

- ▶ Input / data arrives *over time*
- ▶ Need to make decisions without knowing future

Ski Rental Problem

Want to go skiing, but don't know how many times you'll be able to go this year.

Should you rent or buy?

- ▶ Renting skis: \$50
- ▶ Buying skis: \$500
- ▶ Every day you ski and haven't yet bought, need to decide: rent or buy?

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- ▶ What if you ski $M \approx \infty$ times?
- ▶ Should have bought (\$500), instead rented ($M \cdot \50)

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What's the right strategy (for these costs)?

Better Late Than Never

Rent until you realize you should have bought!

BLTN: Rent **9** times, buy on **10**'th.

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If $\text{ski} \geq 10$ times:

- ▶ **ALG = 450 + 500 = 950**
- ▶ **OPT = 500**

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If $\text{ski} \geq 10$ times:

- ▶ **ALG = 450 + 500 = 950**
- ▶ **OPT = 500**

Never more than twice (actually $\frac{19}{10}$ times) what we should have done!

Competitive Ratio

Definition

The *competitive ratio* of algorithm **ALG** is the maximum over all inputs/futures σ of

$$\frac{\mathbf{ALG}(\sigma)}{\mathbf{OPT}(\sigma)},$$

where **ALG**(σ) is the cost of **ALG** on σ and **OPT**(σ) is the optimal cost for σ (knowing the future).

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So on ski rental problem with previous values, competitive ratio is $\frac{19}{10}$.

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$\$r$ to rent, $\$p$ to buy. Assume r divides p for simplicity.

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- ▶ **ALG** = $r \cdot \left(\frac{p}{r} - 1\right) + p = p - r + p = 2p - r$
- ▶ **OPT** = $\min(r \cdot z, p) = p$

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So for all inputs / futures, $\frac{\text{ALG}}{\text{OPT}} \leq 2 - \frac{r}{p}$

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$$\begin{aligned} \frac{\text{ALG}}{\text{OPT}} &= \frac{xr + p}{(x + 1)r} = \frac{xr + p}{xr + r} = 1 + \frac{p - r}{xr + r} \\ &\geq 1 + \frac{p - r}{(\frac{p}{r} - 1)r + r} = 1 + \frac{p - r}{p} = 2 - \frac{r}{p} \end{aligned}$$

Elevator Problem

Trying to get up a building: takes **E** seconds by elevator, **S** seconds by stairs.

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If elevator arrives at $x \leq S - E$:

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Paging

Classical problem in computer systems/theory

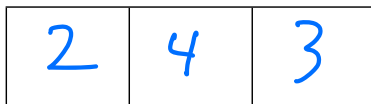
- ▶ Disk (slow) with \mathbf{N} pages
- ▶ Memory (fast) with room for $\mathbf{k} < \mathbf{N}$ pages
- ▶ If OS/application requests a page not in memory: “page fault”
 - ▶ Need to bring requested page into memory, *evict* a page from memory (if currently full)
- ▶ **Question:** What to evict?

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Example: $k = 3$. Requests: 1, 2, 3, 2, 4, 3, 4, 1, 2, 3, 4



(Convention: initial page faults to fill table don't count: only pay when we *evict* a page)

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- ▶ **k = 3, N = 4**

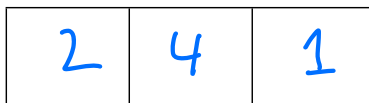
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- ▶ $k = 3$, $N = 4$
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So LRU has competitive ratio $\approx k$

- ▶ LRU evicts every time, **OPT** evicts **1** out of every **k** times.

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Request sequence:

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$$\implies \frac{\mathbf{ALG}}{\mathbf{OPT}} \geq k$$

Marking Algorithm

Get around lower bound by using *randomization*

- ▶ Lower bound argument doesn't apply because can't set request sequence to ask for whatever's not in memory, since that involved randomness! (Oblivious adversary)

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Assume memory initially **1, 2, ..., k**.

Set all pages in memory to be "unmarked"

When page requested:

- ▶ If already in memory, "mark" it
- ▶ If not in memory:
 - ▶ If all pages in memory "marked", unmark all
 - ▶ Choose an *unmarked* page uniformly at random to evict
 - ▶ Bring in new page, mark it

Marking Analysis

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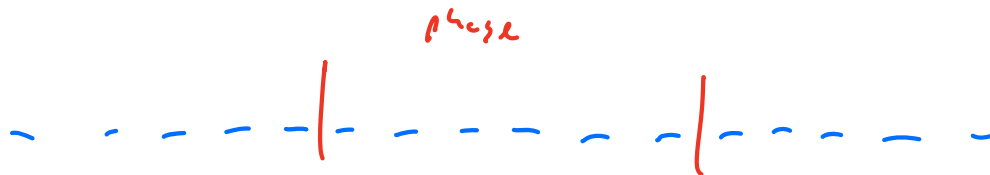
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Phase: time between “unmark all” events.

In each phase:

- ▶ $\text{OPT} \geq 1$, since all N pages requested

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 $\Pr[\text{eviction}] = \Pr[\text{requested page not in memory}] = 1/i$
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At beginning of phase $i = N$, at end of phase $i = 1$. Goes down by one every time page gets marked.

$$\implies \text{expected cost in phase at most } \frac{1}{N} + \frac{1}{N-1} + \frac{1}{N-2} + \cdots + \frac{1}{2} + 1 = O(\log N) = O(\log k)$$