Online Algorithms

- Theory of Computation
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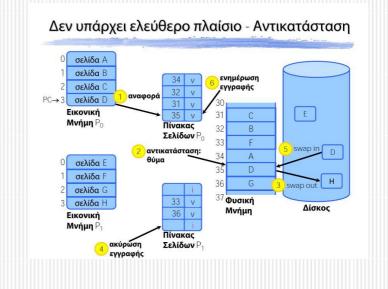
Types of Problems

□ For certain problems, input is not available from the beginning

Certain decisions are requested on the way

Output required





Online vs Offline

- Online Algorithms
 - Input arrives as sequence of input portions
 - The system must react in response to request
 - Future input is unknown
 - Not optimal
- Offline Algorithms
 - Entire input is given in advance
 - Solve the problem at hand
 - Future is known
 - Optimal

Competitive Analysis

- Big O complexity can't be used:
 - For every algorithm there will be a sequence that makes it look foolish
- Competitive Ratio
 - Comparison with an optimal offline algorithm processing the same sequence of requests
 - Maximum cost over all possible input sequences divided by the cost of an optimal offline algorithm
 - Related to minimax concept of game theory
 - Online player vs Adversary

Competitive Analysis

- □ A little formalism:
 - $cost_A(\sigma)$: the cost of an online algorithm A on the input sequence σ
 - $cost_{OPT}(\sigma)$: the cost of the optimal offline algorithm on σ
- Algorithm A is c competitive if there exists a constant b such that on every request sequence σ :

 $cost_A(\sigma) \leq c \cdot cost_{OPT}(\sigma) + b$

The Ski Rental Problem

- Cost for renting a pair of skis
- Cost for buying a pair of skis
- Rent or Buy? When?
 - How do we decide?
- Request = "Take a ski trip"
- Actions = "rent" | "buy" | "use skis already bought"
- $\Box \quad \text{Costs} = 1, \, \text{s}, \, 0 \text{ respectively}$
- On a sequence of t requests any sensible online algorithm is of the form:

"Rent for the first k trips, then buy, then use already bought"



The Ski Rental Problem

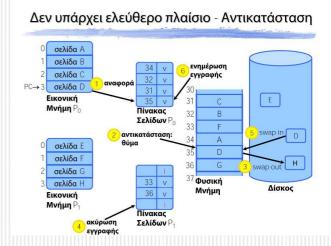
- Online Cost
 Offline Cost
 - t, $t \leq k$
 - k+s, t > k
- Find k that minimizes the competitive ratio.
 For given k, k+1 maximizes the ratio

 $\frac{k+s}{\min(k+1, s)}$

min(s, t)

- Given k, k+1 requests maximize the ratio. The ratio is minimized for k = s 1
- The on-line player should rent until enough ski trips have occurred so that he would have done better if he had bought skis initially

- Memory management scheme
- Memory hierarchy
- Page fault minimization
- Set of n pages
- **RAM** with capacity for k pages
- □ The system receives requests for pages in RAM
- □ If the referenced page is in the RAM, the request can be served
- If not, then a page fault occurs
- The missing page is loaded from secondary storage and an online algorithm has to decide which page to evict



- Common algorithms
 - LRU: evict the page in memory that was requested least recently
 - FIFO: evict the page that has been longest in memory
- Theorem
 - FIFO and LRU are k-competitive, where k the size of main memory in pages
- There exists a more general class of algorithms that achieve a competiveness of k

κολουθία α	ναφορών 9 σφά							ίλματα		
7, 0, 1, I	2, 0,	3, 0	, 4,	2,	3, 0,	3, 2	, 1, 2	2, 0,	1, 7,	0,
↓ 7 7 7 0 0 1 λαίσια μνήι	2 0 1 4ης	2 0 3	2 4 3		2 0 3		2 0 1		7 0 1	

...στο μέλλον.

Ιδανική στρατηγική, χρήσιμη ως μέτρο σύγκρισης.

🗢 κατάσταση ανάλογη του SJF.

Marking

- Each page is associated with a bit called mark
- Initially all pages are set as unmarked
- Stages of page requests
- A page is marked when it is first requested in this stage
- On a fault, an unmarked page is evicted

Theorem

Any marking algorithm is k-competitive

Αλγόριθμοι προσέγγισης LRU

Bit αναφοράς

- ◄ Για κάθε σελίδα στον πίνακα σελίδων, αρχικά 0
- ➤ Οποτεδήποτε γίνεται αναφορά σε αυτή, τίθεται 1
- Ξέρουμε σε ποιες σελίδες έγιναν αναφορές
 δεν τηρείται όμως η σειρά των προσβάσεων
- Όταν πρέπει να φύγει μια σελίδα
 διώξε αυτή που έχει bit αναφοράς μηδενικό

Theorem

 No deterministic online algorithm for the paging problem can achieve a ration smaller than k

Proof

- Optimal Offline Algorithm
 - Belady's greedy algorithm
 - "Sees" in the future
 - On a fault, evict the page whose next request occurs furthest in the future

Proof

- A and OPT start with the same set of pages in memory
- The adversary restricts its request sequence to a set of k+1 pages, the pages initially in the memory and another one
- It always requests the page that is outside of the memory
- This can be continued for an arbitrary number of requests, resulting in a sequence σ on which A faults on every request
- What remains is to show that $cost_{OPT}(\sigma) \leq \left[\frac{|\sigma|}{k}\right]$
- At each fault, the adversary evicts the page whose first request occurs furthest in the future
- The adversary is guaranteed that there will be at least k-1 pages requested between any two faults, so the adversary faults at most on every kth request

Adversaries

- Online algorithms can achieve better performance if they are allowed to make random choices
- The competitive ratio of a randomized algorithm is defined with respect to an adversary
- □ There are three types of adversaries:
 - oblivious adversary (weak)
 - generates the whole request in advance
 - adaptive online adversary (medium)
 - it may observe the online algorithm and generate next request based to all previous requests
 - adaptive offline adversary (strong)
 - knows everything. Even randomization can't face it

Secretary Problem

- Also known as the marriage problem, the game of googol
- □ There is a single secretarial position to fill
- □ There are n applicants for the position

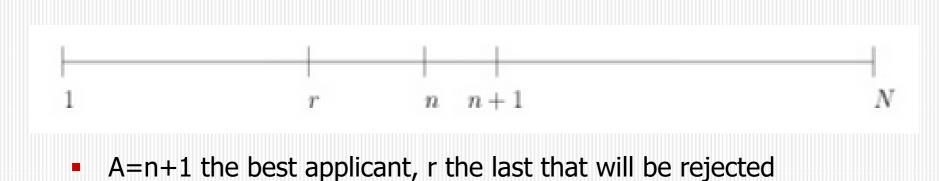


- □ The applicants can be ranked from best to worst unambiguously
- The goal is to have the highest probability of selecting the best applicant of the whole group
- They are interviewed sequentially in random order
- Immediately after the interview, the applicant is either accepted or rejected irrevocably

Secretary Problem

Strategy

- Naive: pick the ith candidate: $P(Success) = \frac{1}{N}$
- Interview the first r applicants for r<n
- Accept the very next applicant that is better than all the first r you interviewed



Secretary Problem

Strategy

- A won't be chosen, unless:
 - $n \ge r$
 - The highest applicant in [1,n] is the same as in [1,r], $P = \frac{r}{n} \frac{1}{N}$

$$P(Success) = P(r) = \frac{1}{N} \left[\frac{r}{r} + \frac{r}{r+1} + \dots + \frac{r}{N-1} \right] = \frac{r}{N} \sum_{n=r}^{N-1} \frac{1}{n}$$

• For the optimal solution, P'(r)=0 => r = $\frac{1}{e} \approx 0.37$

• Coincidentally,
$$P(r_{max}) = \frac{1}{e}$$

Applications and Further Research

- Stock Markets
 - Algorithms for stock prediction
- Large networks
 - Network switches
 - TCP Acknowledgement
- Robot Motion Planning
- Bin Packing
- Storage Allocation and Cache Management
- Job Scheduling

Bibliography

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Questions?

Ευχαριστώ