December 13, 2007

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- Online algorithms
 - Information is revealed to the algorithm in parts
 - Algorithm needs to process each part before receiving the next
 - There is **no information** about the future (in particular, no probabilistic assumptions!)
 - How well can an algorithm do compared to an algorithm that knows everything?
 - Lack of knowledge vs. lack of processing power

Example: Google ad auctions

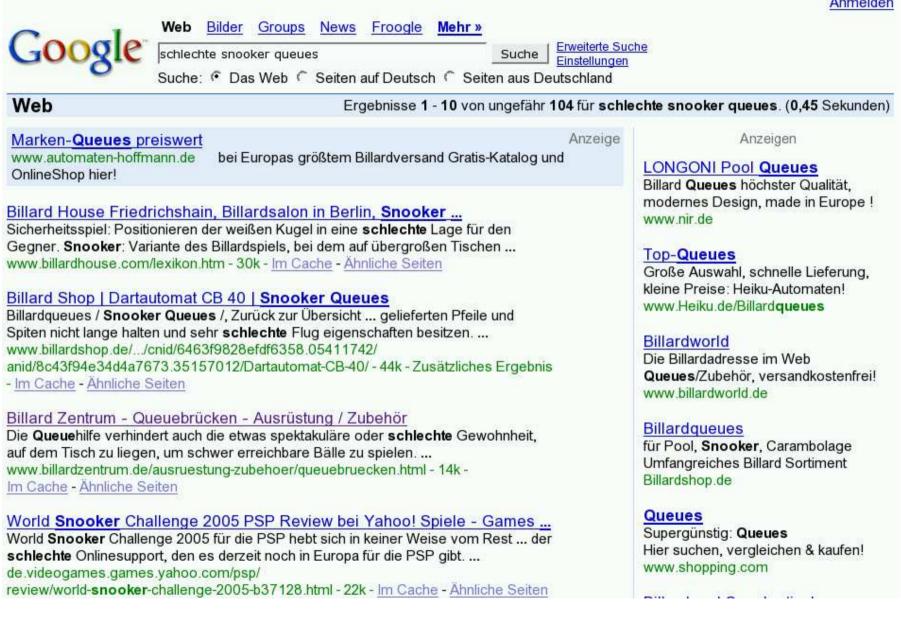


- Google is popular because of its PageRank algorithm
- However, it can only earn money through ads
- Ads are linked to specific keywords
- Advertisers select keywords, maximum price, and daily budget
- They pay only if
 - their ad is shown with the keyword
 - someone clicks on the ad
 - the daily budget has not yet been reached



Suche	
Web Ergebnisse 1 - 10 von ungefähr 202.000 f	ür snooker queues . (0,17 Sekunden)
Marken-Queues preiswert Anzeigen www.automaten-hoffmann.de bei Europas größtem Billardversand Gratis-Katalog und OnlineShop hier! Billard Henzgen www.billard-henzgen.de führender Hersteller von Billardtischen. Gratiskatalog! Billard Billardshop Billazzo - Snooker Queues Queues · Snooker Queues "Dieses Queue "breakt und jumped" von alleine!" sagt Ralf Souquet über sein Mythus! finden Sie in der Rubrik "Queues" www.billazzo.de/queues_snooker_queues.htm - 17k - Im Cache - Ähnliche Seiten Billard Zentrum - Snooker Queues - Billard Queues - Billard Shop	Anzeigen <u>LONGONI Pool Queues</u> Sie suchen höchste Qualität und modernes Design, made in Europe ' www.nir.de <u>Top-Queues</u> Große Auswahl, schnelle Lieferung, kleine Preise: Heiku-Automaten! www.Heiku.de/Billardqueues <u>Billard Shop by AniMazing</u>
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<u>Cuetec Snooker-Queue - Sport - Preis ab € 89,00 im Preisvergleich</u> Preisvergleich für Sport: Cuetec Snooker-Queue mit aktuellen Preis, Foto, Beschreibung und Händler - Angebote zum kaufen. www.preissuchmaschine.de/psm_frontend/main.asp?produkt=351139 - 58k - 4. Dez. 2006 - Im Cache - Ähnliche Seiten	für Pool, Snooker , Carambolage Umfangreiches Billard Sortiment Billardshop.de <u>Billard & Kicker Profis</u> Klasse Preis-Leistungsverhältnis







Second price auction

Advertisers only need to beat competitors: they pay the second price

When no competitors are left, price drops to 0

Google wants to maximize revenue : keep advertisers solvent!

Question for Google:

which ads should be shown each time a certain keyword is looked for?



Optimizing the revenue

Which ads should be shown? Google needs to take into account

- Clickthrough probability for a given ad
- Remaining budget of each advertiser
- **Future** queries on the same day...
- A typical online problem!

Online matching

- Advertisers, bids and budgets are known in advance
- Queries occur over time (during the day)
- Each query must be matched to some ads without knowledge of future queries

Result: it is possible to give an algorithm which gets

$$1 - \frac{1}{e} \approx 0.62312$$

of the optimal offline revenue (knowing all queries in advance)



Competitive analysis

Idea: compare online algorithm ALG to offline algorithm OPT

Worst-case performance measure

Definition:

$$C_{ALG} = \sup_{\sigma} \frac{\mathsf{ALG}(\sigma)}{\mathsf{OPT}(\sigma)}$$

(we look for the input that results in worst relative performance)

Goal:

find ALG with minimal C_{ALG}

A typical online problem: ski rental

Renting skis costs 50 euros, buying them costs 300 euros

You do not know in advance how often you will go skiing

Should you rent skis or buy them?





A typical online problem: ski rental

Renting skis costs 50 euros, buying them costs 300 euros

You do not know in advance how often you will go skiing

Should you rent skis or buy them?

Suggested algorithm: buy skis on the sixth trip

Two questions:

- How good is this algorithm?
- Can you do better?







- Upper bound for ski rental
 - ☐ You plan to buy skis on the sixth trip
 - If you make five trips or less, you pay optimal cost (50 euros per trip)
 - □ If you make at least six trips, you pay 550 euros
 - ☐ In this case OPT pays at least 300 euros
 - Conclusion: algorithm is $\frac{11}{6}$ -competitive: it never pays more than $\frac{11}{6}$ times the optimal cost

Lower bound for ski rental

Suppose you buy skis earlier, say on trip x < 6. You pay 300 + 50(x - 1), OPT pays only 50x

$$\frac{250+50x}{50x} = \frac{5}{x} + 1 \ge 2.$$

Suppose you buy skis later, on trip y > 6. You pay 300 + 50(y - 1), OPT pays only 300

$$\frac{250+50y}{300} = \frac{5+y}{6} \ge 2.$$

Idea: do not pay the large cost (buy skis) until you have paid the same amount in small costs (rent)





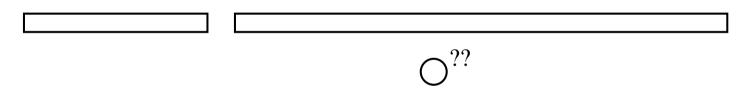
The cow path problem





The cow path problem

- A cow wants to get past a fence
 - There is a hole in this fence, but the cow does not know where it is
- □ How can it find the hole quickly?





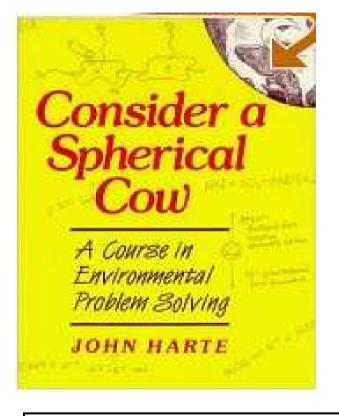
The cow path problem

- A cow wants to get past a fence
 - There is a hole in this fence, but the cow does not know where it is
- □ How can it find the hole quickly?
- Idea: use a doubling strategy





The cow path problem



O??



Algorithm

- \Box Let d = 1, side = Right
- Repeat until hole is found:
 - Walk distance d to current $\ensuremath{\operatorname{side}}$
 - If hole not found: return to starting point, double the value of d and flip side

Analysis

□ What is the worst that could happen?





Analysis

- □ What is the worst that could happen?
 - Answer: hole is slightly beyond a point where you turn around
- Denote its distance from the starting point by $2^i + \varepsilon$, where ε is very small.
- $\Box 2^i + \varepsilon$ is the distance that OPT walks.



Analysis

 $2(1+2+\cdots+2^{i+1})+2^{i}+\varepsilon$



Analysis

$$2(1+2+\cdots+2^{i+1})+2^{i}+\varepsilon$$

$$= 2(2^{i+2} - 1) + 2^i + \varepsilon$$



Analysis

$$2(1+2+\dots+2^{i+1})+2^{i}+\varepsilon$$

$$= 2(2^{i+2} - 1) + 2^i + \varepsilon$$

$$< 8 \cdot 2^{i} + 2^{i}$$



Analysis

$$2(1 + 2 + \dots + 2^{i+1}) + 2^{i} + \varepsilon$$

$$= 2(2^{i+2} - 1) + 2^{i} + \varepsilon$$

$$< 8 \cdot 2^{i} + 2^{i} + \varepsilon$$

$$< 9 \cdot \text{OPT} \qquad \text{OPT} = 2^{i} + \varepsilon$$



The adversary

- An online problem can be seen as a game between two players
- One player is the online algorithm
- The other is the adversary
- The adversary tries to make things difficult for the online algorithm
- The algorithm minimizes the competitive ratio, the adversary maximizes it



List update

- Suppose you have a collection of ℓ unsorted files
- When you need a certain file, you can only find it by going through the list from the start
- After finding a file, you may move it closer to the start of the list
- **Goal:** minimize overall search time
- Linked-list data structure
- **NP-hard** (Ambühl, 2000)



Upper bound

 \Box Every request costs at most ℓ , the length of the list

- ☐ OTP pays at least 1 for each request
- \Box Thus, no algorithm has a competitive ratio above ℓ



Three algorithms

Move-To-Front (MTF): move requested item to start of list

Transpose (TRA): exchange requested item with item before it

Frequency Count (FC): sort items by amount of requests for them



Three algorithms

Move-To-Front (MTF): move requested item to start of list Seems like an overreaction

Transpose (TRA): exchange requested item with item before it
 More conservative

Frequency Count (FC): sort items by amount of requests for them Requires bookkeeping



TRA is bad

$$\Box$$
 Consider list $x_1 \mid x_2 \mid \ldots \mid x_{\ell-1} \mid x_\ell$

Request sequence: $x_{\ell}, x_{\ell-1}, x_{\ell}, x_{\ell-1}, \dots$

] TRA pays ℓ for every request

Before request $1, 3, 5, \ldots$

Before request $2, 4, 6, \ldots$

x_1	x_2	•••	$x_{\ell-}$	1	x_{ℓ}
x_1	x_2	• • •	x_{ℓ}	x	$\ell \ell - 1$



TRA is bad

$$\Box$$
 Consider list $x_1 \mid x_2 \mid \ldots \mid x_{\ell-1} \mid x_\ell$

- Request sequence: $x_{\ell}, x_{\ell-1}, x_{\ell}, x_{\ell-1}, \dots$
- TRA pays ℓ for every request
- OPT moves both items to start of list, and then pays at most 2 per request
- \Box Long sequence of requests: TRA is not better than $\ell/2$ -competitive
- Locality of reference makes request sequences similar to these likely



FC is bad

$$\Box$$
 Let $k > \ell$, initial list is x_1, \ldots, x_ℓ

 \Box Request sequence: k times x_1 , k-1 times x_2,\ldots

 \Box In general, item x_i is requested k + 1 - i times

Example:



FC is bad

- \Box Let $k > \ell$, initial list is x_1, \ldots, x_ℓ
- \Box Request sequence: k times x_1 , k-1 times x_2 ,...

 \Box In general, item x_i is requested k + 1 - i times

□ FC never moves any item

Total cost is

$$\sum_{i=1}^{\ell} i \cdot (k+1-i) = \frac{k\ell(\ell+1)}{2} + \frac{\ell(1-\ell^2)}{3}$$



FC is bad

Total cost of FC is
$$\frac{k\ell(\ell+1)}{2} + \frac{\ell(1-\ell^2)}{3}$$

Optimal: move each page to front on first request

$$\Box$$
 Cost is $\sum_{i=1}^{k} \{i + (k-i)\} = k\ell$

		List			Requests	Optimal cost
x_1	x_2	x_3	x_4	x_5	$x_1, x_1, x_1, x_1, x_1, x_1,$	5
x_2	x_1	x_3	x_4	x_5	$x_2, x_2, x_2, x_2, x_2,$	2 + 3 = 5
x_3	x_2	x_1	x_4	x_5	$x_3, x_3, x_3,$	3 + 2 = 5
x_4	x_3	x_2	x_1	x_5	$x_4, x_4,$	4 + 1 = 5
x_5	x_4	x_3	x_2	x_1	x_5	5



- Free and paid transpositions
 - □ A transposition is the switching of two consecutive items
 - ☐ When a file is found, we can move it closer to the start of the list for free (free transpositions)
- Generally, we might also use paid transpositions: move items although they are not requested

Offline: paid transpositions are necessary!



Paid transpositions are required offline

Initial list $x_1 \mid x_2 \mid x_3$

Request sequence x_3, x_2, x_3, x_2





Move-To-Front

For an algorithm ALG, we define

 \Box ALG_P(σ) = number of paid transpositions (cost 1)

 \Box ALG_{*F*}(σ) = number of free transpositions (cost 0)

 \Box ALG_C(σ) = total cost other than paid transpositions

We show

$$\mathsf{MTF}(\sigma) \le 2 \cdot \mathsf{OPT}_C(\sigma) + \mathsf{OPT}_P(\sigma) - \mathsf{OPT}_F(\sigma) - n$$



Potential functions

- It is often hard to analyze the performance of an online algorithm on a long input sequence at once
- Analysis per request also usually does not work: OPT may have very low cost for some requests
- Solution: potential functions
- Idea: keep track of configurations of OPT and ALG
- If an action of ALG makes its configuration more similar to OPT, it is allowed to cost more
 - Very powerful technique



Potential function

An *inversion* in the list of MTF with respect to the list of OPT is an ordered pair (x, y) for which

 $\Box x$ precedes y on list of MTF

 $\Box y$ precedes x on list of OPT

Let $t_i = \text{cost of MTF}$ for request i.

Let Φ_i = number of inversions after request i.

 Φ_i is a potential function.

We define amortized costs

$$a_i = t_i + \Phi_i - \Phi_{i-1}$$



Potential function

 $\Box \Phi_i$ is always nonnegative

 $\Box \Phi_0 = 0$ (lists are the same at the start)

$$\begin{aligned} \mathsf{MTF}(\sigma) &= \sum_{i=1}^{n} t_{i} = \sum_{i=1}^{n} (a_{i} - \Phi_{i} + \Phi_{i+1}) \\ &= \Phi_{0} - \Phi_{n} + \sum_{i=1}^{n} a_{i} \\ &= \sum_{i=1}^{n} a_{i} - \Phi_{n} \end{aligned}$$

Sanders/van Stee: Approximations- und Online-Algorithmen

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Potential function

- $\Box\,$ We can bound $\mathrm{MTF}(\sigma)$ by bounding the amortized costs
 - For request *i* and for OPT, let s_i be the search cost, and P_i (F_i) the number of paid (free) transpositions

We show

$$a_i \le (2s_i - 1) + P_i - F_i$$

Thus we relate the amortized costs to the optimal decisions

Summing this for all i gives the theorem



From claim to theorem

lf

$$a_i \le (2s_i - 1) + P_i - F_i$$

then

$$\sum_{i=1}^{n} a_{i} \le 2 \sum_{i=1}^{n} s_{i} - n + \sum_{i=1}^{n} P_{i} - \sum_{i=1}^{n} F_{i}$$

Therefore

$$\begin{split} \mathsf{MTF}(\sigma) &= \sum_{i=1}^{n} a_{i} - \Phi_{n} \\ &\leq 2 \cdot \mathsf{OPT}_{C}(\sigma) + \mathsf{OPT}_{P}(\sigma) - \mathsf{OPT}_{F}(\sigma) - n \end{split}$$

which is what we wanted to show



Comparing the lists

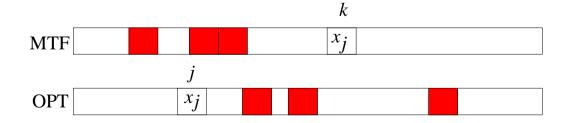
 \Box Consider the number of inversions involving the current request x_j

 x_j is at position j in list of OPT, position k in list of MTF

Suppose there are v items that are before x_j in list of MTF and behind x_j in list of OPT

 \Box Then OPT has at least k - 1 - v items before x_j

$$\Box$$
 Thus, $k-1-v \leq j-1$





Comparing the lists (2)

□ MTF moves x_j to front of list : k - 1 - v new inversions created, v inversions removed

Contribution to amortized cost:

$$k + (k - 1 - v) - v = 2(k - v) - 1 \le 2j - 1 = 2s_i - 1$$

A paid exchange adds at most 1 to the potential function

 \Box A free exchange contributes -1

] This proves the claim $a_i \leq (2s_i - 1) + P_i - F_i$



Randomized algorithms

- A randomized algorithm is allowed to use random bits in its decision-making
- □ We compare its expected cost to the optimal cost
- The adversary knows the probability distribution(s) and chooses the input sequence in advance (oblivious adversary)

] ALG is c-competitive against an oblivious adversary if

 $\mathbb{E}(\mathsf{ALG}(\sigma)) \le c \cdot \mathsf{OPT}(\sigma) + \alpha,$

where α is a constant that does not depend on the input σ



Comparison to approximation algorithms

- □ Here, we are not interested in running times
 - Purpose of randomization is only to decrease competitive ratio
 - Compare weighted vertex cover:
 - deterministic 2-approximation solved linear program (time $O(n^{3.5}L)$)
 - randomized 2-approximation only flipped at most n coins (time O(n))
 - Disadvantage: random bits are not so easy to find



The BIT algorithm

- \Box For each item x on the list, BIT uses one bit b(x)
 - At the start, each bit is set to 0 or 1, independently and uniformly
 - Whenever an element x is requested:
 - flip bit b(x)
 - If b(x) = 1, move x to front, else do nothing

Potential function



- □ Let w(x, y) be the weight of inversion (x, y) = the number of times y is accessed before y passes x in the list of BIT
- \Box We have w(x,y) = b(y) + 1 (= 1 or 2)

$$\Box \text{ Define potential } \Phi \text{ as } \Phi = \sum_{\text{inversions (x,y)}} w(x,y).$$

 \Box We have $\Phi_0=0$ and $\Phi_n\geq 0$

 \Box For amortized costs $a_i = \mathsf{BIT}_i = \Phi_i - \Phi_{i-1}$ we have

$$BIT(\sigma) = \sum_{i=1}^{n} \mathsf{BIT}_i = \Phi_0 - \Phi_n + \sum_{i=1}^{n} \mathbf{a}_i$$



Events

There are two types of events in the sequence:

- □ A paid exchange by OPT
- All other operations from BIT and OPT to serve a request
- For both types, we can show that the amortized cost of BIT is at most 7/4 times the optimal cost

Here we only discuss the first type of event



Paid exchange of OPT

 \Box Suppose event i in the sequence is a paid exchange of OPT

 \Box Then $\mathsf{OPT}_i = 1$ and $\mathsf{BIT}_i = 0$

The exchange might create a new inversion of weight 1 or 2

 \Box If it creates no new inversion, $\Phi_i = \Phi_{i-1}$ and we are done

 \Box Else, recall that w(x, y) = b(y) + 1

 \Box b(y) is 0 or 1, both with probability 1/2, at the start of the algorithm and therefore throughout

$$\Box$$
 Thus $\mathbb{E}(a_i) = \frac{1}{2}(1+2) \leq \frac{3}{2} \cdot \mathsf{OPT}_i$