# Learning in Big Data Analytics Lecture 4 

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## RECAP

- Placing web advertisements means assigning ads to search queries
- Advertisers bid on queries
- Advertisers have overall budget
- Ads have click-through rate
- Ads need to be ranked according to bid, budget, rate to maximize revenue for search engine
- Decision need to be taken online, without delay Online algorithms
- Competitive ratio is fraction of revenue acquired with online relative to optimum offline algorithm
- Ads need to be matched with queries Matching algorithms
- Online matching well covered by greedy algorithms
- We computed the competitive ratio of greedy matching


## The Adwords Problem

## Search Advertizing Principle

Strategy by Overture [2000]

- Overture was company later acquired by Yahoo!
- Advertisers bid on keywords, as appearing in search queries
- All advertisers' links are displayed as response to user who searches keyword, highest-bid first order,
- Advertiser pays if links are clicked on
- Rather useless for users looking primarily for information which are the majority!
- Google adapted idea in system called Adwords
- Advertisers' links displayed separately from generic links


## Adwords System

## Improvements

- Google displayed only limited list of advertisements: requires to decide which to show
- Advertisers have to specify an overall budget, the amount of money to spend for clicked-on ads in a given time (e.g. a month) more involved algorithmic problem
- Google evaluated click-through rates for ads to maximize profit


## The Adwords Problem: Definition

Given

- Set of bids of advertisers for search queries
- Click-through rates for advertiser-query pairs
- Budget for each advertiser (usually specified for a month)
- Limit on number of ads to be displayed

Response to Search Query

- Set of ads no larger than the limit
- Each advertiser in the set has bid on query
- Each advertiser has sufficient budget left to pay bid


## The Adwords Problem: Definition

Adwords Algorithm: Target Function

- Value of ad is product of bid and click-through rate
- Revenue of selection of ads is sum of values
- Merit of an online-algorithm for determining selections of ads is revenue obtained over a month
- Competitive ratio is minimum of revenue for sequence of queries divided by revenue obtained for same sequence by optimum offline algorithm


## Adwords Problem: Greedy Approach

## Simplified Scenario

(a) One ad is shown for each query
(b) All advertisers have the same budget
(c) All click-through rates are the same
(d) All bids are 0 or 1

Alternative formulation of (d): the value (product bid times click-through rate) is the same for each advertiser.

Greedy Algorithm
For each search query, pick arbitrary advertiser

- who bids 1 on query
- has budget left


## Adwords Problem: Note on Reality

Matching Bids with Search Queries

- Advertisers bid on sets of words
- Exact matching: eligible when query matches set of words exactly
- Broad matching: eligible also for inexact matches
- Super- or subsets of words
- Words that have similar meaning
- Charging advertisers follows complicated formulas

Charging Advertisers for Clicks

- First price auction: Advertiser is charged the amount they bid
- Second price auction: Pay (approximate) bid of second placed advertiser
- Second price auctions less susceptible to being gamed by advertisers lead to higher revenues for search engines


## EXAMPLE

- Two advertisers, $A_{1}$ and $A_{2}$, each with budget 2
- Two possible queries, $x$ and $y ; A_{1}$ bids only on $x, A_{2}$ on $x$ and $y$
- Consider sequence of queries $x x y y$
- The Greedy algorithm
- can allocate the two $x$ to $A_{2}$
- $A_{1}$ does not bid on $y, A_{2}$ has no budget left
- Revenue is 2
- The Offline algorithm
- allocates the two $x$ to $A_{1}$, and the two $y$ to $A_{2}$
- Revenue is 4
- The competitive ratio is thus no more than $\frac{2}{4}=\frac{1}{2}$.


## The Balance Algorithm

## Balance Algorithm

- Slight adaptation of Greedy algorithm
- Assigns query to advertiser who
- bids on the query
- has the largest remaining budget
- Ties are broken arbitrarily


## Example Revisited

## Situation

- Two advertisers, $A_{1}$ and $A_{2}$, each with budget 2
- Two possible queries, $x$ and $y ; A_{1}$ bids only on $x, A_{2}$ on $x$ and $y$
- Consider sequence of queries $x x y y$

Balance Algorithm

- Can put first $x$ to $A_{2}$
- But then must put the second $x$ to $A_{1}$
- Puts first $y$ to $A_{2}$
- $A_{2}$ has no budget left to serve second $y$
- Revenue is 3 , so competitive ratio is no more than $\frac{3}{4}$


## Balance: Lower Bound Competitive Ratio

Situation

- Known upper bound on competitive ratio: $\frac{3}{4}$.
- Lower bound not known
- Idea: Establish a suitable lower bound

Claim
(i) A lower bound for the Balance algorithm, in the simple situation sketched (involving only 2 advertisers), is $\frac{3}{4}$
(ii) This establishes $\frac{3}{4}$ as the competitive ratio of the Balance algorithm

Note that $(i i)$ is an immediate consequence of $(i)$, when combining it with the upper bound we established.

## Balance: Lower Bound Competitive Ratio II

Situation

- Two advertisers, $A_{1}$ and $A_{2}$, each of which has budget B
- We need to show that for an arbitrary sequence of queries, Balance achieves at least $\frac{3}{4}$ times the revenue of the optimum offline algorithm

Immediately Possible Assumptions
(*) Given two sequences of queries, we can focus on the sequence that provably yields a smaller ratio

Suffices to show that the smaller ratio is at least $\frac{3}{4}$
${ }^{* *}$ ) The optimum offline algorithm assigns each query to one of $A_{1}$ or $A_{2}$
One can imagine to delete other queries without affecting the revenue, while the revenue of Balance can only decrease

- This yields a sequence whose ratio is smaller, make use of $\left(^{*}\right)$


## Balance: Lower Bound Competitive Ratio III

## Situation

- Two advertisers, $A_{1}$ and $A_{2}$, each of which has budget B
- We need to show that for an arbitrary sequence of queries, Balance achieves at least $\frac{3}{4}$ times the revenue of the optimum offline algorithm


## Immediately Possible Assumptions

$\left(^{* * *}\right)$ Both budgets are consumed by optimum offline algorithm

- If not, consider reduced, but fully consumed budgets
- Revenue of optimum offline algorithm remains the same
- Note that the assumption of equal budget needs to be skipped
- Ratio also applies for unequal budgets exercise!
- Balance revenue can only decrease

Lowers ratio

## Balance: Lower Bound Competitive Ratio IV


(a) Optimum

(b) Balance

Adopted from mmds.org

## Balance: Lower Bound Competitive Ratio V

- Some queries assigned to $A_{2}$


Adopted from mmds.org by Balance could have been assigned to $A_{1}$ by offline optimum (dark queries)

- Let $y$ be number of queries assigned to $A_{1}$ (by Balance)
- Let $x=B-y$ be number of unassigned queries

We seek to show that

$$
\begin{equation*}
y \geq x \quad \text { implying that } \quad y \geq \frac{1}{2} B, \quad \text { yielding } \quad B+y \geq B+\frac{1}{2} B=\frac{3}{2} B \tag{1}
\end{equation*}
$$

## Balance: Lower Bound Competitive Ratio Vi



Adopted from mmds.org

- $x$ is also the number of queries left unassigned by Balance
- All $x$ queries must have gone to $A_{2}$ by the optimum algorithm
- Assigning any of the $x$ queries to $A_{1}$ means that $A_{1}$ would have bid on the queries
- So, because $A_{1}$ had budget left, they would have been assigned to $A_{1}$ also by Balance


## Balance: Lower Bound Competitive Ratio Vi



Adopted from mmds.org

- Consider queries that are assigned to $A_{1}$ by Optimum (dark in figure)
- Recall that all such queries are assigned by Balance, either to $A_{1}$ or $A_{2}$

Two Cases
(i) More than half of dark queries are assigned to $A_{1}$ by Balance
(ii) More than half of dark queries are assigned to $A_{2}$ by Balance

## Balance: Lower Bound Competitive Ratio VII



Adopted from mmds.org

Two Cases
(i) More than half of dark queries are assigned to $A_{1}$ by Balance
(ii) More than half of dark queries are assigned to $A_{2}$ by Balance

CASE (i): This case immediately implies that $y \geq B / 2$, which implies $y \geq x$, so we are done.

## Balance Algorithm: Lower Bound Competitive Ratio VI



Adopted from mmds.org
CASE (ii): More than half of dark queries are assigned to $A_{2}$.
Consider the last dark query assigned to $A_{2}$ by Balance. At that point, $A_{2}$ 's budget must have been at least as great as $A_{1}{ }^{\prime}$ s budget, because otherwise, by the algorithmic principle of Balance, $q$ would have been assigned to $A_{1}(+)$.

Since more than $\mathrm{B} / 2$ dark queries are assigned to $A_{2}, A_{2}$ 's budget was at most $B / 2$ just before $q$ arrived.
Because of (+), this implies that also $A_{1}$ 's budget was at most $\mathrm{B} / 2$, so $A_{1}$ had already collected at least $\mathrm{B} / 2$ queries. So $y \geq B / 2$, implying $y \geq x$.

## Balance Algorithm with Many Bidders

The competitive ratio involving many bidders can be lower than $\frac{3}{4}$, but not much lower.

Worst-Case Scenario

1. There are $N$ advertisers $A_{1}, \ldots, A_{N}$
2. Each advertiser has budget $B=N$ !
3. There are $N$ queries $q_{1}, \ldots, q_{N}$
4. Advertiser $A_{i}$ bids on queries $q_{1}, \ldots, q_{i}$
5. The query sequence consists of $N$ rounds, where the $i$-th round consists of $B$ occurrences of $q_{i}$

Optimum Offline Algorithm

- Assigns all bids of $i$-th round to advertiser $A_{i}$
- Yields revenue $N \cdot B$


## Balance Algorithm with Many Bidders



Balance Algorithm

- Assigns all $B$ occurrences of $q_{1}$ equally to all $A_{i}, i=1, \ldots, N$
- Each advertiser gets $B / N$ of queries $q_{1}$
- Assigns $B$ occurrences of $q_{2}$ equally to all $A_{i}, i=2, \ldots, n$
- Each of $A_{2}, \ldots, A_{N}$ gets $B /(N-1)$ of queries $q_{2}$
- ...


## Balance Algorithm with Many Bidders



Balance Algorithm

- $A_{1}, \ldots, A_{N}$ get $B /(N-i+1)$ of queries $q_{i}$
- Eventually, budgets of higher-numbered advertisers will be exhausted


## Balance Algorithm with Many Bidders



Balance Algorithm

- Eventually, budgets of higher-numbered advertisers will be exhausted
- This happens at lowest round $j$ where

$$
\begin{equation*}
B\left(\frac{1}{N}+\frac{1}{N-1}+\ldots+\frac{1}{N-j+1}\right) \geq B \tag{2}
\end{equation*}
$$

that is, when

$$
\begin{equation*}
\frac{1}{N}+\frac{1}{N-1}+\ldots+\frac{1}{N-j+1} \geq 1 \tag{3}
\end{equation*}
$$

## Balance Algorithm with Many Bidders



Balance Algorithm

- Euler showed that

$$
\sum_{i=1}^{k} \frac{1}{i} \xrightarrow{k \rightarrow \infty} \log _{e} k
$$

- In other words, by approximating (3), we are looking for $j$ where

$$
\begin{equation*}
\log _{e} N-\log _{e}(N-j)=1 \quad \text { or, equivalently } \quad \frac{N}{N-j}=e \tag{4}
\end{equation*}
$$

## Balance Algorithm with Many Bidders



Balance Algorithm

- In other words, by approximating (3), we are looking for $j$ where

$$
\begin{equation*}
\log _{e} N-\log _{e}(N-j)=1 \quad \text { or, equivalently } \quad \frac{N}{N-j}=e \tag{5}
\end{equation*}
$$

- Solving for $j$ yields

$$
\begin{equation*}
j=N\left(1-\frac{1}{e}\right) \tag{6}
\end{equation*}
$$

## Balance Algorithm with Many Bidders



Balance Algorithm

- Solving for $j$ yields $j=N\left(1-\frac{1}{e}\right)$
- So, the approximate revenue of Balance in this worst-case scenario is BN(1- $\frac{1}{e}$ )
- This translates into a competitive ratio of

$$
1-\frac{1}{e} \approx 0.63
$$

## The Generalized Balance Algorithm

Situation
Advertisers' bids are arbitrary and not just 0 or 1
The following generalization of the Balance algorithm can be shown to have a competitive ratio of $1-\frac{1}{e} \approx 0.63$ :
Generalized Balance Algorithm

- Query $q$ arrives
- Advertiser $A_{i}$ has bid $x_{i}$ for query $q$
- Advertiser $A_{i}$ has fraction $f_{i}$ of his budget left unspent
- Let

$$
\begin{equation*}
\Psi_{i}=x_{i}\left(1-e^{-f_{i}}\right) \tag{7}
\end{equation*}
$$

Then assign $q$ to advertiser $A_{i}$ such that $\Psi_{i}$ is maximum.

## General / Further Reading

## Literature

- Mining Massive Datasets, Section 8.4
http:
//infolab.stanford.edu/~ullman/mmds/ch8.pdf

