Advertising Engines

A Guide to Web Research: Lecture 1

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Stuttgart, Spring 2007



Microsoft® adCenter





new algorithmic problems new models and notions

Talk Objective

Industrial solutions

- Google AdWords
- Google AdSense
- Yahoo! SearchMarketing
- Microsoft adCenter
- Amazon recommendations
- Coming soon: personalized ads for webmail, social networks, blogging platforms, phones, computer games, supermarket bills etc.

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Today we show

- (1) single model for distributing personalized ads
- (2) open algorithmic problems motivated by such systems

Outline

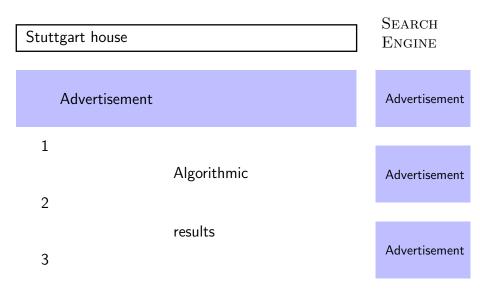
- Architecture of Advertising Engines
 - Component 1: Event
 - Component 2: Advertiser
 - Component 3: Advertising Engine

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- Architecture of Advertising Engines
 - Component 1: Event
 - Component 2: Advertiser
 - Component 3: Advertising Engine
- Algorithmic Challenges
 - Target optimization
 - Click Volume
 - AdRank Computing
 - Ad Coverings

Part I: Architecture of Advertising Engines

Example: Sponsored Search



Example: Context Ads

STUTTGART ESTATE AGENCY

Advertisement

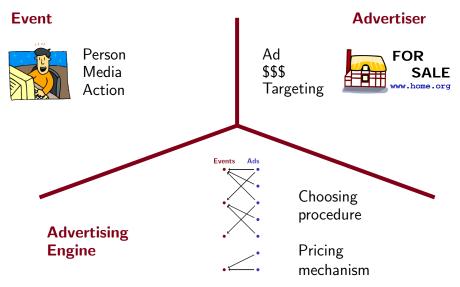
Main

Advertisement

content

 ${\sf Advertisement}$

Three Components: Event, Advertiser, Engine



Component 1: Event



Collect all available information:

- Person: What do we know about him/her?
 - Age, geographic location, previous actions, interests etc
- Media: What is situated around the ad placement?
 - Content and typical audience of website, tv program, newspaper
- Action: Current relations between person and media?
 - Current search query, purchasing a book, signing up to a service

Component 2: Advertiser



Setting new campaign:

- Ad: What will be displayed?
 - Text, image, video, hyperlink, phone number, advertiser's website
- \$\$\$: Size of campaign?
 - Monthly/daily budget, maximal admissible price (bid) for click/impression
- Targeting: Who is target audience?
 - Location, specific query keywords, category of landing page

Targeting in general: any subset of event space $P \times M \times A$

Basic routine of advertising engine:

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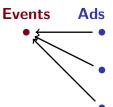
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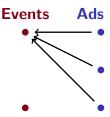
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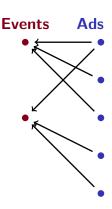
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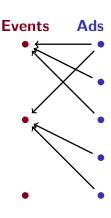
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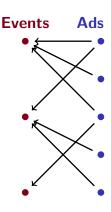
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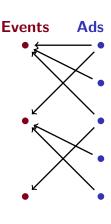
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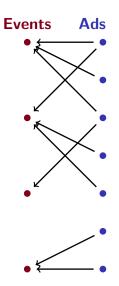
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More objectives?

Part II: Algorithmic Challenges

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Disclaimer: my style is

- At first, think independently (e.g. pose new problems)
- Only after that look into literature

Hence, the following problems might be already known and heavily studied!

Target optimization (1/4)

Informally

Advertiser sets target audience. Advertising engine should help:

- Some potentially interested people are missed
- Exclude people who will be offended by this ad
- Proper setting of target audience is difficult
- Advertising engine knows much more about event space

Target optimization (2/4)

Formalization

Events are vectors

Advertiser provides some **sample** events $S = \{e_1, \dots, e_k\}$ from the target

Advertising engine produces an effective membership procedure for ${\bf optimized\ target\ } \bar{S}$

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How to define **optimized target**?

Target optimization (3/4)

Solution

Let B(e, r) be the ball in event space with center e and radius r

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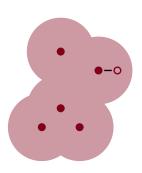
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Target optimization (4/4) Questions Around

- Other definitions for optimized target?
- Exploiting historical information for target optimization
- Target construction based on advertisement content

Informally

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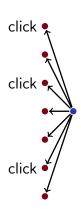
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Take daily history event-ad-?click:

$$(e_1, a_1, b_1) \dots (e_n, a_n, b_n)$$

Use similarity-between-ads function S for computing click volume V:

$$V(a_{new}) = \sum S(a_{new}, a_i) \cdot b_i$$

Basic Formula

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Any comments/objections?

Click Volume (3/4) Corrected Formula

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First step: prediction of click-through rate for a given event-ad pair

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Second step: using click rates

$$V(a_{new}) = \sum_{1 \le i \le n} CTR(e_i, a_{new})$$

Questions Around

- Computing ad volume (the amount of advertisements that can get positive response at the given event)
- Fast algorithm for predicting click volume for all ads in the system
- Exploiting metric inside event space

Input: event e_{new} , set of all ads A. Choosing-ads principles:

- Take the most content-relevant
- Take the ones with best click-through rate
- Take ads with maximal bids

$$AdRank(e_{new}, a) = Bid(a) \cdot (ContRel(e_{new}, a) + CTR(e_{new}, a))$$

AdRank Computing (2/2) Questions Around

Actually, finding content-closest ads to the given input is just the nearest neighbor problem. **We need:**

- Data structure for A for fast computing of best AdRank(e_{new}, a) values
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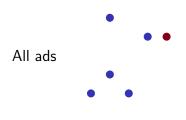
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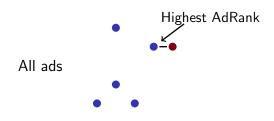


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Ad Coverings

Informally

Consider any **publishers-subscribers** graph (say, RSS feeds):

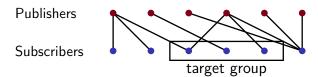
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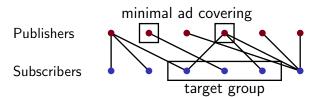


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Other Directions in Advertising Engines

- Optimal ad distribution in case when interested audience is larger than budget
- Machine learning for advertising engines
- Weighted targeting (some events are preferable to others)
- Advertising engines for social networks
- Auction design for sponsored search
- Click fraud

Call for participation

Know a relevant reference?

Have an idea?

Find a mistake?

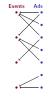
Solved one of these problems?

- Knock to my office 1.156
- Write to me yura@logic.pdmi.ras.ru
- Join our informal discussions
- Participate in writing roadmap-paper

Summary

Three components:







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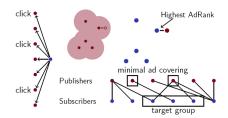




FOR SALE

www.home.org

Four problems:



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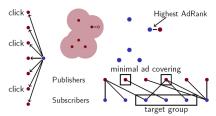




FOR SALE

www.home.org

Four problems:



Vielen Dank für Ihre Aufmerksamkeit! Fragen?

Sources

Course homepage

http://logic.pdmi.ras.ru/~yura/webguide.html



Daniel C. Fain and Jan O. Pedersen

Sponsored Search: a Brief History

http://www.bus.ualberta.ca/kasdemir/ssa2/fain_pedersen.PDF



Alexander Tuzhilin

The Lanes Gifts v. Google Report
http://googleblog.blogspot.com/pdf/Tuzhilin_Report.pdf



Moira Regelson and Daniel C. Fain

Predicting ClickThrough Rate Using Keyword Clusters http://www.bus.ualberta.ca/kasdemir/ssa2/regelson_fain.pdf



Juan Feng, Hemant K. Bhargava and David M. Pennock

Implementing Sponsored Search in Web Search Engines: Computational Evaluation of Alternative Mechanisms

http://research.yahoo.com/node/338/2371



Panel Discussion at SSA2

Models for Sponsored Search: What are the right questions?

http://research.microsoft.com/~hartline/papers/panel-SSA-06.pdf