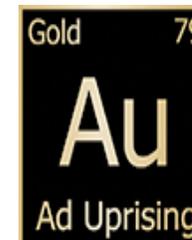


Perspectives on Measuring and Analyzing Online Ads

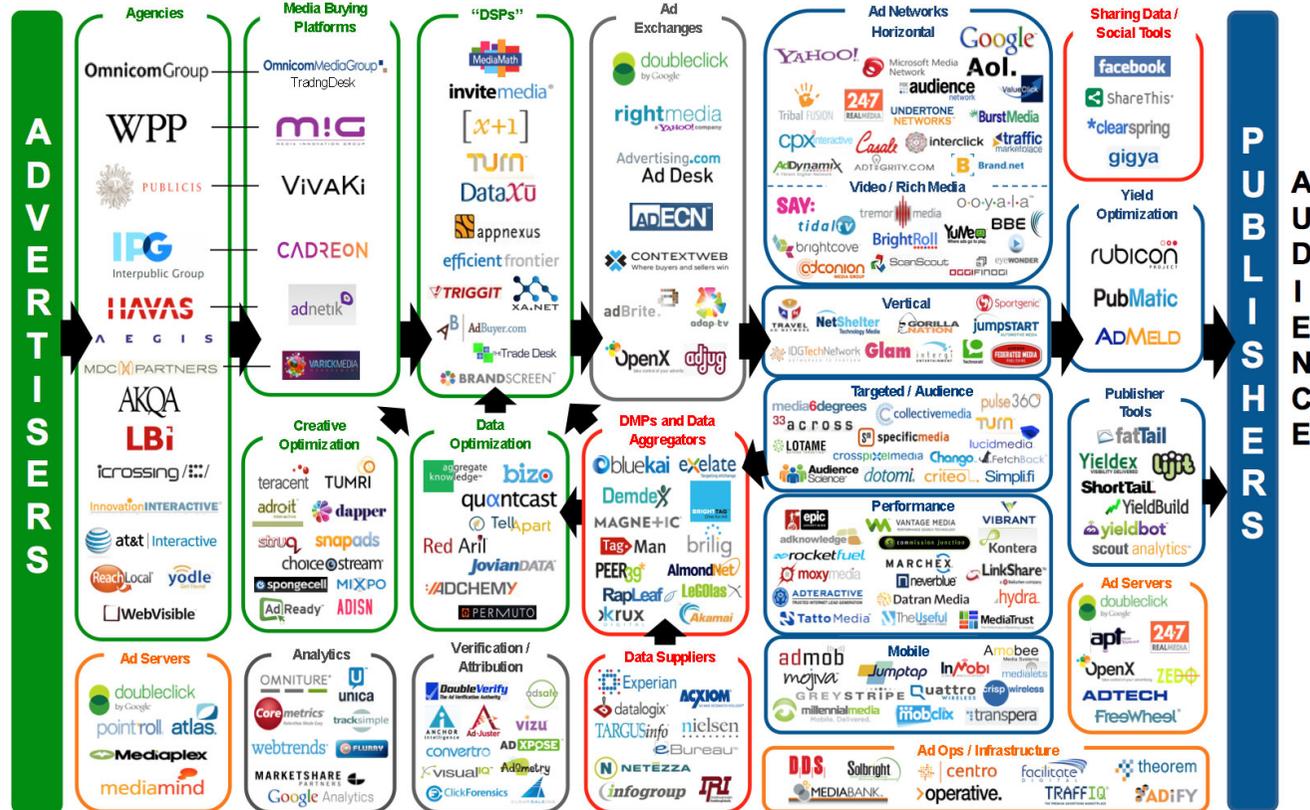
Paul Barford

Spring 2014



Motivation

Display Advertising Technology Landscape



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Outline

- **Overview**
 - Background, objectives and challenges
- **The publisher perspective**
 - m.Labs & PPV nets
- **The user perspective**
 - Ad Uprising & adscape

Objectives

- **Investigate behavior and characteristics of the ad eco-system**
 - **Develop techniques for measuring aspects of the online ad eco-system**
 - **Compile diverse data repositories**
- **Develop new mechanisms that improve performance, yield, security and privacy**
 - **Many opportunities!**
- **Commercial impact**
 - **But it's a cluttered space!**

Publisher revenue deletions

Invalid traffic includes both clicks and impressions that Google suspects to not be the result of genuine user interest

- **Standard means for *valid* traffic - AdWords**
- **Google simply notifies publishers that invalid traffic led to \$XX deduction from you account**
 - Large internal group that monitors traffic quality
- **m.Labs – Web user & traffic quality analytics**
 - Identify invalid impressions and clicks in real time



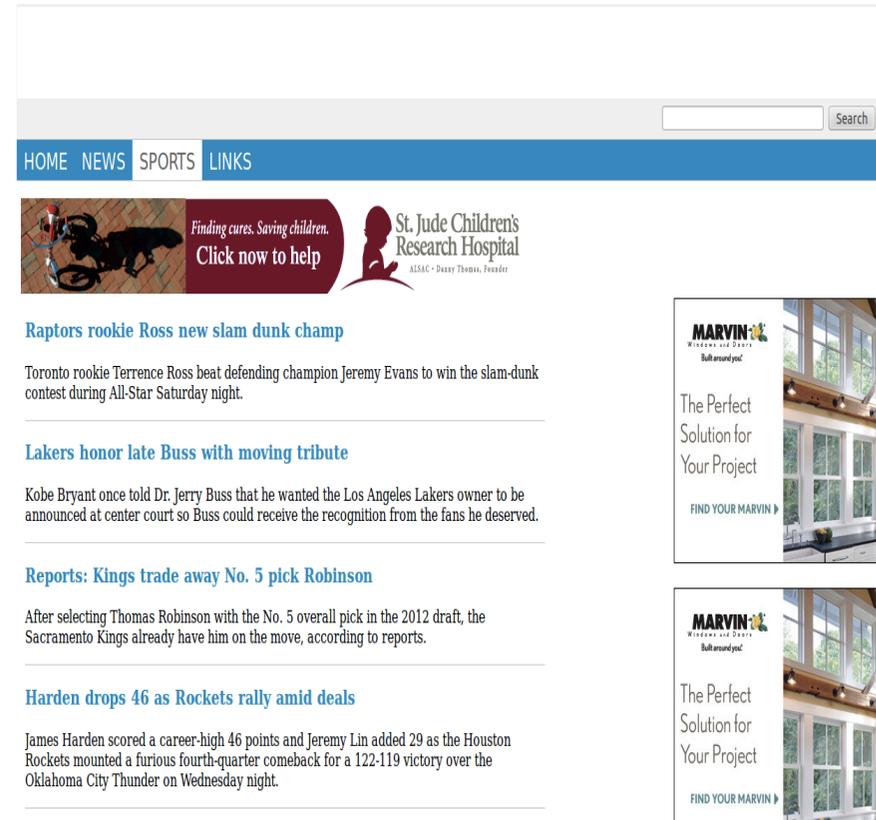
Traffic generation

- **Type “purchase web traffic” in Google**
 - MANY traffic generation offerings
- **Simple threats: script-based page retrieval**
 - Ubiquitous - \$12/10K impressions
- **More complex threats: botnets**
 - Geotargeting, clicks, and other characteristics
 - As much as \$100/10K impressions
- **Pay-per-view networks**
 - Websites that load 3rd party pages in an obfuscated fashion when accessed by users

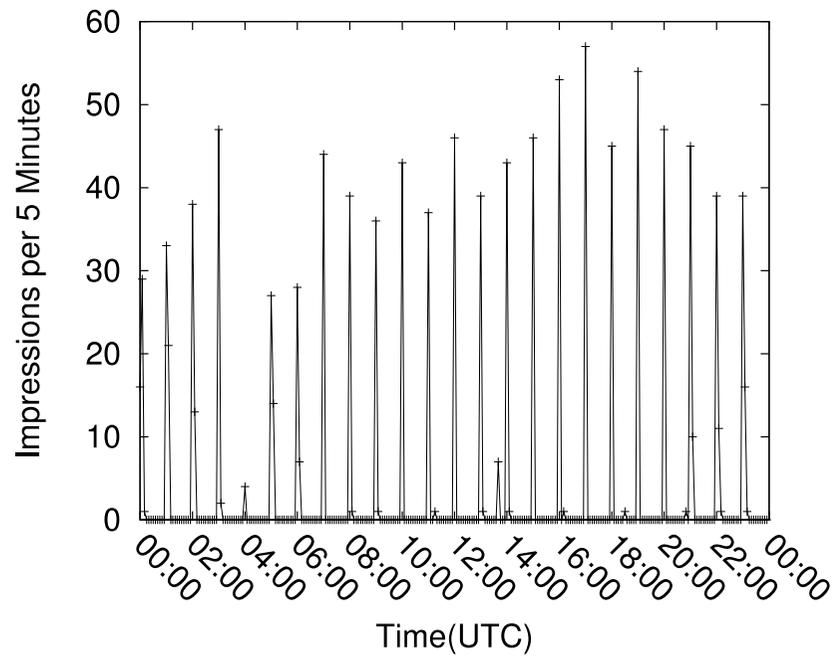


Honeypot websites

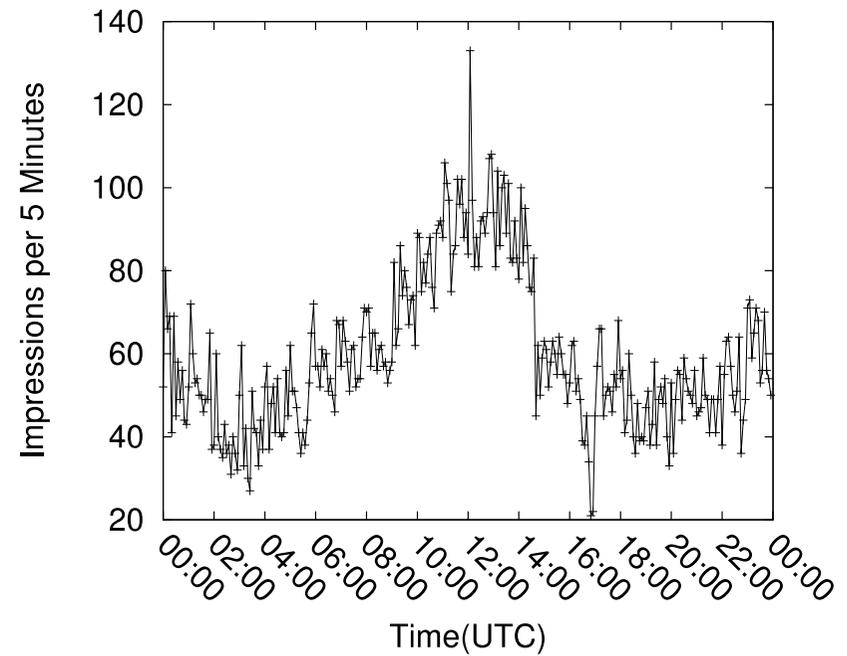
- **Series of websites developed to be targets for traffic generation**
 - Look and feel of a “real” site
- **Instrumentation**
 - Gather as much data per access as possible



Purchased traffic



BuildTraffic arrival process



AeTraffic arrival process



Deep dive into PPV nets

- **When a user accesses a site running a PPV tag a pop-under window is generated**
 - Typically requires a user action
- **Pop-under calls PPV network server**
 - Delivers details on user and site
- **PPV network will deliver URL's of sites buying traffic**
 - Often to 0 height frames
 - Frequent reloads
- **K. Springborn and P. Barford, "Impression Fraud in On-line Advertising via Pay-Per-View Networks" To appear in the USENIX Security Symposium, 2013**



Scope and impact of PPV nets

- **Many PPV sites publish their volume**
 - **Average of 17.16M unique visitors and 6.29B page views per provider per day are claimed**
- **We searched Jan-June '12 Common Crawl DB for PPV tags from 10 providers**
 - **Over 4M PPV tags found on over 11K domains**
- **We used MuStats to estimate daily page views on identified pages**
 - **Over 168M daily page views**
- **Over \$15M/month in wasted ad spend from 10 PPV networks alone!**



Reset: the user perspective

- **Research question: *what display ads are being delivered to users?***
- **Targeting problem: select an ad to deliver to a user accessing a particular web page**
 - Objectives: build awareness, click through
 - Context, geography, placement, behavior, etc.
 - Targeting mechanisms are intrinsic to online ad eco-system
- **AdUprising – Identifying the Internet Adscape**
 - What is being shown where and to whom?

Challenges in Adscape identification

- **Scope – millions of publisher sites**
- **Complexities of ad campaigns**
 - Demo, geo, site lists, caps, etc.
- **Complexities of publisher ad placements**
 - Premium, exchanges, backfill, etc.
- **Complexities of ad delivery and targeting mechanisms**
 - This is what we're seeking to understand

Building a display ad crawler

- **We seek to understand ad delivery by harvesting ads from a broad set of sites**
- **Ad crawler requirements**
 - Distinguish and collect ads from other images
 - Collect related data
 - Accommodate gigantic scale and highly dynamic nature of ads in a gentle fashion
 - Personalization
- **We developed a scalable, profile-based ad crawler based on Firefox/firefly**

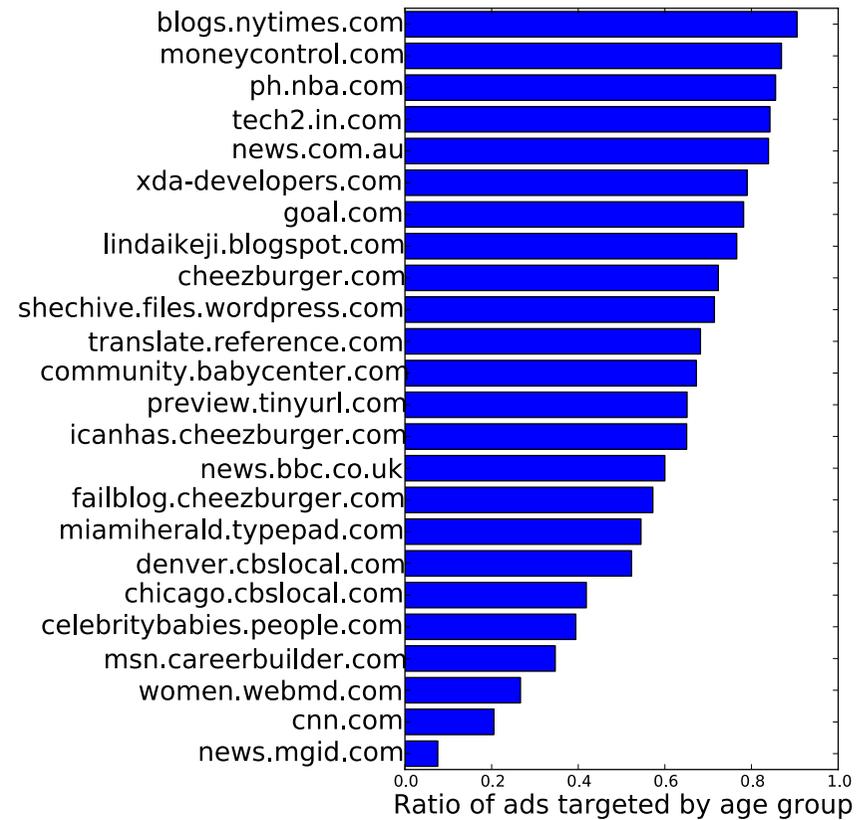
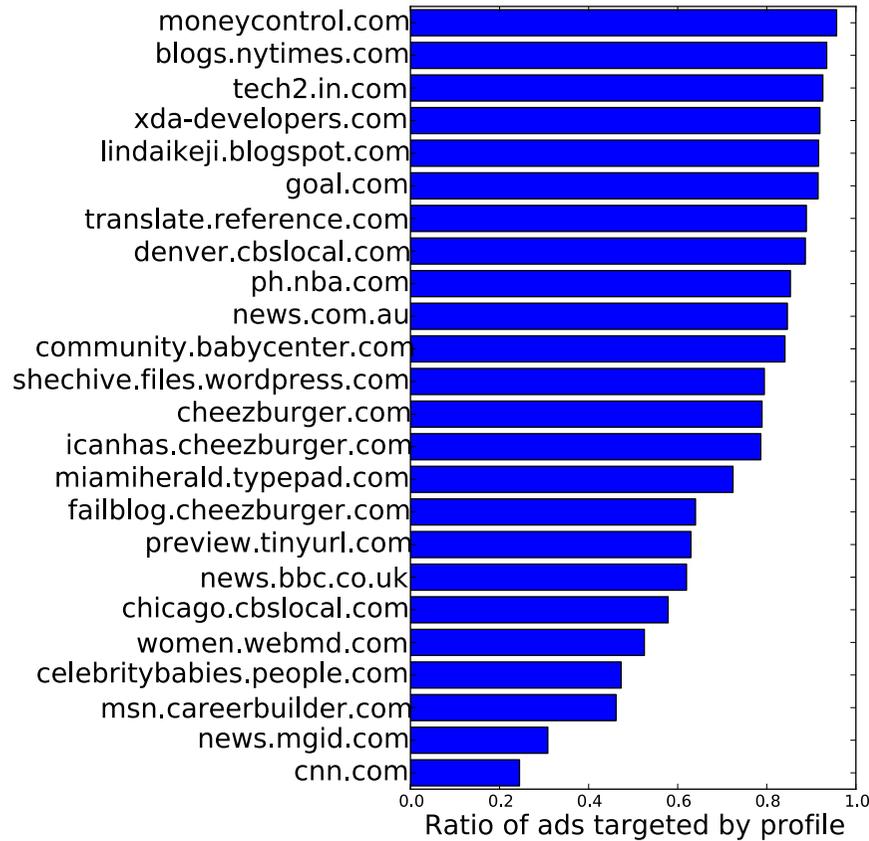
Building user profiles

- **We assume *profiles* can be built based on browsing careful selection of sites**
 - Assume single interest users
- **Site selection is base on Alexa categories**
- **Profiles are created by browsing top 100 websites from an Alexa category**
 - Profiles are basis for ad collection
- **Profile maintenance is a challenge when gathering ads from different sites**
 - We find that profiles change significantly

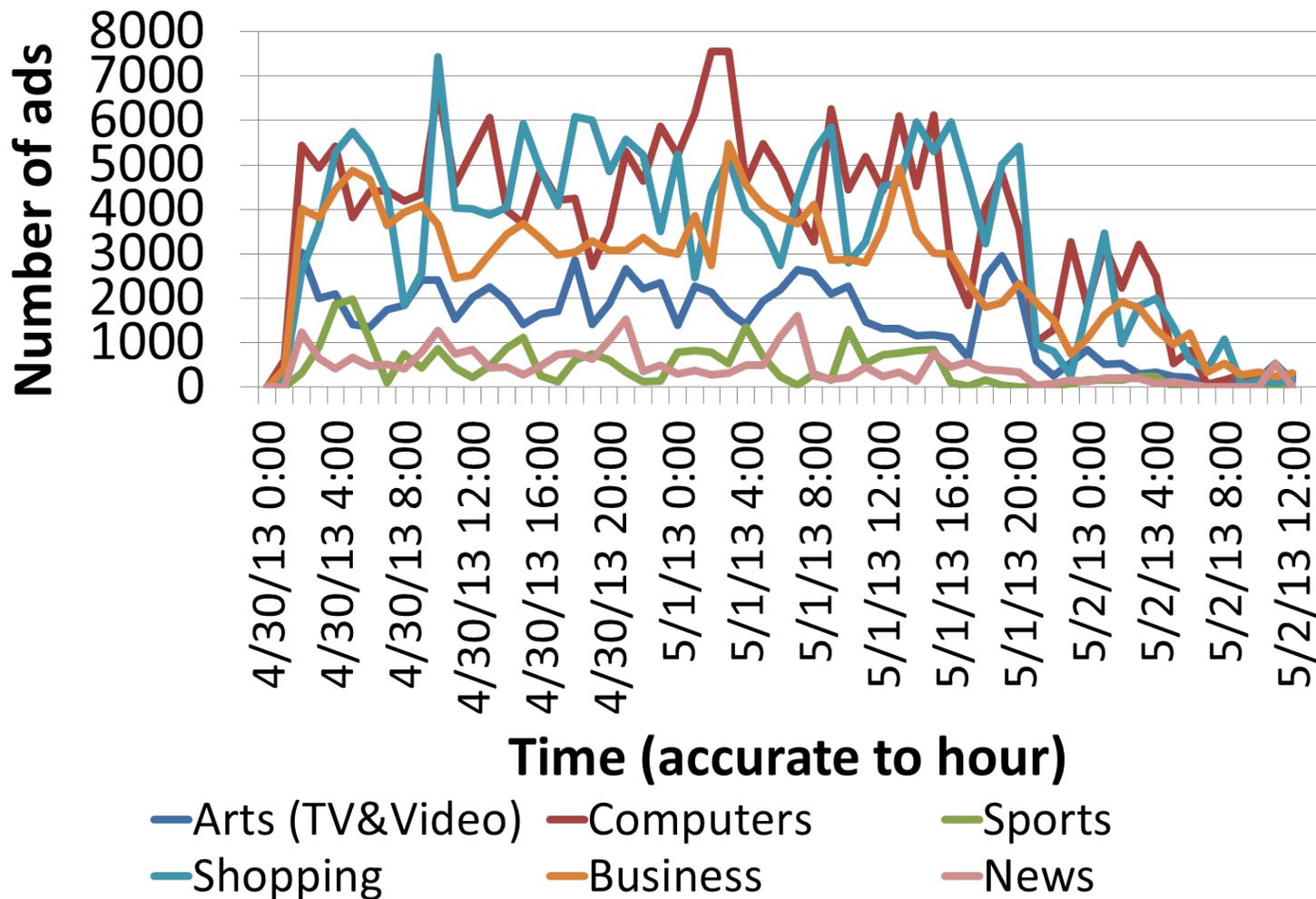
Collecting ads

- **Objective: gather as many distinct ads as possible**
- **Simplifications**
 - **Single collection site (no geo diversity)**
 - **No consideration of time varying effects**
 - **Limited crawl to fixed period and fixed sites**
- **134 sites crawled using 60 different profiles**
 - **462 [website, profile] pairs**
- **Over 175K unique ads gathered from ~3.5K entities**

Ad targeting



Ads impressions by category

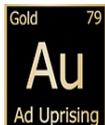


Profiles vs. ad categories

	Arts	Business	Computers	Games	Health	Home	Kids_and _Teens	News	Recreation	Reference	Science	Shopping	Society	Sports
Arts	0.098	0.239	0.199	0.037	0.031	0.016	0.048	0.026	0.022	0.038	0.000	0.187	0.029	0.030
Business	0.099	0.299	0.145	0.015	0.018	0.011	0.024	0.044	0.015	0.083	0.000	0.173	0.038	0.036
Computers	0.125	0.099	0.415	0.021	0.012	0.008	0.031	0.007	0.012	0.030	0.000	0.193	0.025	0.022
Health	0.070	0.164	0.181	0.020	0.083	0.014	0.017	0.033	0.027	0.101	0.001	0.199	0.065	0.025
Home	0.102	0.230	0.287	0.002	0.011	0.030	0.003	0.017	0.034	0.100	0.010	0.123	0.032	0.019
News	0.070	0.313	0.197	0.014	0.038	0.008	0.003	0.050	0.037	0.060	0.002	0.116	0.037	0.056
Recreation	0.135	0.204	0.199	0.006	0.014	0.012	0.005	0.037	0.086	0.039	0.001	0.154	0.039	0.070
Reference	0.058	0.171	0.235	0.009	0.021	0.024	0.007	0.007	0.030	0.163	0.002	0.168	0.102	0.003
Shopping	0.106	0.132	0.231	0.004	0.012	0.005	0.004	0.012	0.013	0.036	0.000	0.409	0.021	0.015
Sports	0.087	0.207	0.249	0.001	0.025	0.012	0.012	0.017	0.032	0.108	0.001	0.158	0.029	0.063

Alexa profiles in rows, ad categories in columns, percentage of ads shown in cells

Barford et al. "Adscape: Harvesting and Analyzing Online Display Ads" Under submission to ACM IMC, 2013.



Summary and status

- **Fraud is a gigantic problem in online advertising**
 - **From simple scripts to sophisticated bots to PPV networks**
- **m.Labs has developed filters to identify fraud**
- **m.Labs open experimental platform**
 - **Data repository and API for ad fraud detection**
- **The Internet Adscape is huge and diverse**
- **Initial study to characterize the Adscape**
- **Ad Uprising: ongoing data collection and analysis**

Thank you!

- **Igor Canadi**
- **Darja Krushevskaja**
- **Qiang Ma**
- **Muthu**
- **Kevin Springborn**
- **Charles Thomas**