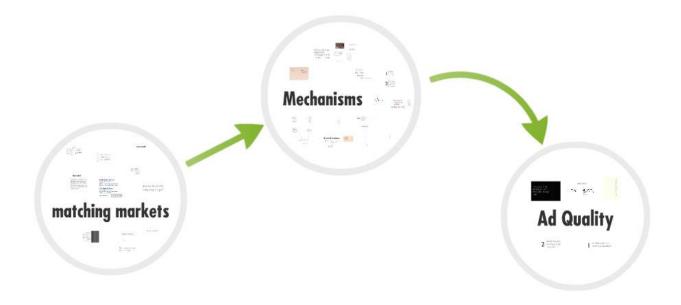


#### Sponsored Search Markets

hari sundaram hs1@illinois.edu how one makes money on the web









SUTHEBYS

Web search

**Game Theory** 

**Auctions** 







Display Ads



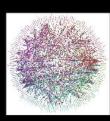
Behavioral targeting



**Recommender systems** 



**Privacy** 



**Networks** 



**Emerging areas** 



**Final Presentations** 

## bonsored Search advertizing makes most of Google's revenue Markets

hari sundaram hs 1@illinois.edu

how one makes money on the web





fact: keyword based advertizing makes most of Google's revenue

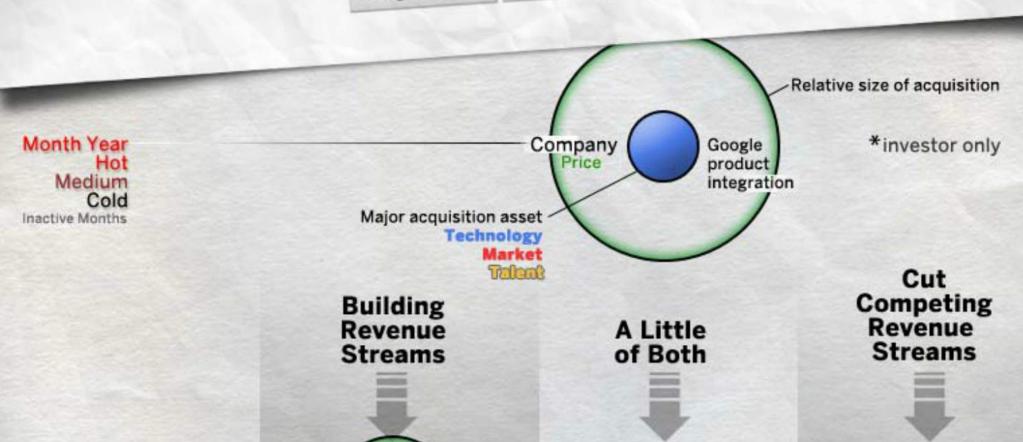
## Google (graphic)

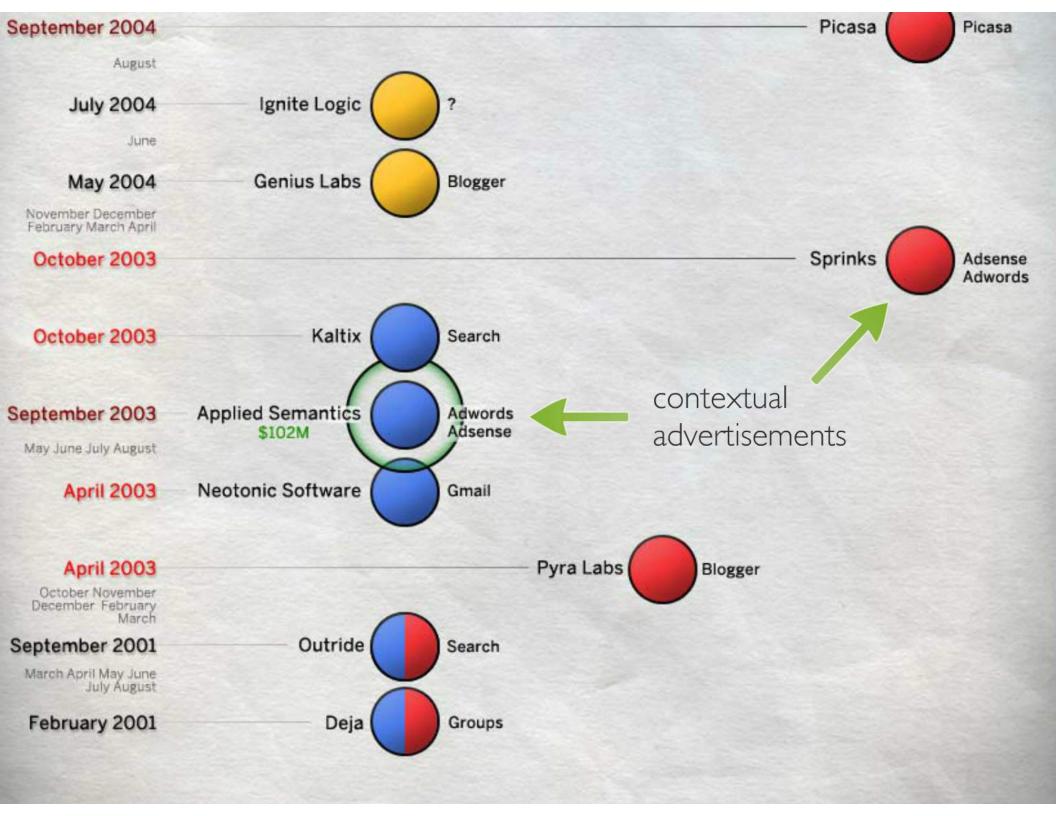
Google's Acquisition Appetite

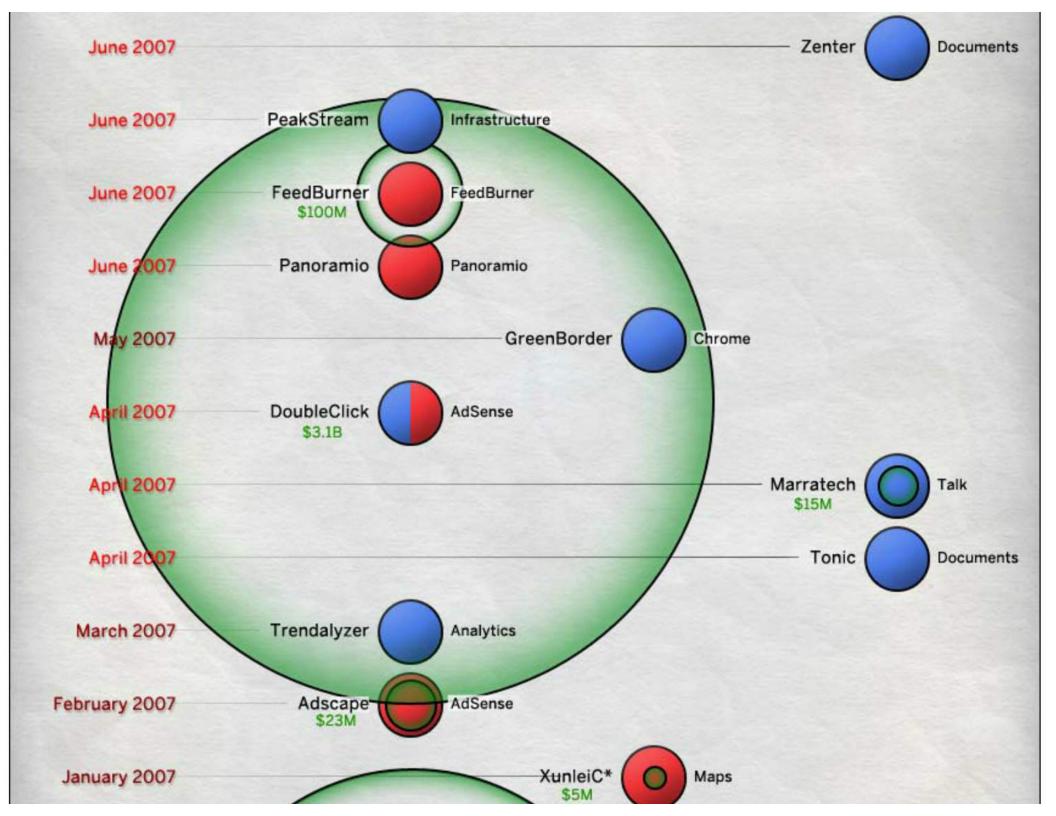
August 2010

Google Search

I'm Feeling Lucky







#### Alaska Travel Information

www.alaskatravel.com 1 (877) 484 5851 Alaska Vacation Packages, Railroad,

Hotels & Cruises. Free Guide & Map.

75% off Alaska Cruises

www.alaskancruise.com

Huge selection of discounted Alaska cruises. All cruise lines.

See your ad here »

paying per click



## paying per click

# are auctions the only way to go?

#### highest click through rate

lets assume that we have a list of slots that need to be filled



75% off Alaska Cruises
www.alaskancruise.com
Huge selection of discounted Alaska
cruises. All cruise lines.

See your ad here »

# we need to make some assumptions

Advertisers know the click through rates.

We assume that the click through rate depends only on the slot itself and not on the ad that is placed there.

3

Third, we assume that the click through rate of a slot also doesn't depend on the ads that are in other slots.

We assume that each advertiser has a revenue per click: the expected amount of revenue it receives per user who clicks on the ad.

4

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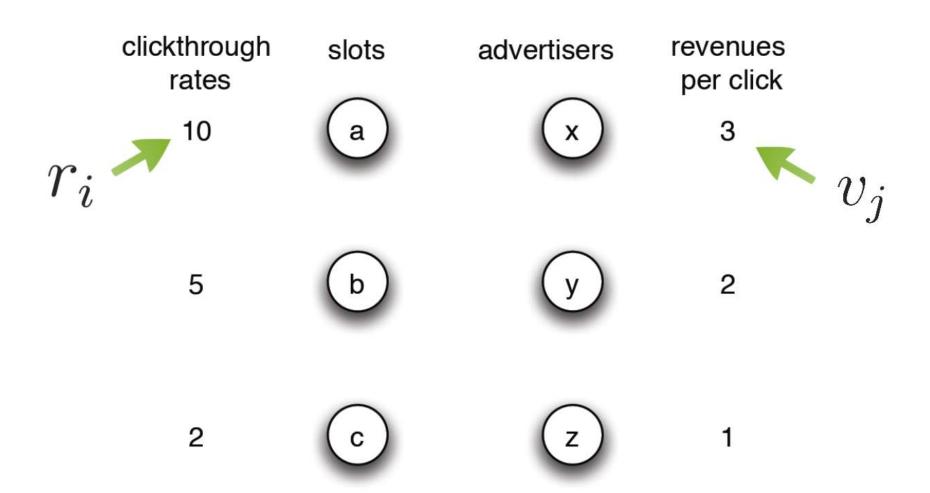
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## we need to make some assumptions

### a basic model



#### a basic model

clickthrough rates $r_i$	slots	advertisers	revenues per click 3	$v_{i}$
5	b	У	2	J
2	С	Z	1	

$$v_{ij} = r_i v_j$$

slots

advertisers

valuations

(a)

(x)

30, 15, 6



 $v_{ij} = r_i v_j$ 

(b)

У

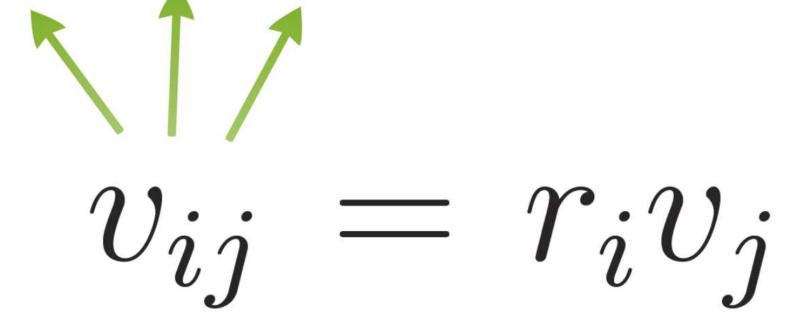
20, 10, 4

 $\bigcirc$ 

 $\left(z\right)$ 

10, 5, 2

## 30, 15, 6

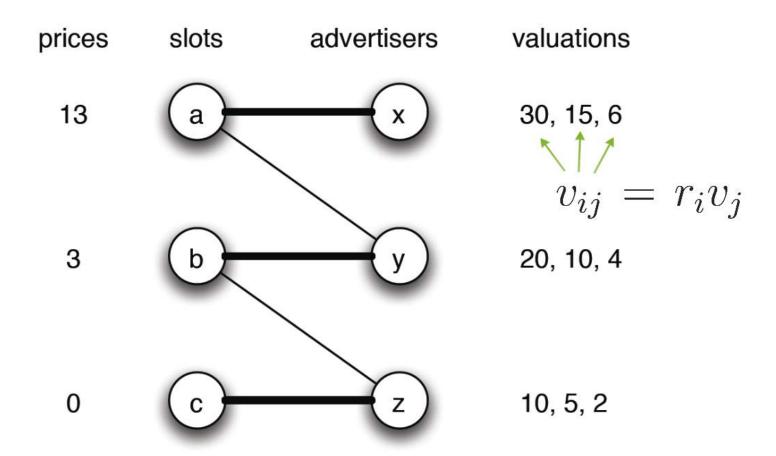


# Each seller i announces a price

Each buyer j evaluates her payoff for choosing a particular seller i

 $v_{ij} - p_i$ 

### Build a preferred-seller graph

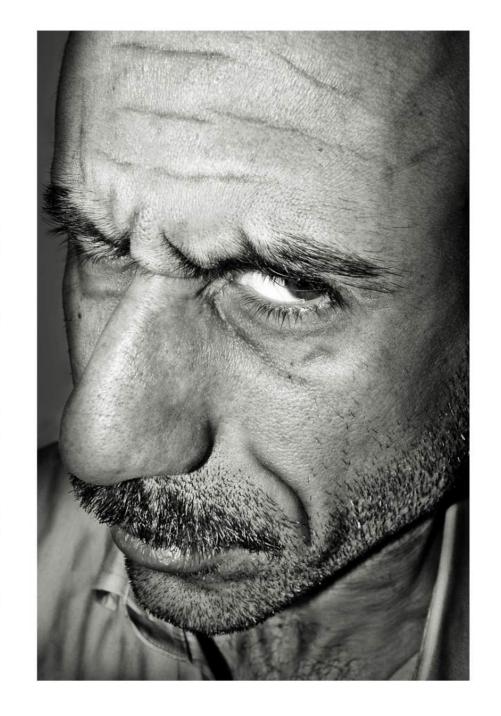


The prices are market-clearing if this graph has a perfect matching

# The valuation function can be arbitrary

$$v_{ij} = r_i v_j$$

In reality, search engines do not know the valuations of the advertisers



In the early days, we saw variants of first price auctions: advertisers were simply asked to bid for a slot



#### VCG auctions are designed to encourage truthful bidding generalize single item second price auctions

The second-price auction produces an allocation that maximizes social welfarethe bidder who values the item the most gets it

lets assume that we have a set of valuations for one item

$$v_1, v_2, v_3, \ldots, v_n$$

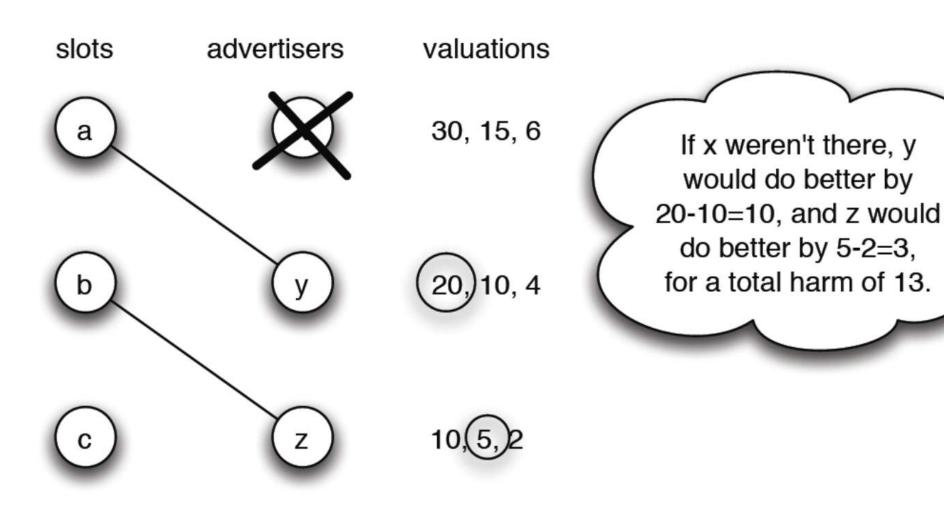
the winner of the auction is charged an amount equal to the "harm" he causes the other bidders by receiving the item

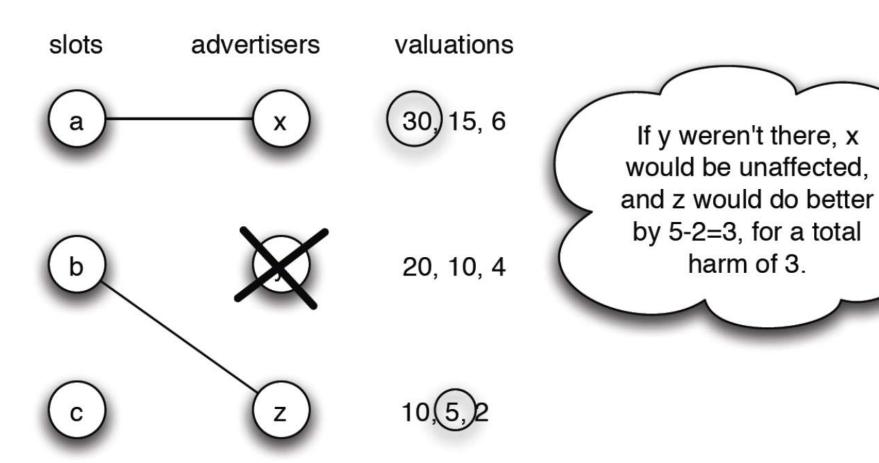
each individual is charged a price equal to the total amount better off everyone else would be if this individual weren't there



we first assign items to buyers so as to maximize total valuation

the price buyer j should pay for seller i's item is the harm she causes to the remaining buyers through her acquisition of this item.





VCG for the more general case

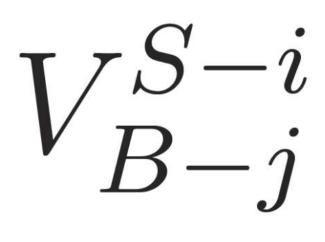


 ${\bf social\ optimum-maximizes\ valuation}$ 

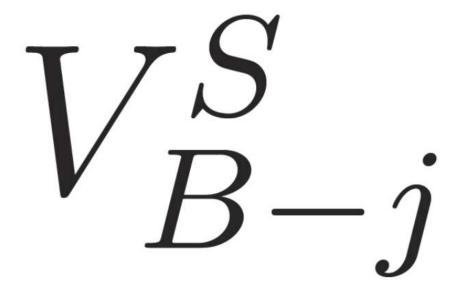
the set of sellers with seller i removed

the set of buyers with buyer j removed

removing item i and seller j, then the best total Vvaluation the rest of the buyers



if buyer j simply didn't exist, but item i were still an option for everyone else, then the best total valuation the rest of the buyers







The VCG price that we charge to buyer j for item i is then:

$$p_{ij} = V_{B-j}^S - V_{B-j}^{S-i}$$

price

when j is absent

when j buys item i

### the overall mechanism

Ask buyers to announce valuations for the items.

Choose a socially optimal assignment of items to buyers

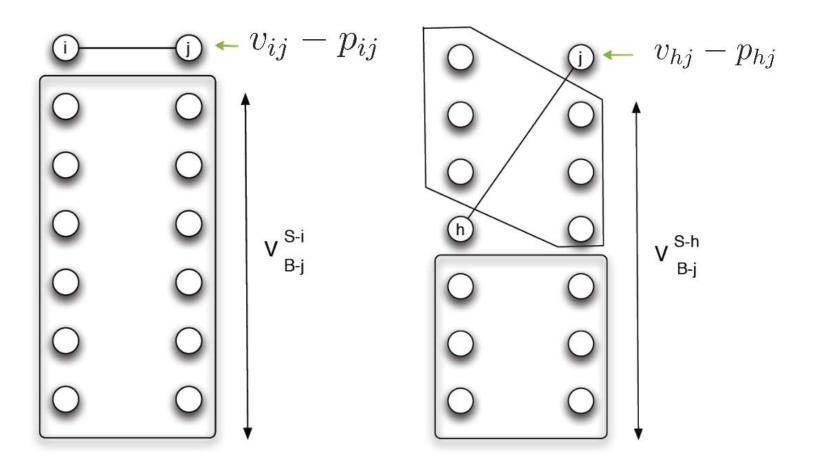
Charge each buyer the appropriate VCG price

## Claim:

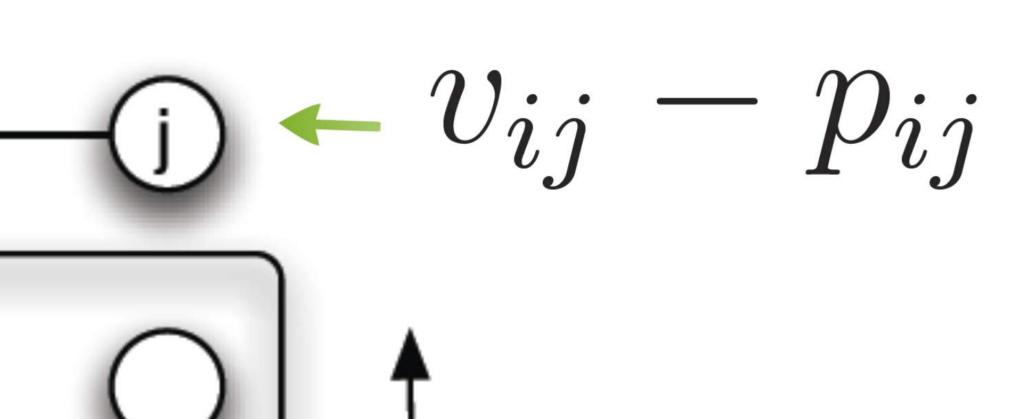
Truth telling is a dominant strategy

#### j tells the truth

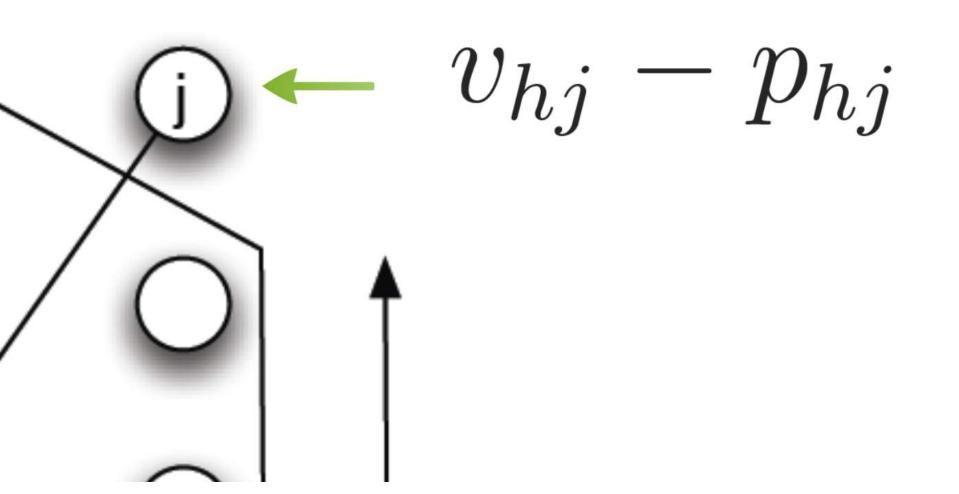
#### j lies



#### reno mie mani



## jlies



#### what we have to show:

$$v_{ij} - p_{ij} \ge v_{hj} - p_{hj}$$

truthtelling

$$v_{ij} - [V_{B-j}^S - V_{B-j}^{S-i}] \ge v_{hj} - [V_{B-j}^S - V_{B-j}^{S-h}]$$



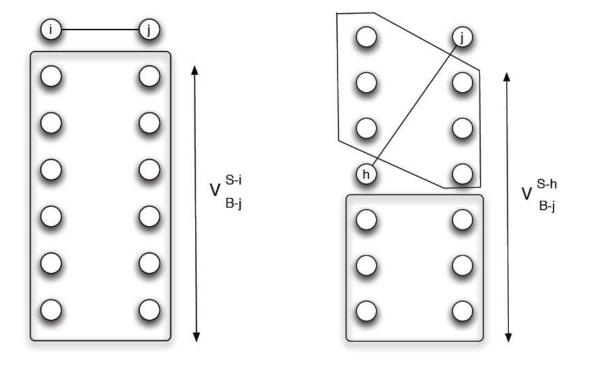
$$v_{ij} + V_{B-j}^{S-i} \ge v_{hj} + V_{B-j}^{S-h}$$

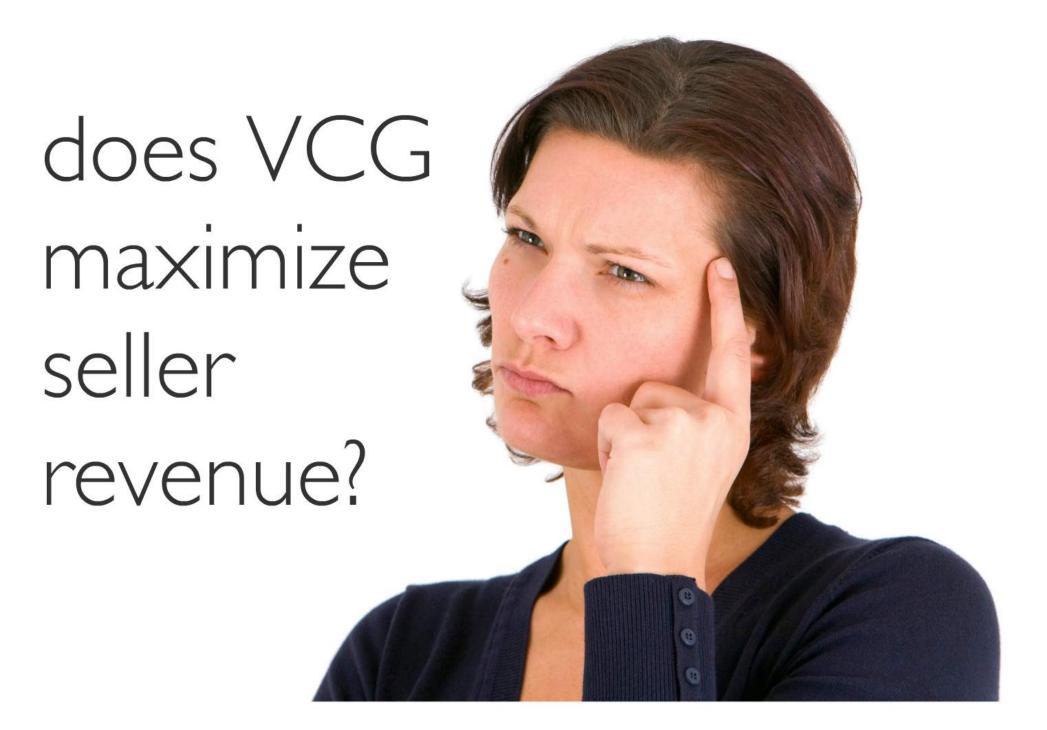


$$v_{ij} + V_{B-j}^{S-i} = V_B^S$$

#### valuation with a constraint

$$v_{hj} + V_{B-j}^{S-h} \le V_B^S$$

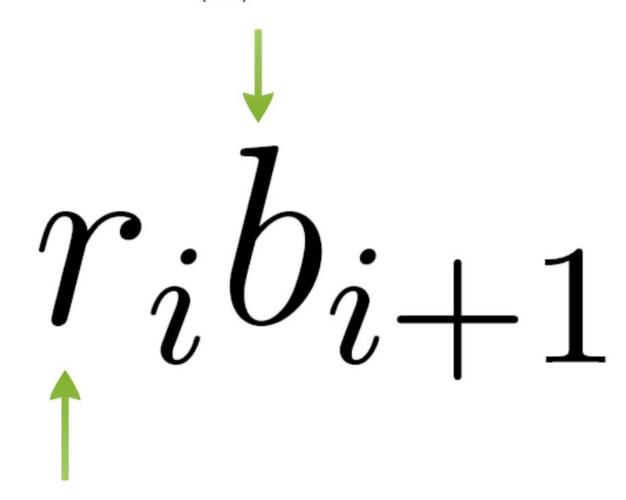




Each advertiser announces a bid consisting of a single number bi the price it is willing to pay per click

Then, after each advertiser submits a bid, the GSP procedure awards each slot i to the ith highest bidder, at a price per click equal to the (i + I)st highest bid.

(i+1)th bid



click through rate

# there can be multiple Nash equilibria

there can be multiple Nash equilibria

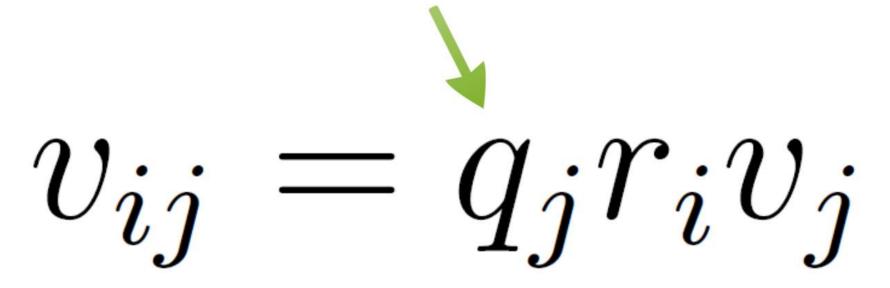
truth-telling may not be one of them!

Search engine revenue can be greater than VCG, but it depends on the equilibrium!

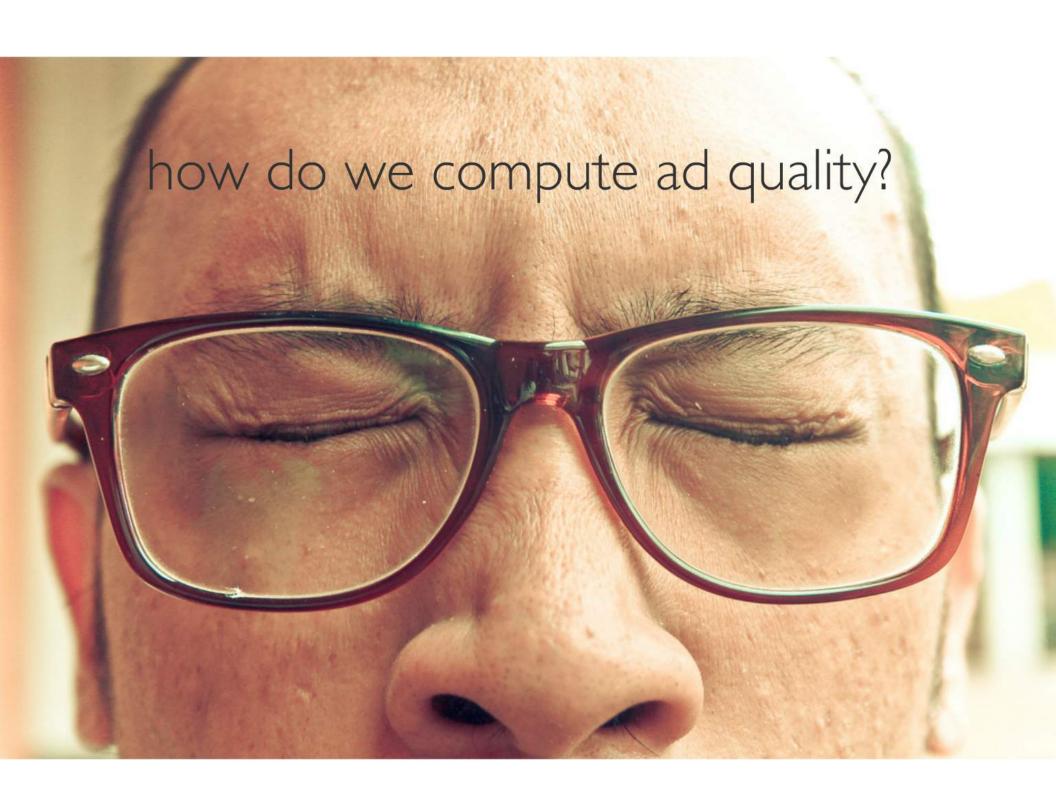
how good is the assumption of a fixed click-through rate?



#### quality factor



hidden from bidders



# auctions are now more complicated!

# Search engines now have a lot of power!