An Auctioning Reputation System Based on Anomaly Detection

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Online Auctioning

- Huge volume: eBay hosted 440,100,000 new listings in Q2 2005
- In this talk: trustworthiness of online auctioning
- Why do we buy in an online auction?

A. to find a rare/collectable item

- B. to find a bargain; commodity at a "good" price
- eBay financial report (expected 2005):
 - Clothing & Accessories --- \$3.3 billion (2nd)
 - Consumer Electronics --- \$3.2 billion (3rd)
 - Computers
 --- \$2.9 billion (4th)

Data suggests that most people use eBay to find bargains

- Inherently untrustworthy environment:
 - Pseudonymous sellers
 - Pseudonymous buyers
 - Delivery? Warranty? Quality?
- Reputation system: a tool to establish trust



G Great seller, extra quick shipping thanks A++++

- eBay's reputation system provides little help
 - Based on feedback: vulnerable to "poisoning" attack

ANOTHER GREAT DEAL !! I WILL BE BACK !!

G WORKS GREAT !! GREAT DEALER!! My GOD I Can't SPELL!!

🚯 WORKS GREAT !! DREAT DEALLER!!

🕒 WORKS GREAT !! DREAT DEALLER!!

🔂 WORKS GREAT !! DREAT DEALLER!!

- eBay's reputation system provides little help
 - Based on feedback: vulnerable to "poisoning" attack
 - Does not provide information on price



- eBay's reputation system provides little help
 - Based on feedback: vulnerable to "poisoning" attack
 - Does not provide information on price
 - Does not differentiate among the majority of sellers

Goals

- Alice—a buyer, Bob—a seller
- Develop a trustworthy mechanism that helps Alice:
 - Achieve her goal: what are the chances that Alice can find a bargain in Bob's auctions?
 - Warn her from fraudulent activities: are the prices in Bob's auctions artificially inflated?
 - Provide her assurance against poisoning attack: why should Alice trust the mechanism?

Contributions

- A reputation system that helps buyers avoid sellers who seem to be inflating prices
 - Formulated the "seem to be inflating prices" as an anomaly detection problem
 - Business level anomaly detection: the basic events are auctions, bidding.
 - Behavioral system: based on how human behave/act rather than on people feedback.
- Only a first step, some goals still ahead

Outline

- Motivation: find a bargain and avoid fraud
- Contributions: anomaly detection system to identify price inflating sellers:
 - The N model
 - The M model
 - The P model
- Case studies

Auctioning 101

- Pseudonymous sellers and bidders
- Auctions end after a predefined time (e.g., 7 days)
- Highest bid wins
- Seller sets minimum starting bid
- Shilling: a group of bidders that place fake bids to inflate the final price

Methodology

- Collect data from eBay
 - three weeks of data in the category: Laptop Parts & Accessories
 - 127,815 auctions, 12,331 sellers,
 - 604 high-volume sellers: posted more than 14 auctions controls 60% of the market
- Use statistical model to predict seller behavior
 - 95% of the sellers are "normal"
 - 5% are abnormal, or suspicious

Step 1: Average Number of Bids

- What is an indication that prices are high?
 - high number of bids
- Goal: identify sellers with abnormally high number of bids



 95% of high-volume sellers have less than 7 bids per auction

•Model is insensitive to supply: number of auctions posted by a seller

Step 1: The N Model



Correlation: many auctions implies low number of bids

Suspicious seller: one that posts many auctions and still attracts many bids

Outline

- Motivation: find a bargain and avoid fraud
- Contributions: anomaly detection system to identify price inflating sellers:
 - The N model: a seller is suspicious if they post many auctions that attract many bids
 - The M model
 - The P model
- Reputation example

Step 2: Average Minimum Starting Bid

- Legitimate explanation for high number of bids: low minimum starting bid
- Goal: identify sellers with abnormally high number of bids and high minimum bid
- Problem: how do you know that the minimum bid is high?

| Relative minimum | winning_bid – minimum_bid |
|------------------|---------------------------|
| bid (RMB) | = winning_bid |

Step 2: The M Model



Correlation: low minimum starting bid implies many bids

M suspicious seller: starts with high minimum bid and attracts many bids

M+N suspicious seller: posts many auctions, attracts many bids, starts with high minimum bid

Step 3: Bidders' Profile of a Seller

- Fraudulent explanation for high number of bids: shilling
- Goal: identify group of bidders that repeatedly bid and lose in a seller's auctions
- Suspicious seller:
 - N: sellers with abnormally high number of bids and
 - M: high starting bid **and**
 - P: has a group of bidders that repeatedly bid and lose

Bidder Presence Curve



Bidder Presence/Win Curves



Bidder Presence/Win Curves (Normal case)



10% of the bidders participated in 20% of the auctions and won 20% of the times

Outline

- Motivation: find a bargain and avoid fraud
- Contributions: anomaly detection system to identify price inflating sellers:
 - The N model: a seller is suspicious if they post many auctions that attract many bids
 - The M model: a seller is suspicious if they attract many bids and start with high minimum bid
 - The P model: a seller is suspicious if they have a group of bidders that repeatedly participate and lose
- Reputation example



Results Summary

- 54 sellers classified as abnormal with respect to at least one model
- 3 sellers classified as abnormal with respect to all three models
- No confirmed fraud



Summary

- Trust: do we get what we expected?
- Reputation system as anomaly detection
 - Attempt to identify price inflation
 - Work at the business level
 - Consider poisoning attack (see paper)

Thank you. Questions?