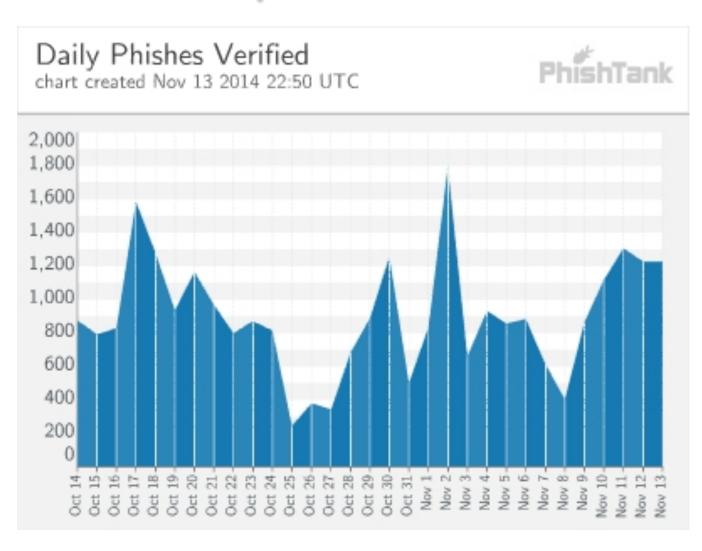
# CS259D: Data Mining for Cybersecurity

# Phishing

- Goal:
  - Account information
  - Logon credentials
  - Identity information
- Attack vectors:
  - Legitimate-looking emails
  - Legitimate-looking websites

# Scale of the problem



#### Detection

- Toolbars
  - Spoofguard
  - Netcraft
- Email filtering
  - Examples:
    - SpamAssassin
    - Spamato
  - Advantages:
    - More complete context (content, headers, etc.)
    - Completely shield user from decision-making process

- IP-based URLs
  - Example:
    - http://192.168.0.1/paypal.cgi?fix account
  - Compromised PCs with no DNS entries
  - Binary feature
- Age of linked-to domain names
  - Registered legitimate-sounding domain names
    - Example: playpal.com, paypal-update.com
  - Typically short life-span
    - Registered using stolen credit cards, canceled by registrar
    - Domain caught by anti-phishing monitors
    - Often lasting only ~ 48 hours
  - Obtained using a WHOIS query
  - Binary feature: Lifetime < 60 days</li>

- Non-matching URLs
  - Example: <a href="badsite.com"> paypal.com</a>
  - Binary feature: URL text different from HREF
- "Here" links to non-modal domain
  - Example: Click here to restore your account
  - Modal domain: domain most frequently linked to
  - Binary feature: link with text "link", "click", "here" that links to a domain other than modal domain

- HTML emails
  - Binary feature: email section with MIME type text/html
- Number of links
  - Numeric feature: # links in HTML part(s) of email
  - Link defined by an <a> tag with href attribute
    - Including mailto: links

- Number of domains
  - Domain names for URLs starting with http/ https
  - Only the main part of the domain name
    - What registrar gets paid for
      - Not necessarily same as combination of top- & 2nd-level domain
    - Example:
      - university.edu for www.cs.university.edu
      - company.co.jp
        - Top-level: .jp, second-level: .co
  - Numeric feature: #distinct domains

- Number of dots
  - Subdomains:

http://www.my-bank.update.data.com

Redirection script:

http://www.google.com/url?q=http://www.badsite.com

- Looks to naïve user to be from google.com
- Redirects browser to badsite.com
- Numeric feature: Maximum number of dots in any of the links in the email

- Contains javascript
  - Binary feature: string "javascript" appears in email
- Spam filter output
  - Binary feature: class assigned to email by SpamAssassin

- 10-fold cross validation
- Classifier: Random forest
  - 10 decision trees
  - Each decision made on a random attribute
  - Trees pruned

- SpamAssassin ham corpora
  - ~6950 non-phishing non-spam
- Publicly available phishingcorpus
  - ~ 860 phishing messages
  - Challenge with WHOIS queries
    - Only 505 domains out of 870 domains
    - Increases false negative rate

Feature	Non-Phishing Matched	Phishing Matched
Has IP link	0.06%	45.04%
Has "fresh" link	0.98%	12.49%
Has "nonmatching" URL	0.14%	50.64%
Has non-modal here link	0.82%	18.20%
Is HTML email	5.55%	93.47%
Contains JavaScript	2.30%	10.15%
SpamAssassin Output	0.12%	87.05%

Feature	$\mu_{ ext{phishing}}$	$\sigma_{ m phishing}$	$\mu_{ ext{non-phishing}}$	$\sigma_{ m non-phishing}$
Number of links	3.87	4.97	2.36	12.00
Number of domains	1.49	1.42	0.43	3.32
Number of dots	3.78	1.94	0.19	0.87

	•	
Classifier	False Positive Rate $fp$	False Negative Rate $fn$
PILFER, with S.A. feature	0.0013	0.036
PILFER, without S.A. feature	0.0022	0.085
SpamAssassin (Untrained)	0.0014	0.376
SpamAssassin (Trained)	0.0012	0.130

#### Review of TF-IDF

- Measure importance of word in document
- TF = frequency of word in document
- IDF = measure popularity of word in corpus
  - Log(N/#{documents having the term})
- tf-idf  $(t, d, D) = tf(t, d) \times idf(t, D)$

# Robust hyperlinks

- Lexical signatures for identifying URLs
- Signature words chosen using TF-IDF
- Experiments: 5 terms enough for unique page identification

#### Observation

- Minimal changes to original page detectable via robust hyperlinks
- Phishing sites often include brand names
  - Common on brand's webpages
  - Rare on the web

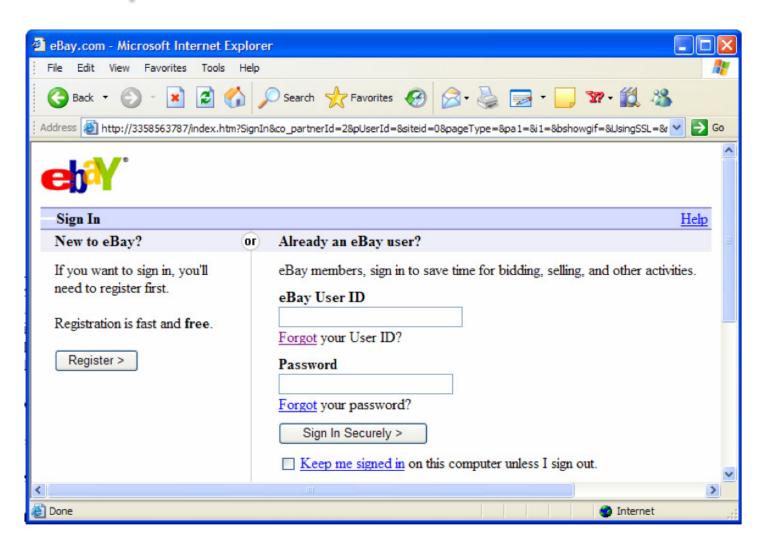
# Algorithm

- Compute term TF-IDFs
- Find top 5 terms
- Submit terms as query to Google
- Check if domain is among top-N results
- Assumption: phishing pages have low pagerank

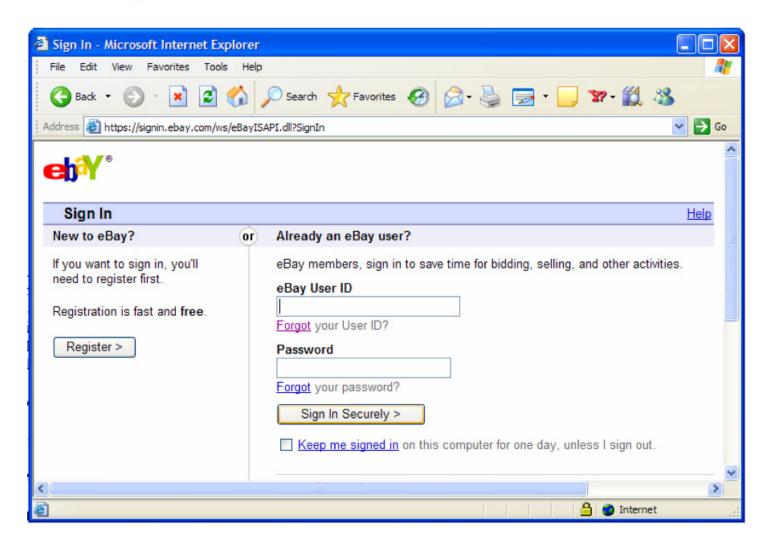
# Lowering false positives

- Include domain name in lexical signature
- Heuristic: Zero results Means Phishing

## Example



# Example



# Example

 Top terms: eBay, user, sign, help, forgot

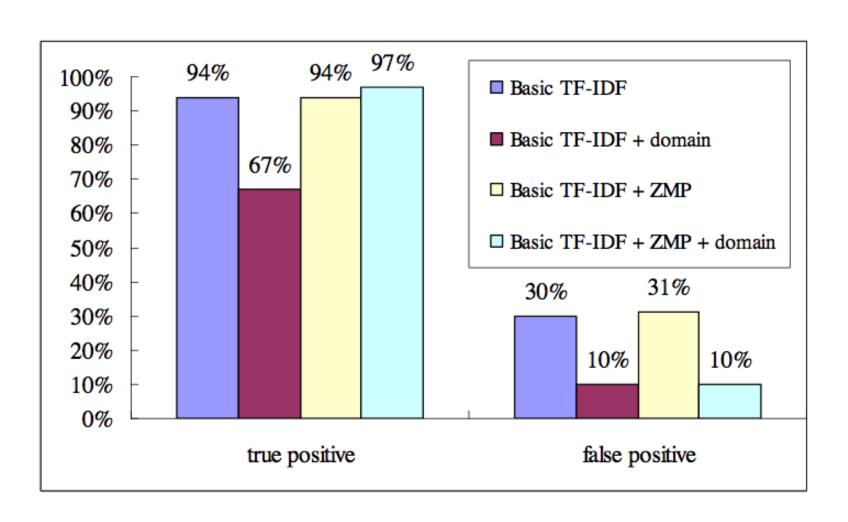


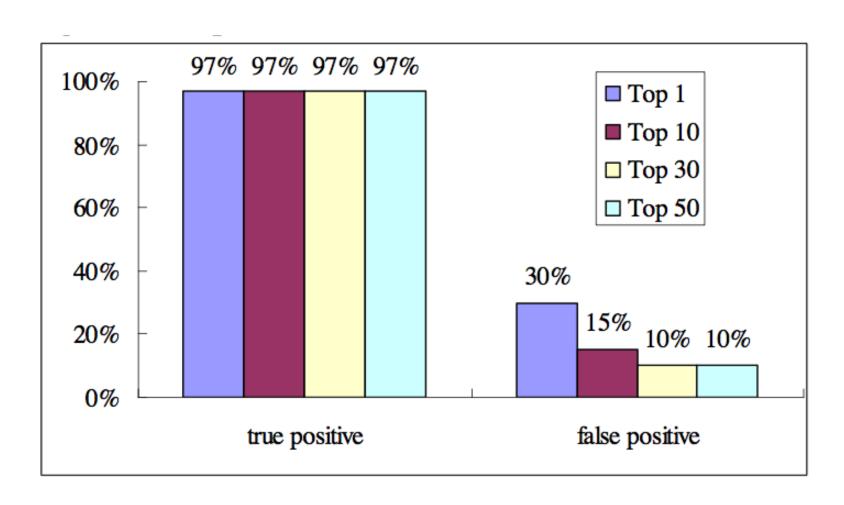
#### Other features

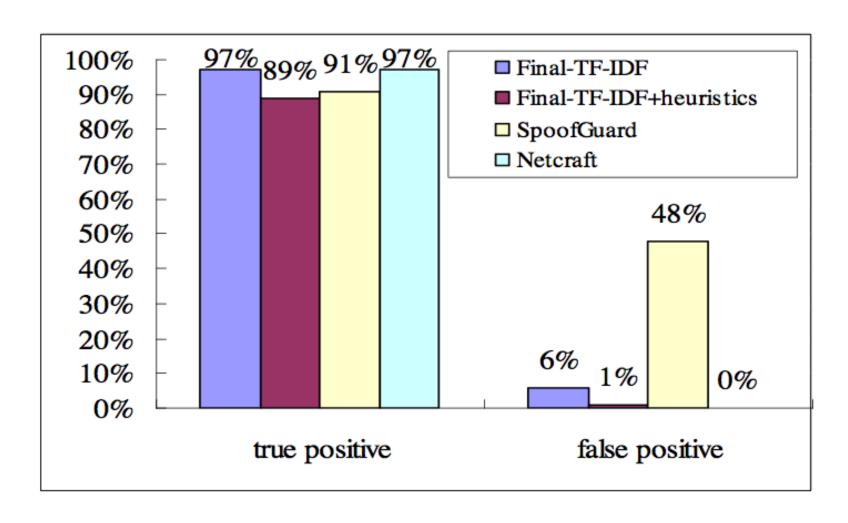
- Age of domain
- Known images
  - Presence of inconsistent well-known logos
  - Top-10 identified targets: eBay, PayPal,
     Citibank, Bank of America, Fifth Third Bank,
     Barclays Bank, ANZ Bank, Chase Bank, and
     Wells Fargo Bank
- Suspicious URL
  - Contains @ or in domain name

#### Other features

- Suspicious links
  - Same as suspicious URLs
- IP address as domain
- Dots in URL
  - Binary: #Dots > 5
- Forms
  - HTML <input> tag, with text such as "credit card", "password"







#### References

"Learning to Detect Phishing Emails",
 Fette et al, 2007

 "Cantina: A content-based approach to detecting phishing websites", Zhang et al, 2007