## Spam Filtering and Global Social Email Networks

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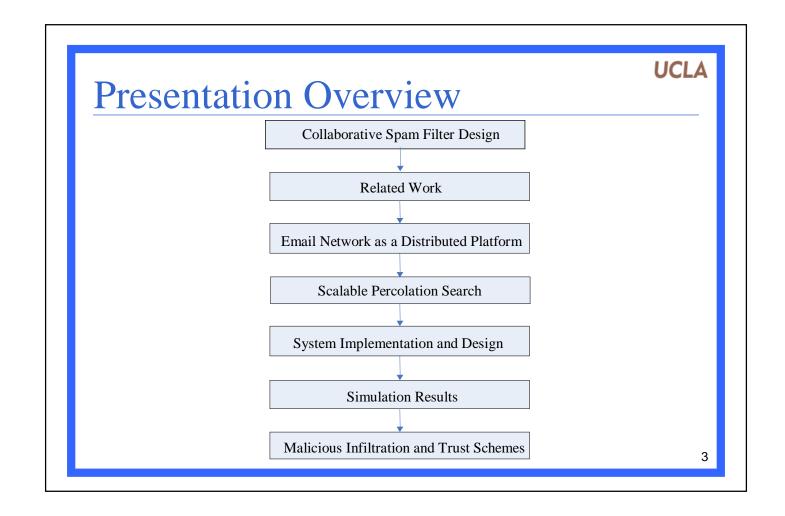
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#### Motivation

- In the previous presentation, we see that the *personal* email network can be harnessed to filter spam.
- This naturally begs the question: can the *global* social email network, which reaches hundreds of millions of users worldwide, be harnessed to help fight spam?



## Collaborative Spam Filter Design

- A collaborative spam filter uses the collective memory of, and feedback from, the users to reliably identify spams.
- There are three main challenges facing collaborative spam filter design:
  - 1. How to find and connect users?:
  - 2. How to make the search scalable?:
  - 3. Who to trust?

## **Existing Design and Solutions**

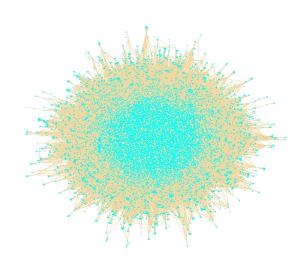
- SpamNet: central server design, server-based search, seniority trust scheme.
- Vipul's Razor (open source version of SpamNet).
- SpamWatch: distributed hash table (DHT) P2P network design, approximate text addressing (ATA) search scheme. [Zhou03]

#### Harnessing the Email Network

- The basic idea: instead of using dedicated central servers or P2P systems, we use our latent social email network as the distributed spam filter platform.
- All queries and communications are exchanged via background email through personal contacts.
- Motivation: we should exploit our existing network to find and connect users.

## **Topology of Email Networks**

- Email network: nodes = email addresses; edges = email exchanges in either direction.
- Email network data [Ebel02]: 56,969 nodes, 84,190 edges
- PL degree distribution:  $P(k) \propto k^{-1.81}$
- We can utilize the scalable percolation search algorithm on this naturally power-law network to search.



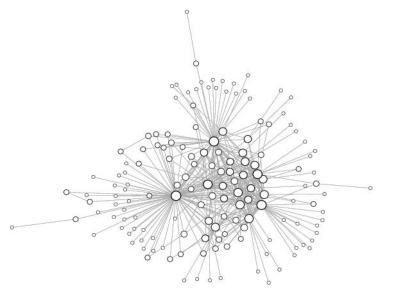
Real-world email network [Ebel02]

#### Percolation Search [Sarshar04]

- N. Sarshar, P. Boykin, and V. Roychowdhury. Percolation search in power law networks: Making unstructured peer-to-peer networks scalable. In *Proceedings of the 4<sup>th</sup> IEEE international conference on peer-to-peer computing*, 2004.
- Developed in the UCLA Complex Networks Lab led by Prof. Roychowdhury.
- Provably solves the search scalability problem in unstructured P2P networks.
- Even rare items can be found with probability 1, while keeping the traffic scalable  $(O(\log^2 N))$  for the optimal network topology).
- Offers comparable performance to Distributed Hash Table (DHT) object location schemes but in an unstructured network.

## **Bond Percolation Theory**

• Consider the random network below (153 nodes, 366 edges).



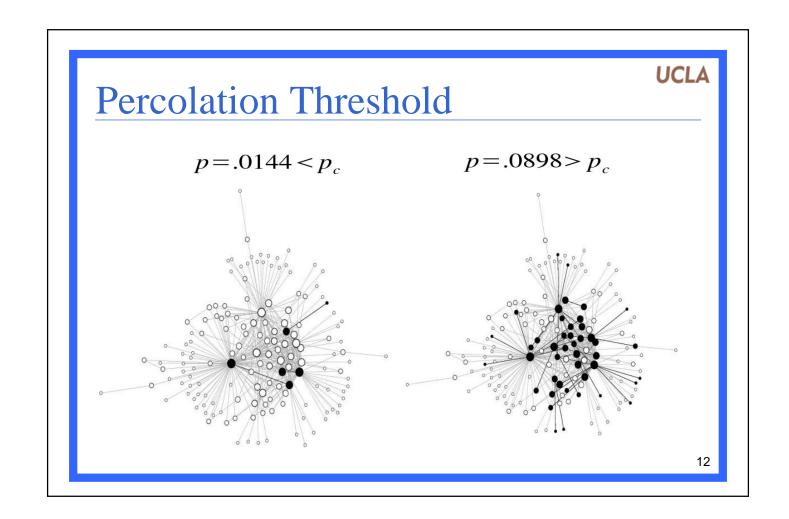
## **Bond Percolation Theory**

• Bond percolating with probability p means to keep each edge with probability p.



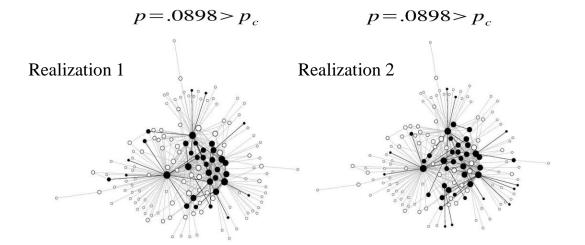
#### Percolation Threshold

- The percolation threshold  $p_c$  of a network is defined as followed:
- for a network  $p_c \approx \frac{\langle k \rangle}{\langle k^2 \rangle}$ ,  $p_c \approx .0359$  for our sample network.
- if  $p < p_c$ , the percolated network consists of small and disconnected components;
- if  $p > p_c$ , a giant connected component (GCC) emerges.



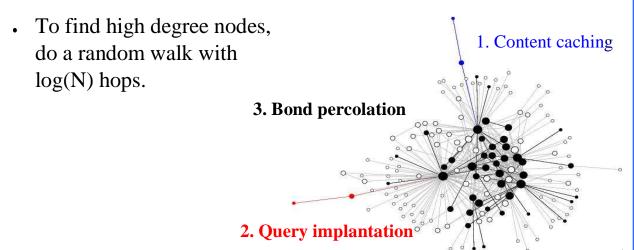
#### Percolation Search

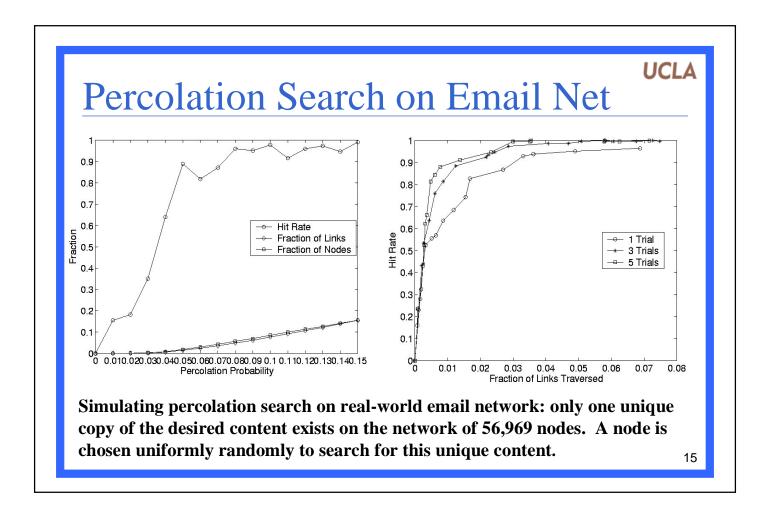
• For  $p > p_c$ , the high degree nodes almost suredly remain in the giant connected component of the percolated network.



#### Percolation Search

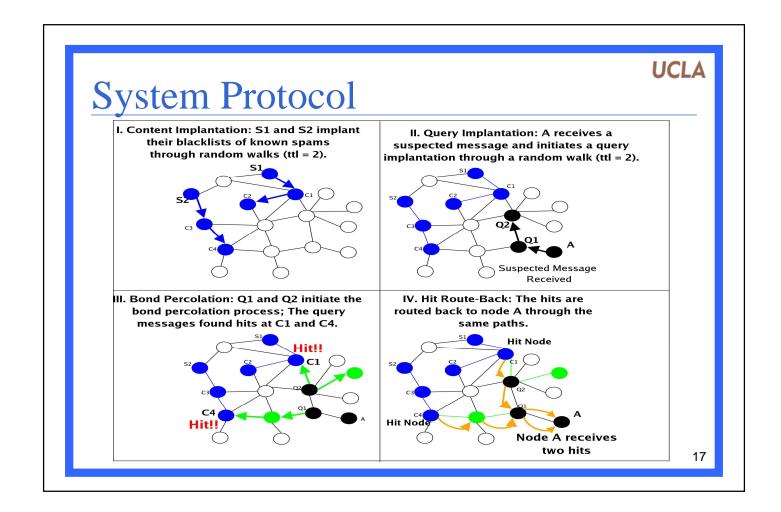
• The idea: start queries from high degree nodes AND cache contents at high degree nodes; high degree nodes can find each other by percolation.

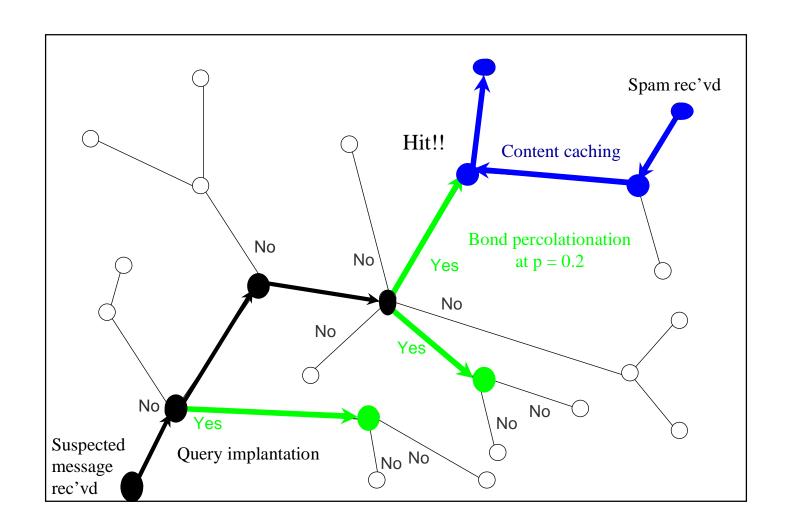




## Build a Distributed Spam Filter!

- Use the digest-based mechanisms to index spams.
- Use our latent social email network as the distributed platform.
- Use the efficient percolation search protocol to query for suspected spam messages indexed by digests.





#### Distributed Client Implementation

- Implement the system as a plug-in to any email client such as Sendmail.
- Any other existing spam filter can be used to do initial filtering.
- All system messages (e.g. queries, hit returns and content caching) can be done via background email exchanges.

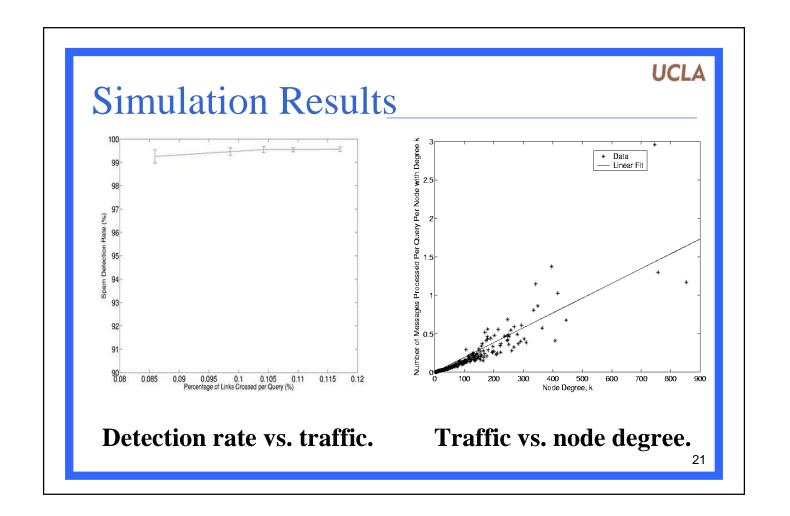
```
Algorithm 1 PROCESS-MAIL(Email E)
 1: if DefinitelySpam(E) then
      Mark E as Spam
 3: else if DefinitelyNotSpam(E) then
      Mark E as not Spam
 5: else
      D_e = Digest(E); \{Gray SPAM\};
      Implant percolation of D_e on a random walk of
      length l
      Wait(T);
      H_e = HitScore();
10:
      if H_e < threshold then
11:
        Mark E as not Spam
12:
        Mark E as spam
13:
14:
      end if
15: end if
Algorithm 2 Publish-Spam(Email E)
```

1:  $D_e = Digest(E)$ ;

2: Implant  $D_e$  on a random walk of length l

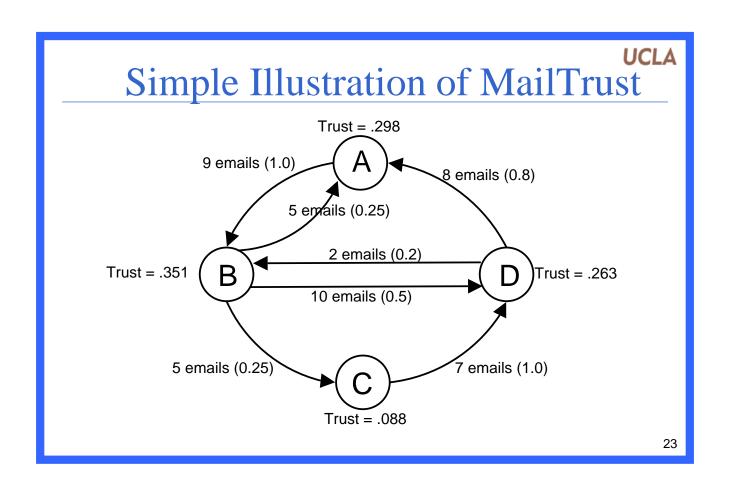
## Simulation on Real Email Net

Number of Nodes	56,969
Number of Edges	84,190
Node Degree Distribution	Power-Law ( $\exp = 1.8$ )
Bond Percolation Threshold	$\frac{\langle k \rangle}{\langle k^2 \rangle} = .0169$
Time-to-Live (TTL)	50
# of arrivals of the same spam	500
# of hits needed to id spam	2
# of runs	30



#### Infiltration and MailTrust

- Malicious users will attempt to subvert the system.
- We propose an eigen-based trust scheme.
- The topological structure of the social email networks can be used to assign trust or reputation to individual users (just like PageRank).
- Model each email contact as placing a unit of trust on the recipient.
- Model the whole graph as a discrete time Markov chain.



#### Markov Chain (Cont'd)

• Using the example on the previous slide, we can construct the following transition matrix:

$$P = \begin{pmatrix} 0 & 1.0 & 0 & 0 \\ 0.25 & 0 & 0.25 & 0.5 \\ 0 & 0 & 0 & 1.0 \\ 0.8 & 0.2 & 0 & 0 \end{pmatrix}$$

#### Markov Chain (Cont'd)

- In addition, we must ensure that the Markov Chain is ergodic.
- In email network, this means that nodes must out-links.
- This can be achieved by having nodes with zero out-degree assign uniform trust to a set of pretrusted nodes who have been carefully picked.

#### Markov Chain (Cont'd)

- In order to find the MailTrust score, compute the steady state probability distribution, which corresponds to the principal eigenvector of *P*.
- Can be done efficiently using the *Power Iteration* method.
- Using this method, the MailTrust score vector for our example are found to be:

· (.298 .351 .088 .263)

• The MailTrust scores are computed in a distributed fashion (see a paper by Kamvar et al. ACM WWW 03).

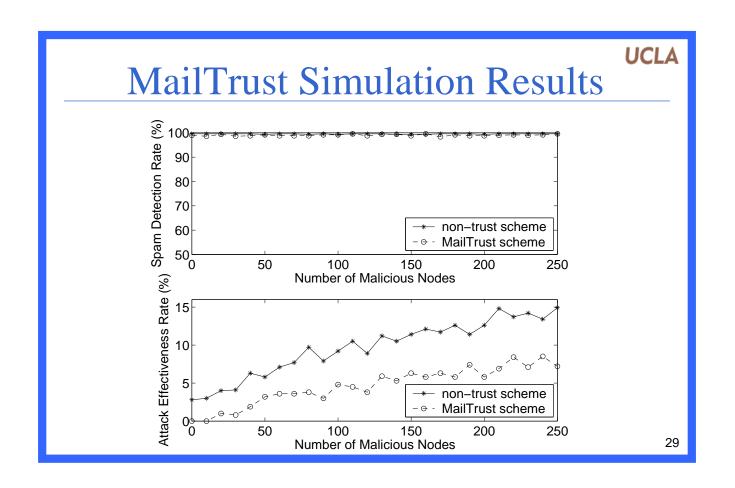
## MailTrust Simulation Setup

- A malicious node can subvert the system by introducing blacklists of well-known valid messages into the network.
- Note that this form of attack will only raise the false positive rate of the system and it has no impact on the spam detection rate.
- a small fraction of nodes in the network (250 nodes) will be labelled as malicious nodes and these malicious nodes will blacklist nonspams from popular mailing lists;

# MailTrust Simulation Setup (Cont'd)

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- The malicious nodes will follow all specifications of the protocol such as forwarding and routing queries and storing cache implants for other nodes.
- The MailTrust scheme results in about 50% improvement in lowering the false positive rate.



#### Summary

- We presented a distributed collaborative anti-spam system.
- Can be implemented as a simple plug-in.
- Delivers superior performance while keeping bandwidth cost on the internet extremely low.
- MailTrust scheme is effective in deterring malicious users.
- A more comprehensive version is available on: <a href="http://arxiv.org/abs/physics/0504026">http://arxiv.org/abs/physics/0504026</a>
- Journal version will soon appear on IEEE Computer.