

Exploiting Transport-Level Characteristics of Spam

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Outline

- 1 Background
- 2 Experimental Methodology
- 3 Learning and Prediction
- 4 Open Questions



The Spam Arms Race

Attackers, scammers and thieves quickly adapt to defenses. Most effective solutions exploit *fundamental* weaknesses of attackers

Current Best Practices:

- Content Filtering ... response: modify word tokens
- Reputation Analysis ... response: dynamic, fresh addresses
- Collaborative Filtering ... response: mail uniqueness
- And the cycle continues: Authentication Schemes, computational puzzles, etc.



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The Spam Arms Race

We propose a different approach:

- No panacea; existing solutions all have weaknesses
- Our solution, “SpamFlow,” is distinct from current practice

Question:

Are traffic characteristics a fundamental weakness of spam?



Hypothetical Question

Specifically:

- What is the *transport* (TCP/IP packet stream) character of spam?
- Are there *differences* between spam and ham flows?
- How to exploit differences in a way which spammers cannot easily evade?

Why ask this question?



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Specifically:

- What is the *transport* (TCP/IP packet stream) character of spam?
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Transport-Level Characteristics of Spam

Two Observations

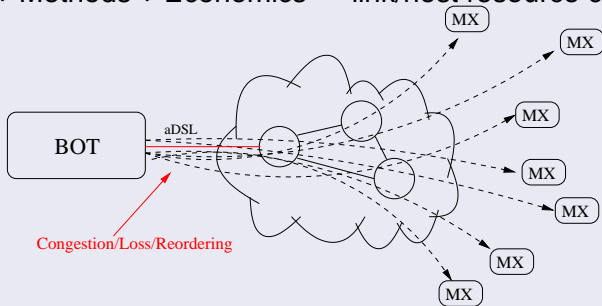
- 1 Low Penetration:
 - due to existing filters, user ambivalence
 - → huge volumes of spam
- 2 Sending Methods:
 - Open mail relays, email trojans, botnets, dialup
 - → Low asymmetric bandwidth, widely distributed



Transport-Level Characteristics of Spam

Combining Observations: Low Penetration + Sending Methods

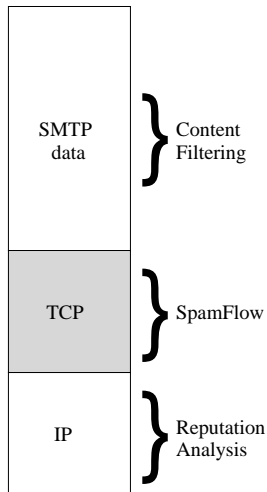
Volume + Methods + Economics → link/host resource contention



Contention:

Contention manifests as TCP/IP loss, retransmission, reordering, etc.

Understanding SpamFlow



- Not looking at IP header
- Not looking at data
- SpamFlow: TCP stream, incl timing
- (look at combining methods later)



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A Brief Diversion on TCP/IP

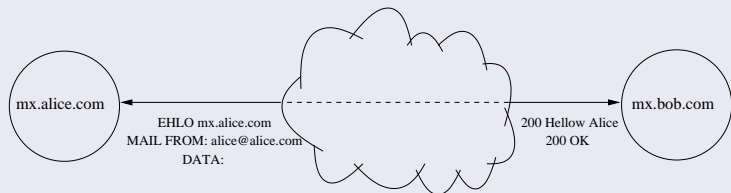
Transmission Control Protocol (TCP):

- Reliable, bi-directional, in-order byte transmission abstraction
 - Acknowledgments
 - State Machine
- Flow and congestion control
 - Reacts to loss, persistent congestion
- Multi-flow fairness and efficient resource utilization (AIMD)
 - Round trip time (RTT) estimation
 - Bandwidth probing



SMTP and TCP

Transmission Control Protocol:



- Simple Mail Transport Protocol (SMTP) uses TCP for transport
- Sequence of SMTP handshaking between Mail Transport Agents (MTAs)
- Mail contents are packetized

How do Spam Connections Behave?

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How do Spam Connections Behave?

...or, a quick look at `netstat`

```

RcvQ  SndQ  Local          Foreign Addr          State
0      0      srv:25        92.47.129.89:49014   SYN_RECV
0      0      srv:25        ppp83-237-106-114.:29081 SYN_RECV
0      0      srv:25        88.200.227.123:25068 SYN_RECV
0      0      srv:25        92.47.129.89:49014   SYN_RECV
0      0      srv:25        ppp83-237-106-114.:29084 SYN_RECV
0      0      srv:25        88.200.227.123:25068 SYN_RECV
0      0      srv:25        88.200.227.123:25069 SYN_RECV
0      0      srv:25        88.200.227.123:25070 SYN_RECV
0      0      srv:25        88.200.227.123:25074 SYN_RECV
0      0      srv:25        84.255.150.15:4232   SYN_RECV
0      25     srv:25        222.123.147.41:50282 LAST_ACK
0      28     srv:25        adsl-pool-222.123.:1720 LAST_ACK
0      31     srv:25        222.123.147.41:50152 LAST_ACK
0      15     srv:25        222.123.147.41:50889 LAST_ACK
0      9      srv:25        88.245.3.19:venus    LAST_ACK
0      25     srv:25        78.184.155.70:1854   FIN_WAIT1
0      23     srv:25        190-48-30-225.spe:50920 FIN_WAIT1
0      23     srv:25       .dsl.dynamic812132:48154 FIN_WAIT1
0      23     srv:25        ip-85-160-91-16.e:48093 FIN_WAIT1
0      23     srv:25        88.234.141.158:48389 FIN_WAIT1
0      23     srv:25        p5B0FBB5D.dip.t-d:11965 FIN_WAIT1
...

```



How do Spam Connections Behave?

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```

RcvQ  SndQ  Local                Foreign Addr          State
0      0      srv:25              92.47.129.89:49014    SYN_RECV
0      0      srv:25              ppp83-237-106-114 :29081    SYN_RECV
0      0      srv:25              88.200.2...
0      0      srv:25              92.47.12...
0      0      srv:25              ppp83-23...
0      0      srv:25              88.200.2...
0      0      srv:25              88.200.2...
0      0      srv:25              88.200.2...
0      0      srv:25              84.255.1...
0      25     srv:25              222.123...
0      28     srv:25              adsl-poo...
0      31     srv:25              222.123...
0      15     srv:25              222.123...
0      9      srv:25              88.245.3...
0      25     srv:25              78.184.1...
0      23     srv:25              190-48-3...
0      23     srv:25              dsl.dyna...
0      23     srv:25              ip-85-16...
0      23     srv:25              88.234.14...
0      23     srv:25              p5B0FBB5D.dip.t-d:11965  FIN_WAIT1
...

```

TCP Stuck in States

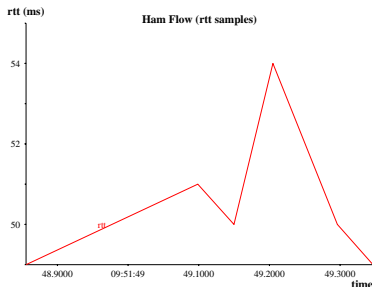
- Stays in these states for minutes
- Half-open connections
- Remote MTAs that “disappear” mid-connection
- Remote MTAs that send FIN and disappear



What about RTT?

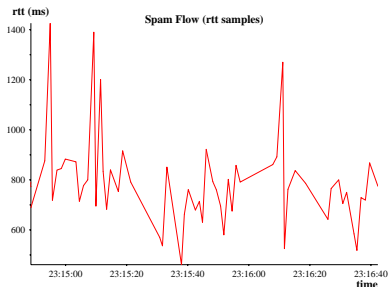
...building more intuition

```
Received: from vms044pub.verizon.net
From: "Dr. Beverly, MD" <b@ex.com>
Subject: thoughts
Dear Robert,
I hope you have had a great week!
```



Ham

```
Received: from unknown (59.9.86.75)
From: Erich Shoemaker <ried@ex.com>
Subject: Replica for you
A T4g Heuer w4tch is a luxury statement
on its own.
In Prestlge Replicas, any T4g Heuer...
```

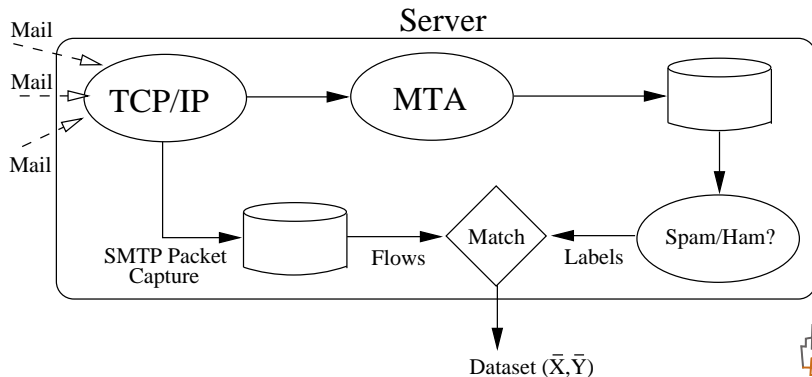


Spam



Data Collection

- Instrument a Mail Transport Agent (MTA) server
- Collect SMTP packet trace
- Match labeled emails to packet flows

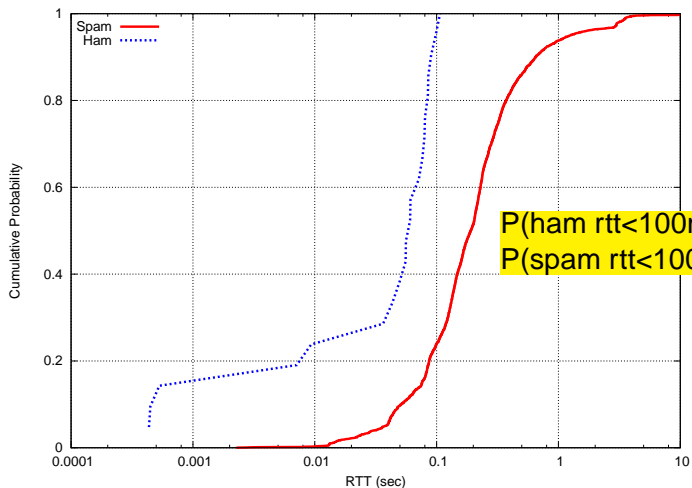


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Round Trip Time



Round Trip Time

cont'd

Bayes' Rule

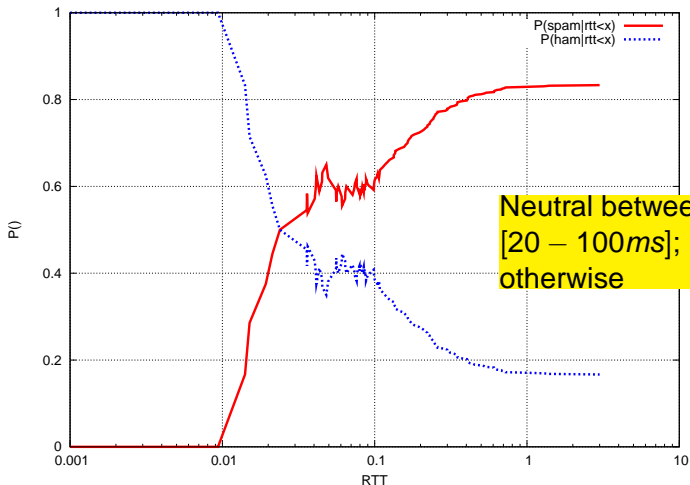
- Use causal information to form diagnosis

$$P(spam|rtt > x) = \frac{P(rtt > x|spam)P(spam)}{P(rtt > x)} \quad (1)$$



Round Trip Time

cont'd



Selecting Features

Wait! You're disenfranchising distant servers!

- Yes; may be a good thing
- $\simeq 5\% > 1s$
- More importantly...

Other Transport "Features:"

- Packets, Retransmits, OutOfOrder, RSTs, FINs
- Zero Window, Minimum Cong. Window, Max Idle, Jitter, etc.
- *Adaptable* per-user, per-network

Key Insight

Statistical flow properties can provide differentiation

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Key Insight

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Non-Features

Non-Features

- Many intuitively “good” features turn out not to be
- Strength of statistical approach

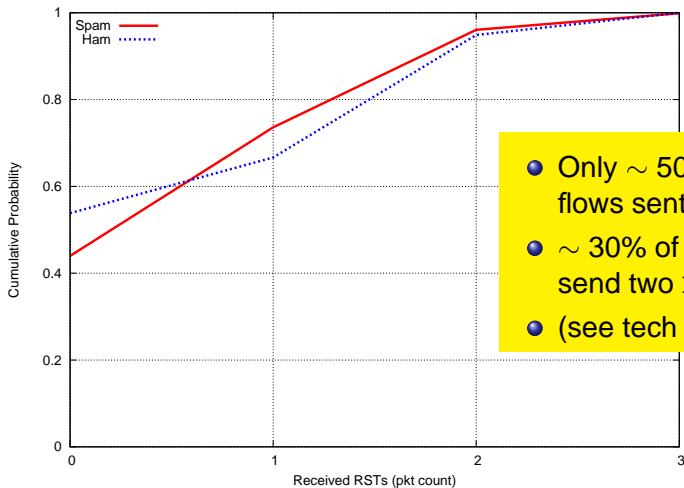
One Example in Detail:

- RSTs as abortive close on socket
- A good indication of misbehaving flows?



Non-Features

Example: Received RSTs



- Only ~ 50% of ham flows sent no RSTs!
- ~ 30% of ham flows send two RSTs!
- (see tech report for why)

Outline

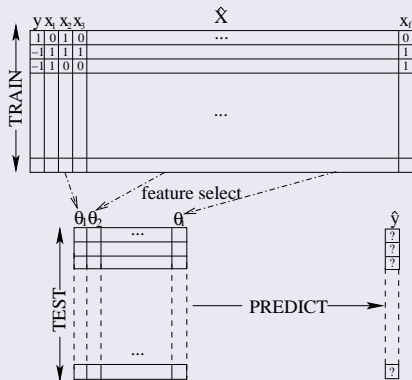
- 1 Background
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Picking Features

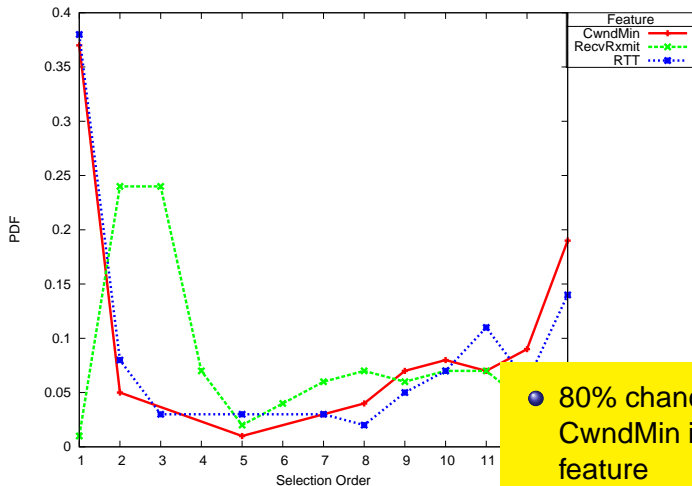
So, which features provide discrimination?

- Feature selection
- Simple method is forward fitting
- Greedily choose one available feature to minimize training error



Picking Features

cont'd

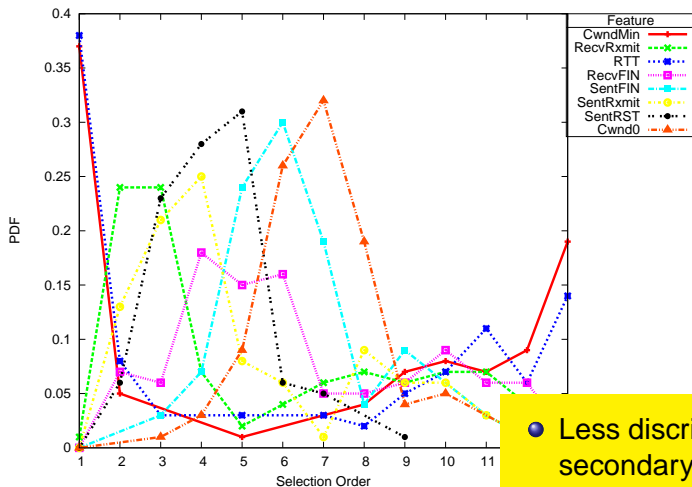


• 80% chance that RTT or CwndMin is best single feature



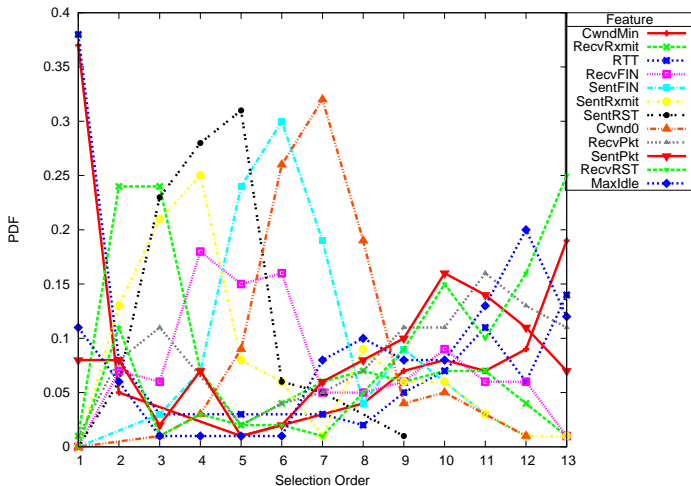
Features

cont'd



Features

cont'd



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SpamFlow

Based on observations, build a model

- Supervised learning, binary classification
- E.g. Bayes Nets, Support Vector Machines, etc.

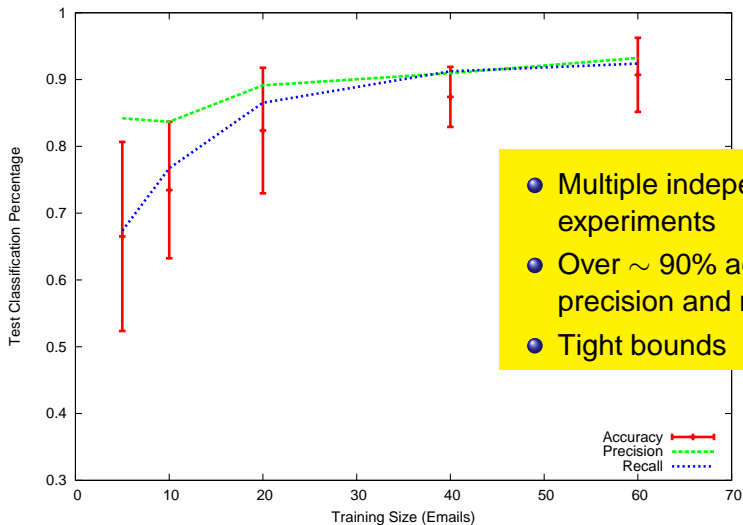
SpamFlow

- A working implementation of the ideas using SVMs

Evaluation

- FP = ham marked as spam
- FN = spam marked as ham
- $accuracy = \frac{TP+TN}{P+N}$
- $precision = \frac{TP}{TP+FP}$

Prediction Performance



- Multiple independent experiments
- Over $\sim 90\%$ accuracy, precision and recall
- Tight bounds



SpamAssassin False Negatives

False Negatives

- Against our data set, SpamAssassin gives 127 false negatives
- SpamFlow detects 78% of those
- → useful to combine methods!

For example...



SpamAssassin False Negatives

Received: (qmail 12851 invoked from network); 24 Jan 2008 05:14:58 -0000

Received: from 201-213-46-215.net.prima.net.ar (201.213.46.215:8963)

by ralph.rbeverly.net with SMTP; 24 Jan 2008 05:14:58 -0000

Received: from unknown (HELO deviant) (192.168.0.5) by mail6.colossal.com
with SMTP; Thu, 24 Jan 2008 00:14:58 -0500

Date: Thu, 24 Jan 2008 00:14:58 -0500

To: rbeverly@grdata.com, rcmsjm@grdata.com, reb3@grdata.com, roots.nojunk@grdata.com, russell_s

From: "Jordan Abrams" <inclusionVito@familyhistree.com>

Subject: Canadian Pharmacy Online! - 70-80% OFF!

Content-Length: 76

Lines: 6

Re" Your Pharmacy order # 85493899

Pls Go ' [www.protectfair](http://www.protectfair.com) ' dot com



SpamAssassin False Negatives

```

Received: (qmail 12851 invoked from network)
Received: from 201-213-46-215.net.prima.net.
  by ralph.rbeverly.net with SMTP; 24 Jan 2008 00:14:58 -0500
Received: from unknown (HELO deviant) (192.168.1.1)
  with SMTP; Thu, 24 Jan 2008 00:14:58 -0500
Date: Thu, 24 Jan 2008 00:14:58 -0500
To: rbeverly@grdata.com, rcmsjm@grdata.com, rbeverly@grdata.com
From: "Jordan Abrams" <inclusionVito@family.com>
Subject: Canadian Pharmacy Online! - 70-80% OFF!
Content-Length: 76
Lines: 6

```

Re" Your Pharmacy order # 85493899

Pls Go ' www.protectfair ' dot com

SpamAssassin:

```

X-Spam-Status: No,
score=3.5 required=5.0
tests=BAYES_50,
FS_OBFU_PRMCY,
SORTED_RECIPS,
UNPARSEABLE_RELAY
autolearn=no version=3.2.3

```



SpamAssassin False Negatives

Received: (qmail 12851 invoked from network); 24 Jan 2008 05:14:58 -0000

Received: from 201-213-46-215.net.prima.net.8

by ralph.rbeverly.net with SMTP; 24 Jan 2008

Received: from unknown (HELO deviant) (192.168.1.1)

with SMTP; Thu, 24 Jan 2008 00:14:58 -0500

Date: Thu, 24 Jan 2008 00:14:58 -0500

To: rbeverly@grdata.com, rcmsjm@grdata.com, r

From: "Jordan Abrams" <inclusionVito@family1

Subject: Canadian Pharmacy Online! - 70-80% OFF

Content-Length: 76

Lines: 6

Re" Your Pharmacy order # 85493899

Pls Go ' www.protectfair ' dot com

SpamFlow:

SntPkt: 45 RcvPkt: 29
 SntRxmit: 0 RcvRxmit: 1
 SntRST: 0 RcvRST: 0
 SntFIN: 1 RcvFIN: 1
 Cwnd0: 0 MinCwnd: 65280
 MaxIdle: 1.366636
 RTT: 0.162413



Open Questions

Spam is an Arms Race:

- How would spammers react?
- Adapt by slowing down, sending less mail
- Could spammers tweak TCP stacks and circumvent?

Future Work:

- Gather additional data sets
- Package, distribute
- Explore method's potential in other domains



Summary

- Attacking spam at a different layer
- Correct predictions with over 90% accuracy, precision and recall *without* content or reputation analysis
- SpamFlow finds 78% of SpamAssassin false-negatives
- No implementation hurdle, easily combined with existing techniques

Thanks!

Questions?





SpamFlow FAQ

- 1 *Can SpamFlow be more conservative in using RTT:* Yes, even a highly conservative filter can still leverage RTT to eliminate extremely large RTT spam flows.
- 2 *Doesn't SpamFlow privilege well-connected senders?* Personal, home or small business servers do not have the same volume requirement as spammers and thus are unlikely to induce the same TCP congestion effects we observe. SpamFlow only discriminates against sources that are *both* poorly connected *and* injecting large volumes of mail.
- 3 *What about email lists?* In contrast to spam, which must be sent continually, email list traffic can be scheduled in order to not cause local congestion.

Support Vector Machines

Dual-Form, Constrained Optimization:

$$\sum_{t=1}^n \alpha_t - \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \alpha_i \alpha_j \mathbf{K}(\phi(\mathbf{x}_i), \phi(\mathbf{x}_j)) \text{ s.t. } \mathbf{C} \geq \alpha_t \geq 0, \sum_{t=1}^n \alpha_t \mathbf{y}_t = 0 \quad (2)$$

- Separate training set into two classes in most general way
- **Main insight:** find hyper-plane separator that maximizes the minimum margin between convex hulls of classes
- **Second insight:** if data is not linearly separable, take to higher dimension
- **Result:** generalizes well, fast, accommodate unknown data structure

What's going on here?

Example: Received RSTs

Google sends SMTP QUIT, then active close, then RSTs passive close

```
11:55:57.807504 googl > srv: P 187089:187095(6) ack 143 win 5720
11:55:57.807510 googl > srv: F 187095:187095(0) ack 143 win 5720
11:55:57.807628 srv > googl: . ack 187096 win 32614
11:55:57.807863 srv > googl: P 143:167(24) ack 187096 win 32614
11:55:57.808181 srv > googl: F 167:167(0) ack 187096 win 32614
11:55:57.834759 googl > srv: R 46149836:46149836(0) win 0
```

Yahoo! sends SMTP QUIT, srv performs active close. Yahoo! then sends three RSTs when srv goes to TIME_WAIT

```
11:20:35.023406 srv > yahoo: P 113:137(24) ack 1426 win 32120
11:20:35.023782 srv > yahoo: F 137:137(0) ack 1426 win 32120
11:20:35.023983 yahoo > srv: F 1426:1426(0) ack 113 win 33304
11:20:35.024073 srv > yahoo: . ack 1427 win 32120
11:20:35.076591 yahoo > srv: R 776208340:776208340(0) win 0
11:20:35.076969 yahoo > srv: R 776208340:776208340(0) win 0
11:20:35.077381 yahoo > srv: R 776208341:776208341(0) win 0
```

Abortive close in Postfix source; normal behavior

What's going on here?

Example: Received RSTs

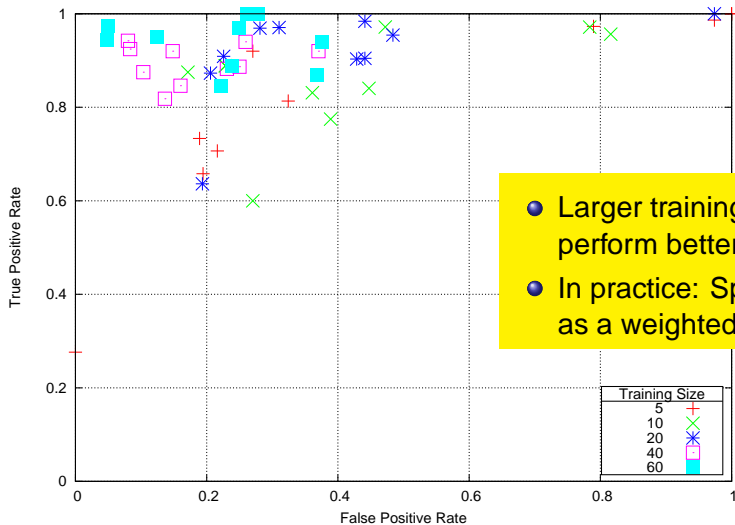
Is abortive close a common “normal” SMTP technique?

Postfix Source

```
static void start_connect(SESSION *session) {
    int fd;
    struct linger linger;
    linger.l_onoff = 1;
    linger.l_linger = 0;
    if (setsockopt(fd, SOL_SOCKET, SO_LINGER, (char *) &linger,
        sizeof(linger)) < 0)
        ...
}
```



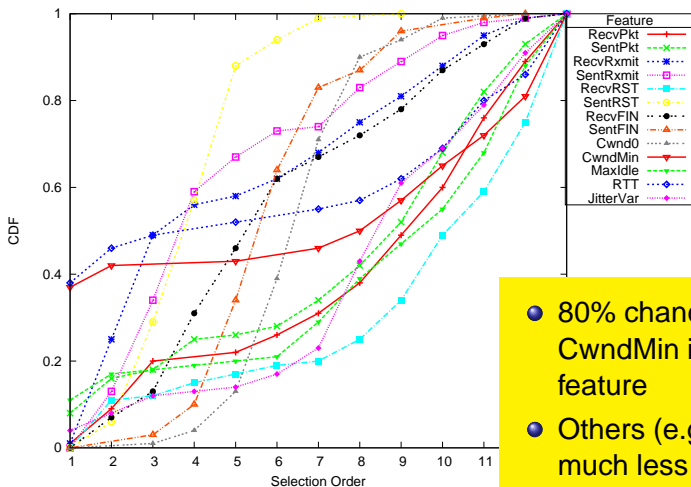
ROC Curve



- Larger training sizes perform better
- In practice: SpamFlow as a weighted voter

Features

cont'd



- 80% chance that RTT or CwndMin is best first feature
- Others (e.g. RecvRST) much less discriminatory

Data Collection

Dataset:

- One week, January 2008
- $\sim 18k$ emails, only ~ 200 legitimate ham
- Normalize spam and ham count for each experiment, randomly select spams
- Dataset is small; future work corrects this

This talk: method, intuition, validation

