

Combating Snowshoe Spam with Fire

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Overview

Introduction

Methodology

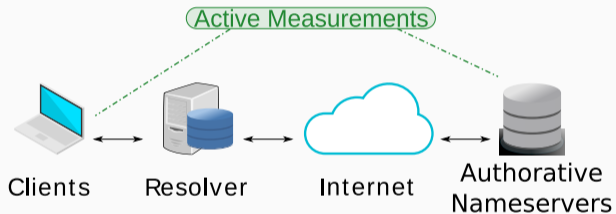
Results

Conclusions

Introduction

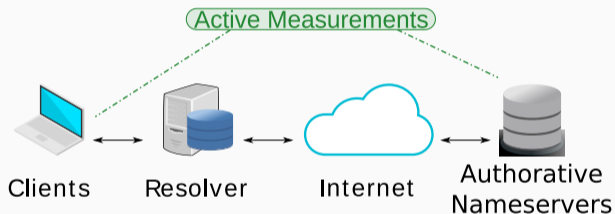
- Active DNS Measurements

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Background Info

- Active DNS Measurements
- Snowshoe Spam



- Snowshoe spam is hard to detect

Starting Point

- Snowshoe spam is hard to detect
- Sender Policy Framework (SPF)

Starting Point

- Snowshoe spam is hard to detect
- Sender Policy Framework (SPF)
- DNS domain

Research Question

How can we detect snowshoe spam through active DNS measurements?

Methodology



(source)



(processing)



(storage)



(validation)

- Active DNS Measurement Platform

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- Queries more than 60% of registered domain names

Datasets & Features

- Two types of datasets
 - Labeled
 - Unlabeled

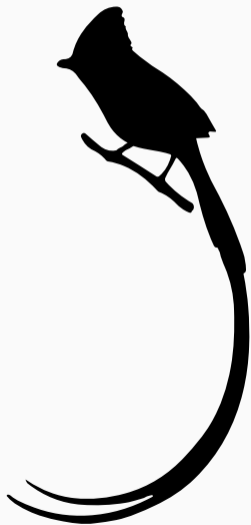
Datasets & Features

- Two types of datasets
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- 37 Features

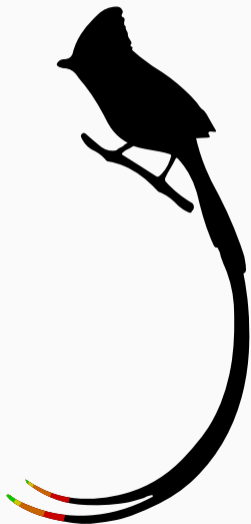
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- Long Tail Analysis

Long Tail Analysis



The long tail of the DNS



The **long** tail of the DNS

- Trained and evaluated many classifier algorithms

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- Ranked performance based on 'precision' metric

$$\text{Precision} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$$

- Trained and evaluated many classifier algorithms
- Ranked performance based on 'precision' metric
- Selected AdaBoost Classifier as classifier of choice (110 false positives out of 10851 ham domains)

- DNS based way of hosting a blacklist

Realtime Blackhole List

- DNS based way of hosting a blacklist
- Daily detections

Realtime Blackhole List

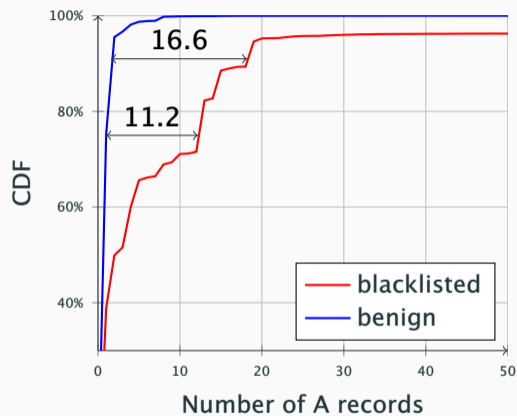
- DNS based way of hosting a blacklist
- Daily detections
- Compared to other blacklists

- SURFmailfilter

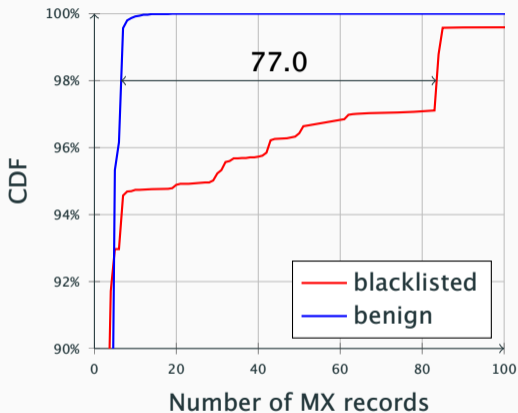
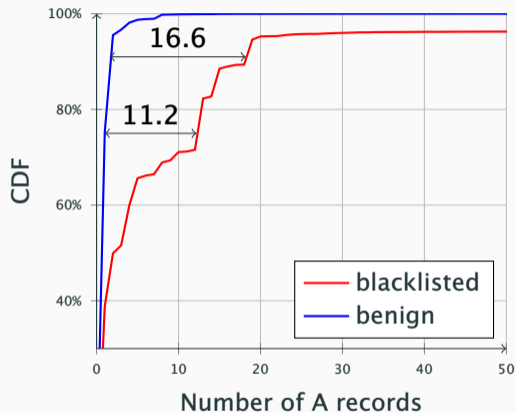
- SURFmailfilter
- Initially in evaluation mode

Results

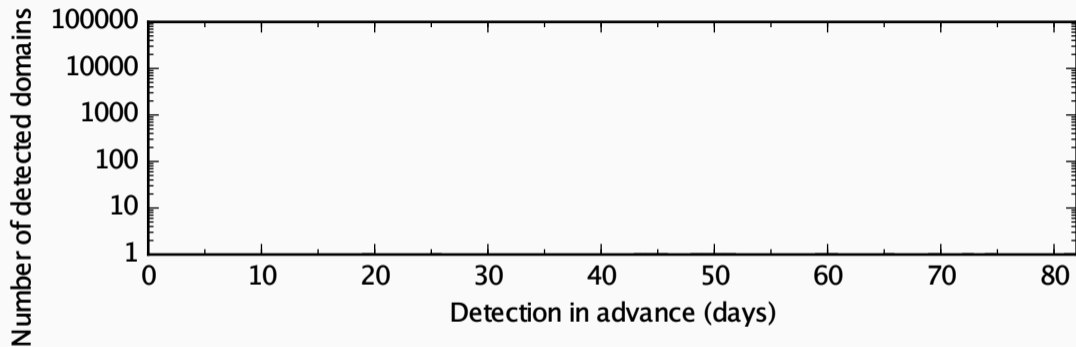
Comparison training data



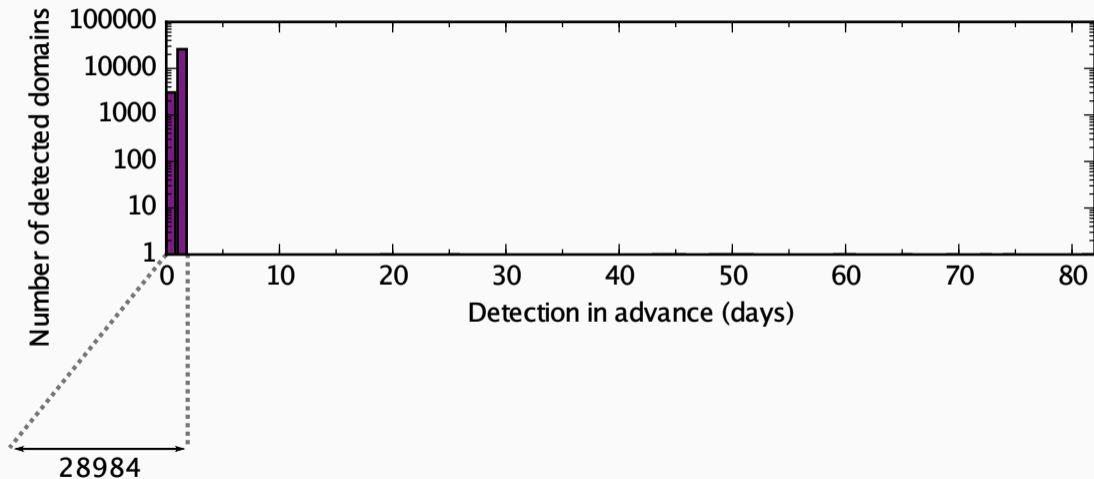
Comparison training data



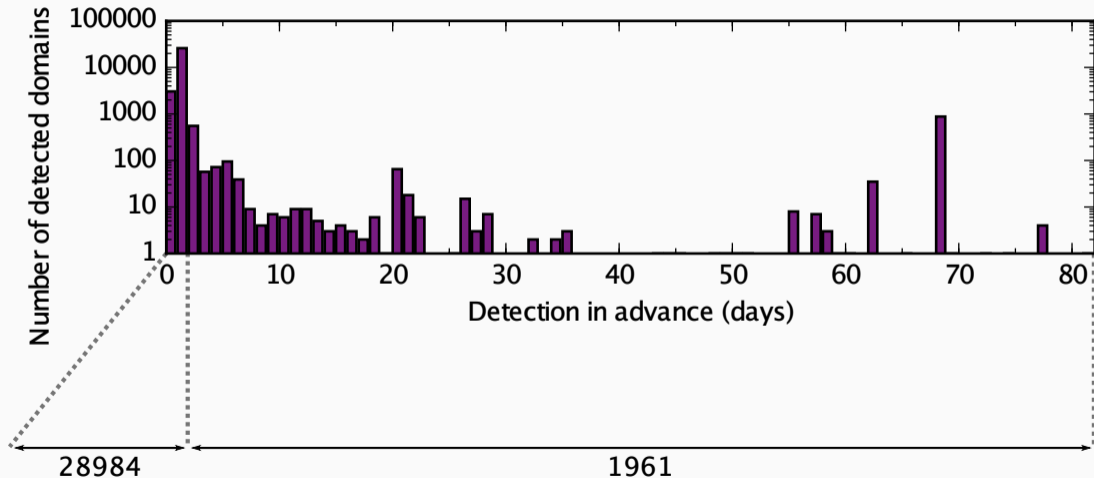
Early Detection



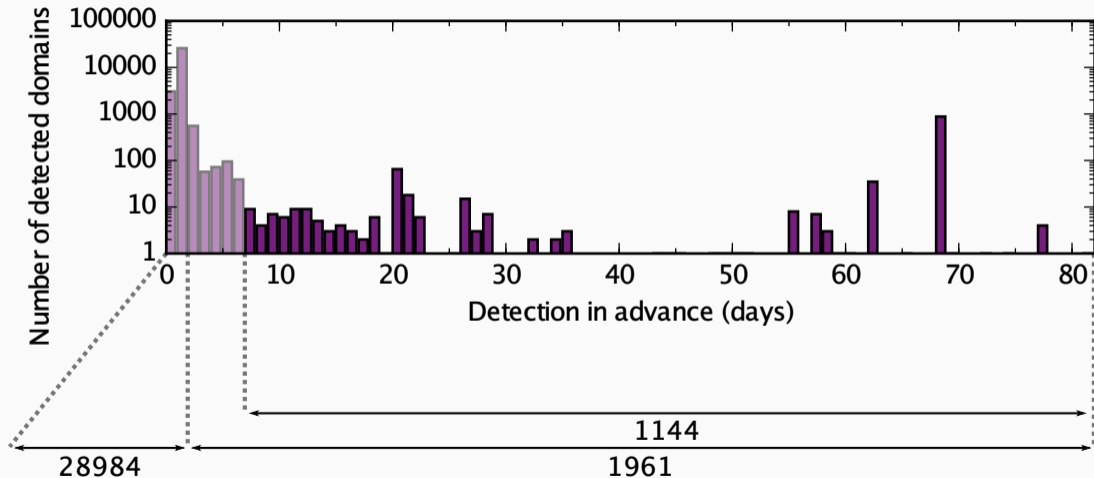
Early Detection



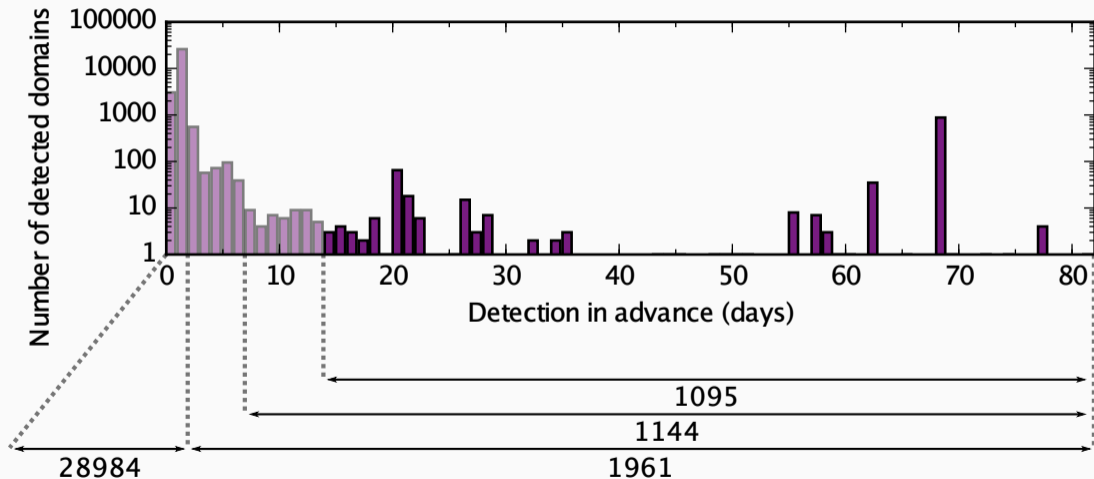
Early Detection



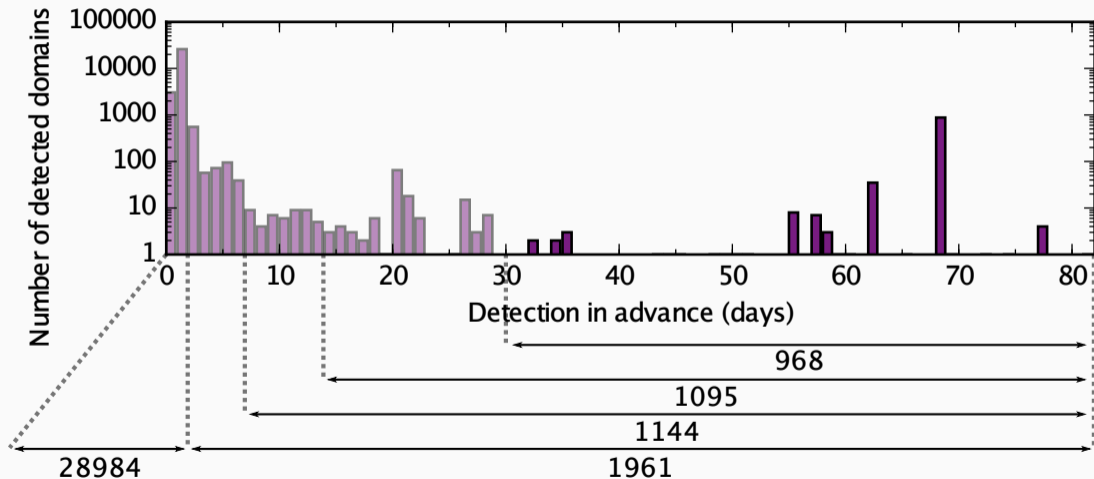
Early Detection



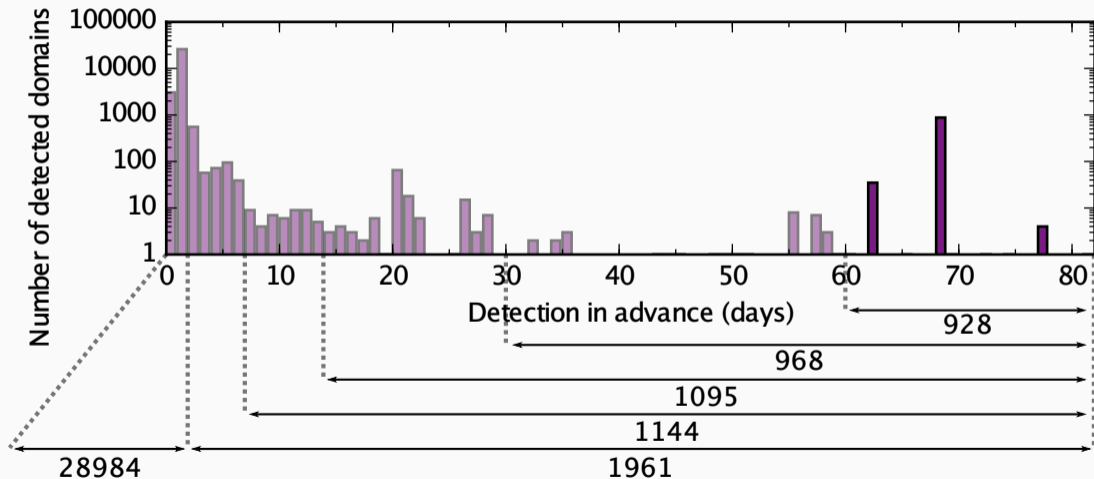
Early Detection



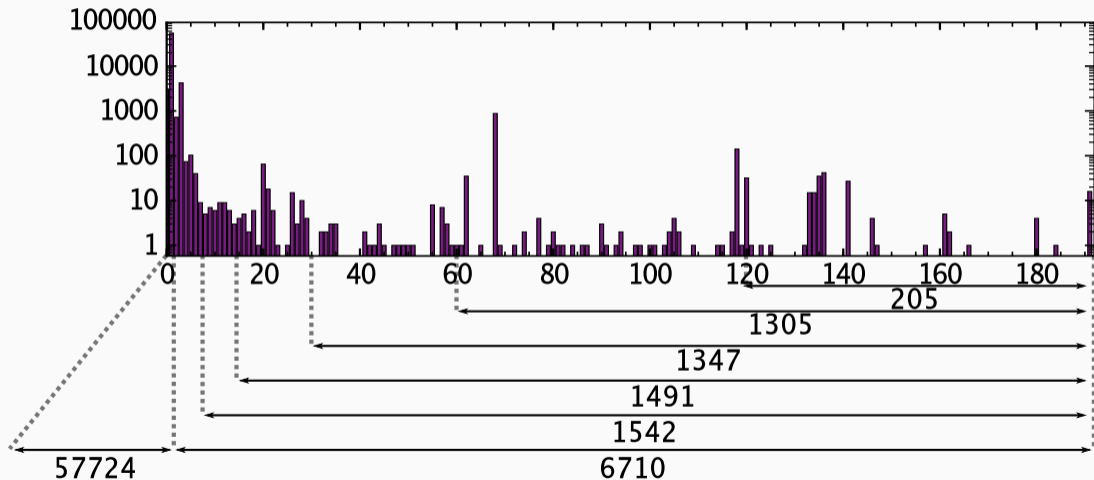
Early Detection



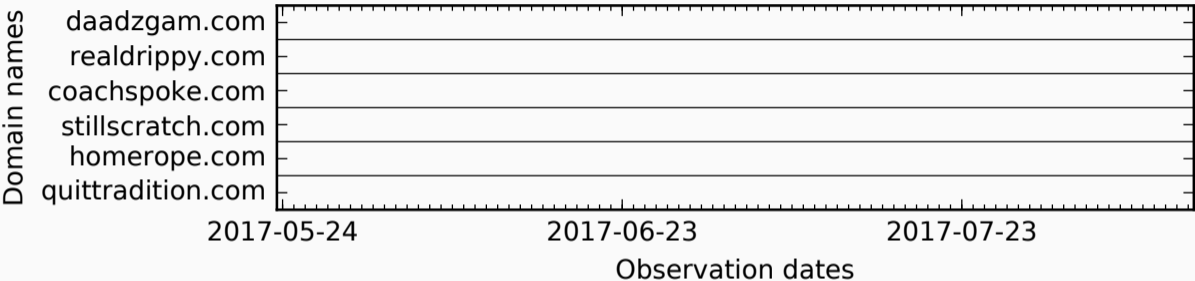
Early Detection



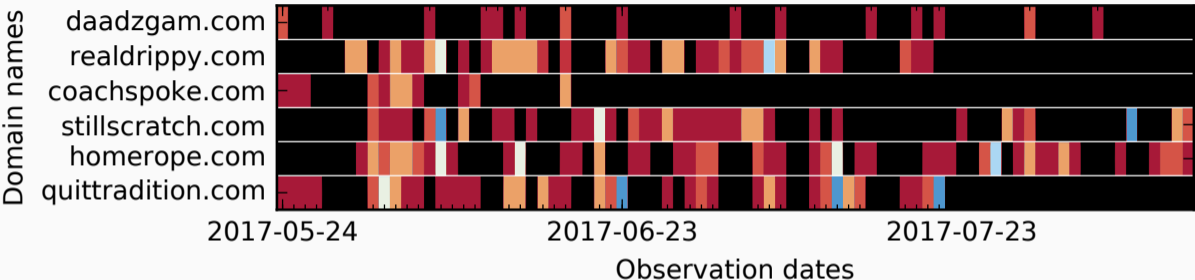
Early Detection (update)



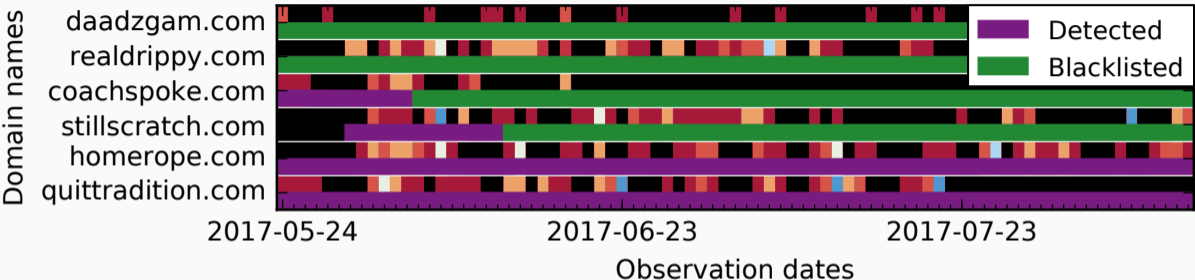
SURF Results



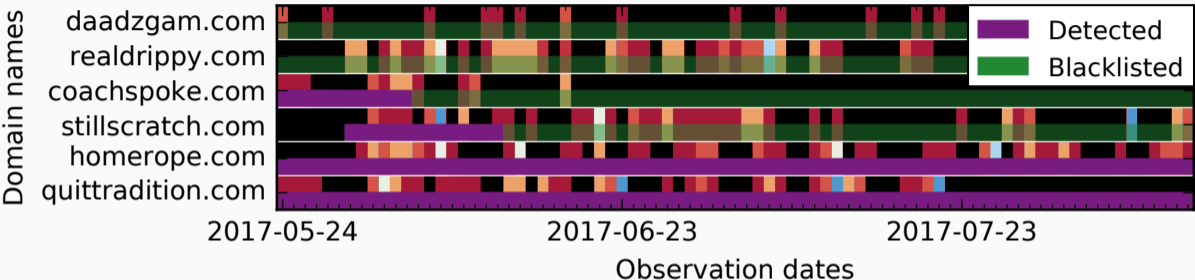
SURF Results



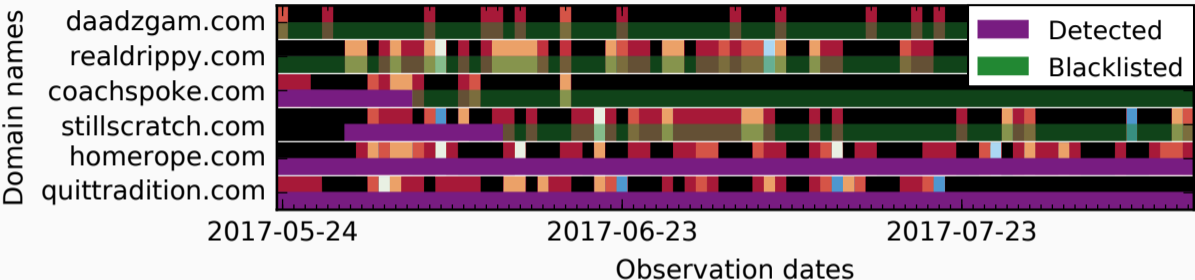
SURF Results



SURF Results

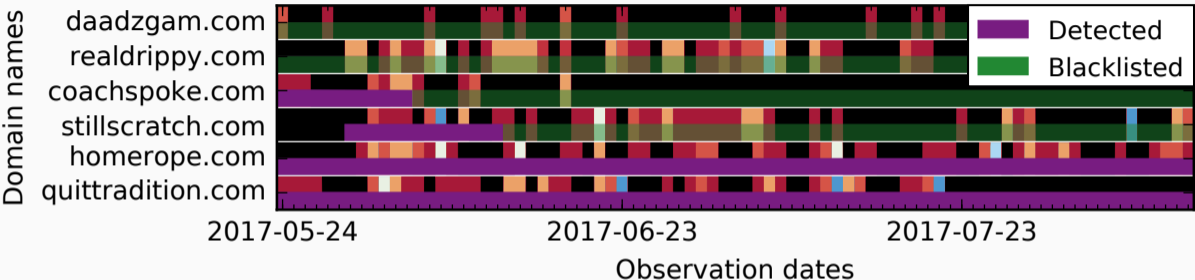


SURF Results



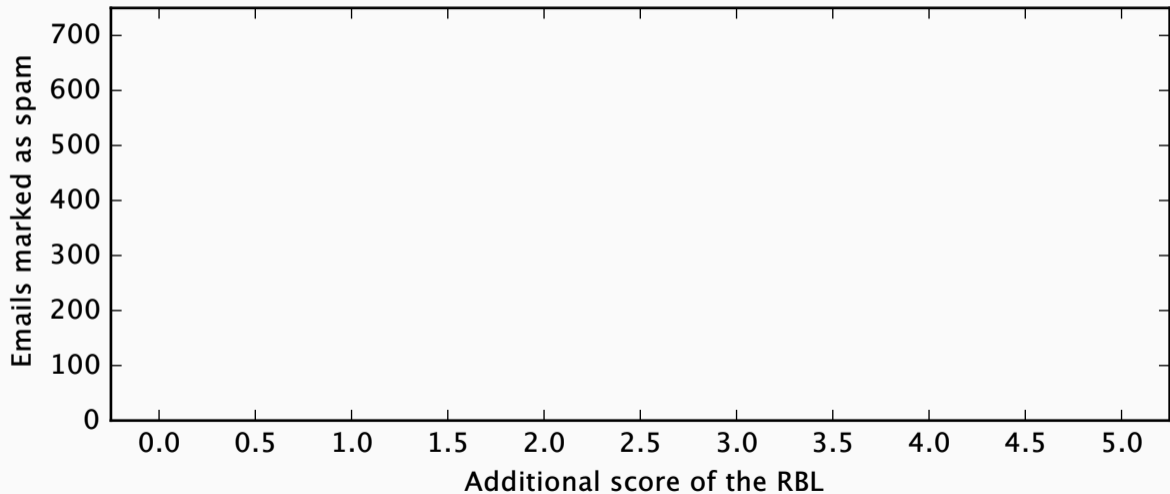
- 1080 emails
- 447 (41.39%) emails with a score of five or higher

SURF Results

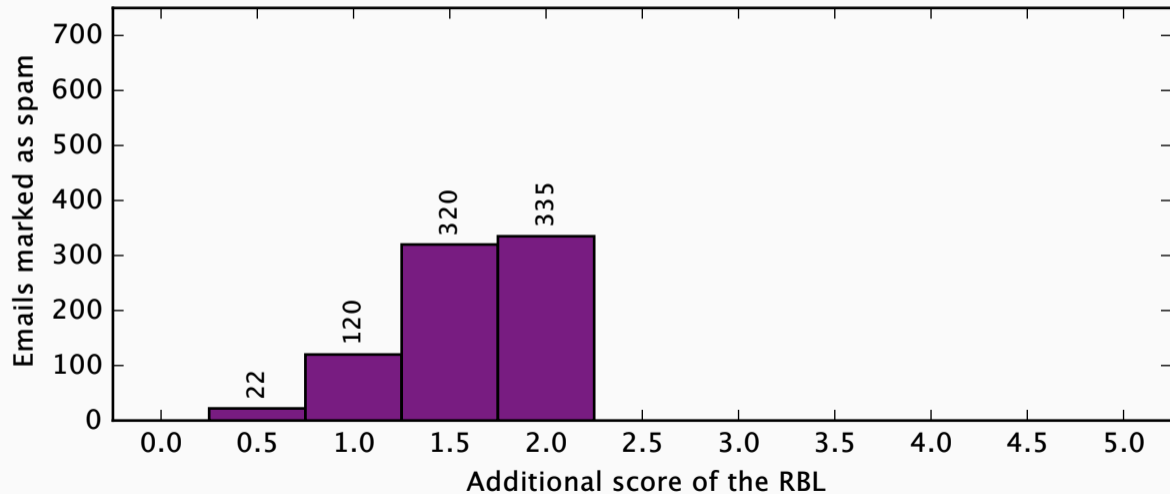


- 633 (58.61%) emails have a score below five
- 52 unique domains in the body
- of which 13 domains have never appeared in an email classified as spam
- these 13 domains appeared in 31 emails (2.87%)

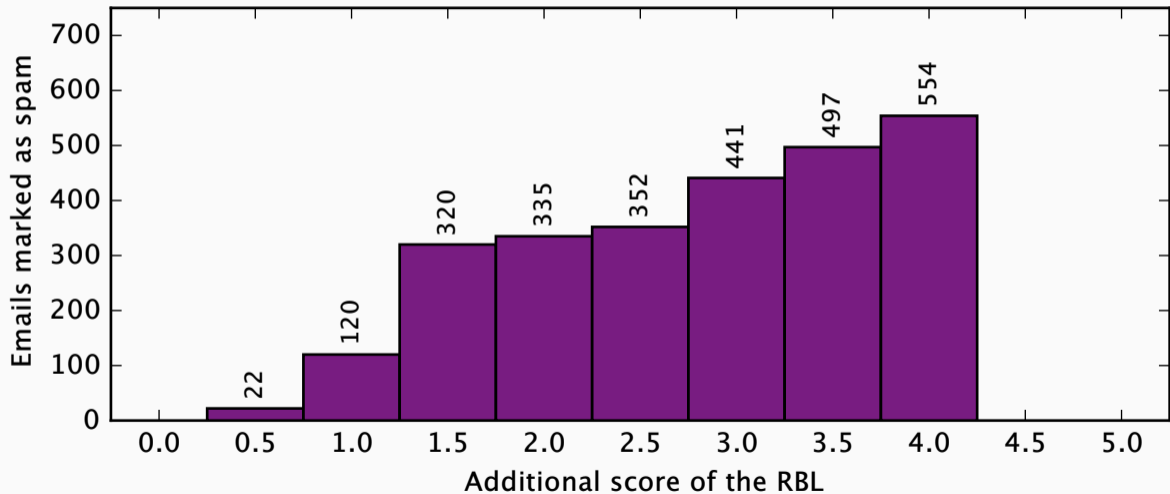
Additional email blocked



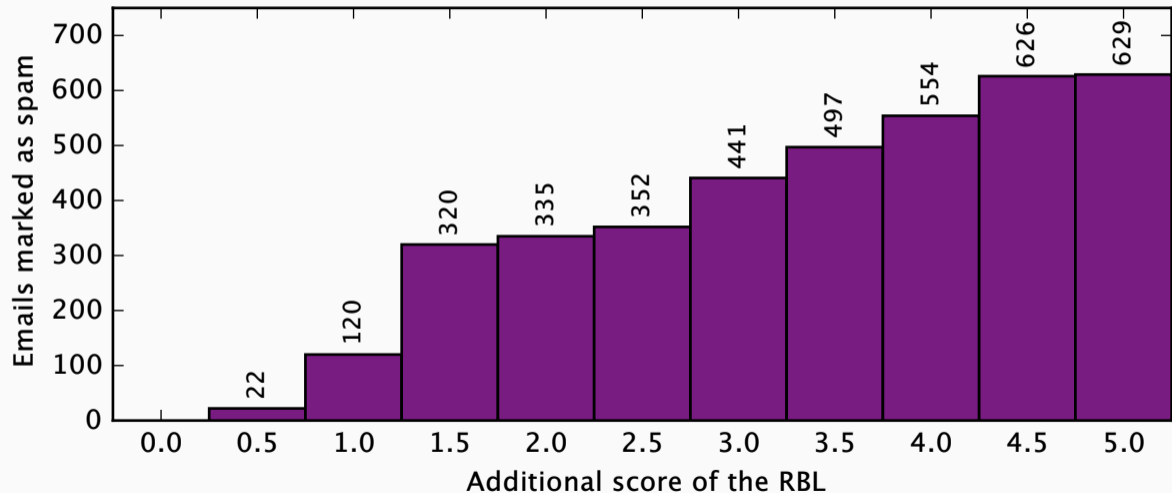
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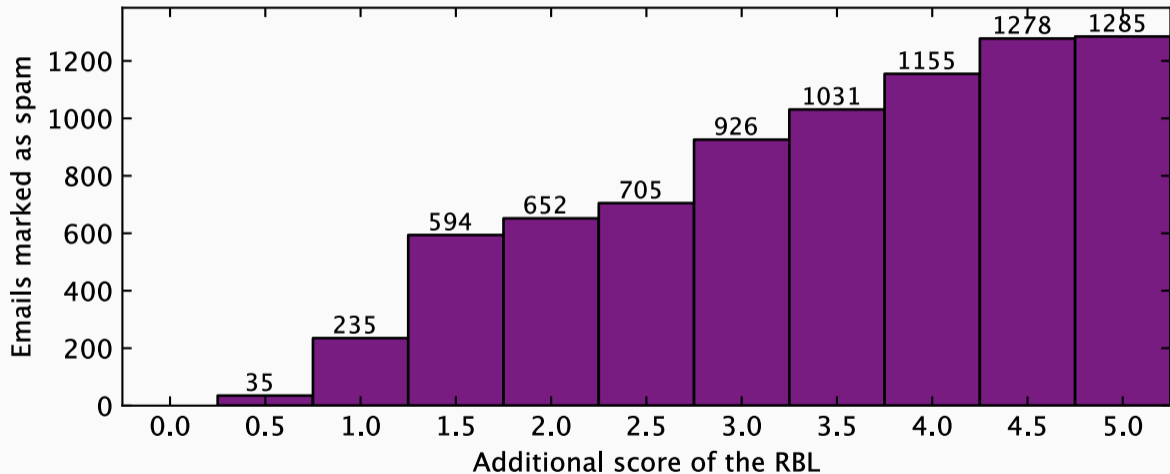
Additional email blocked



Additional email blocked



Additional email blocked (update)



Conclusions

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