

# Optimized Invariant Representation of Network Traffic for Detecting Unseen Malware Variants



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# Motivation – Network Security Challenges

- Large variability of malicious samples
  - 100k new or modified malware samples every day
- Lack of labeled data (obtaining additional labels is costly)
  - Most of existing methods rely on signature matching or feeds
    - ↑ High precision
    - ↓ Low recall (detect only known threats)
- Behavior changes introduce problems when training detectors
  - Attackers change the behavior frequently to remain undetected

# Malicious Traffic and HTTP(S)



# Our Goal

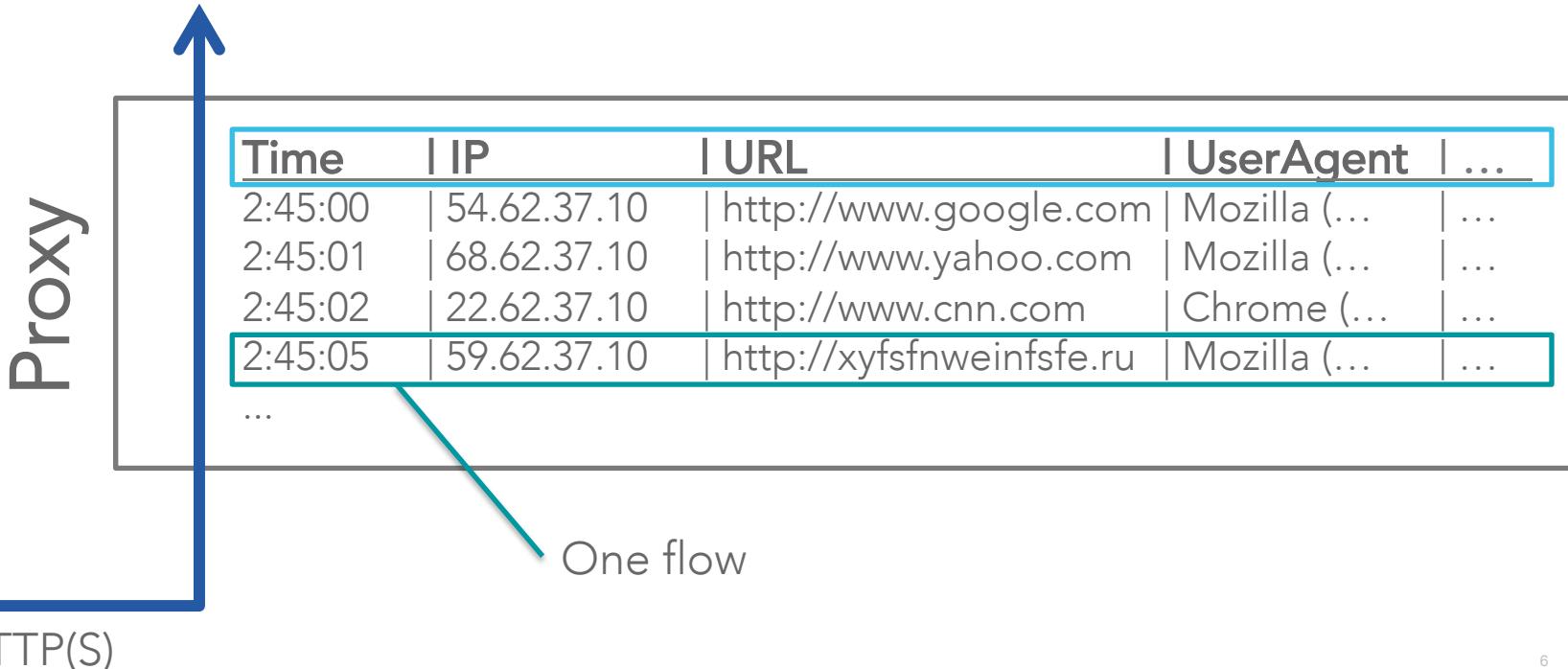
Build a representation of malware behavior robust to most of the modifications done by the attacker in the future:

# Our Goal

Build a representation of malware behavior robust to most of the modifications done by the attacker in the future:

- Change in malicious code, payload, obfuscation
- Change in hostname or server IP address
- Change in the intensity
- Change in timing
- Change in URL path, parameters, etc.

# Input data – proxy log records



# Flows are Grouped into Bags

BAG

= Flows from one user/device to one hostname in the given time interval  
Contains user-hostname communication

# Malware Bags

1.48M malware flows

15 330 malware bags

35%



## Single-flow bags

5.4k flows

5 404 bags



34% of legit

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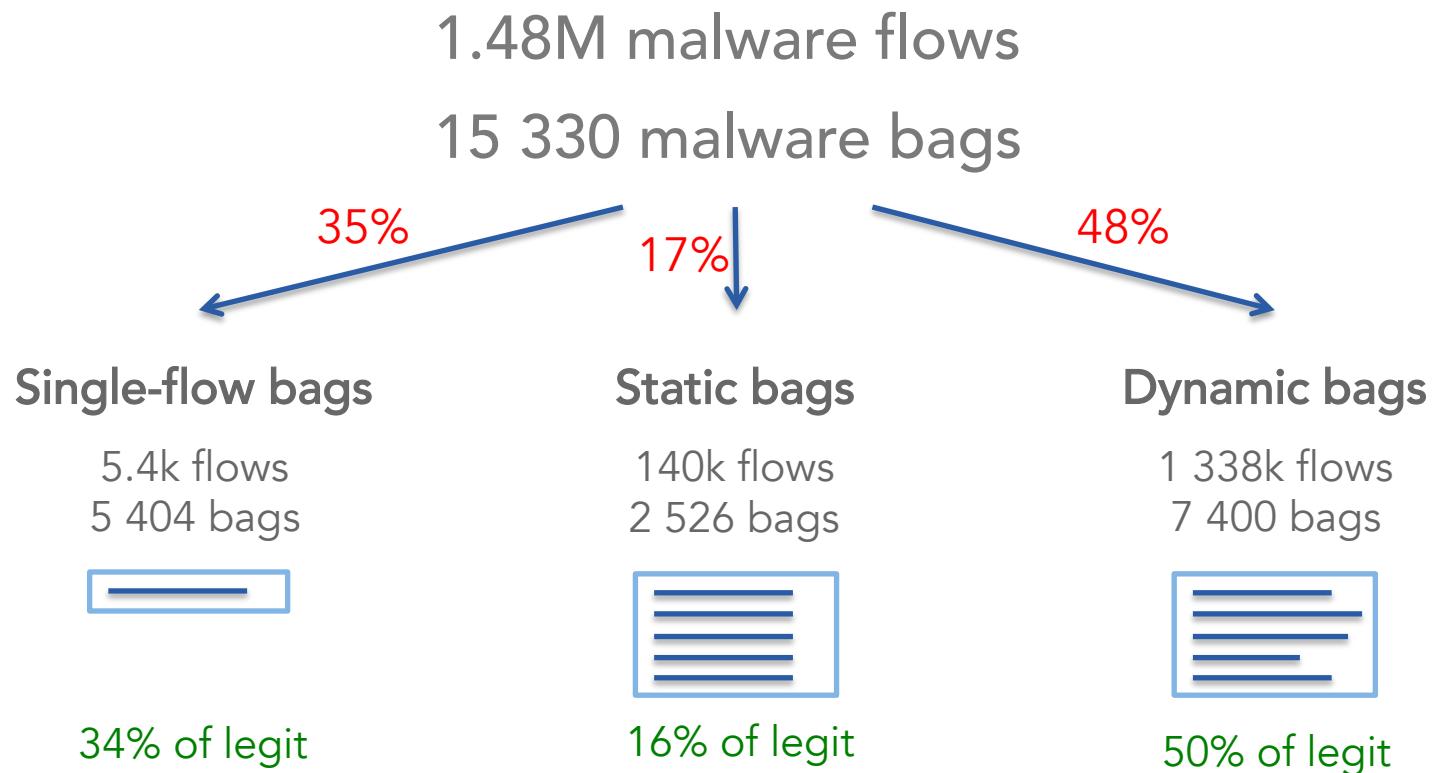
## Static bags

140k flows  
2 526 bags



16% of legit

# Malware Bags



# Malware Bags

## Single-flow bags

5.4k flows  
5 404 bags



Percent of  
malware bags:      35%

Features:      Flow-based

## Static bags

140k flows  
2 526 bags

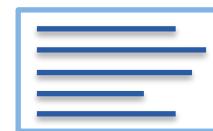


17%

Flow-based

## Dynamic bags

1 338k flows  
7 400 bags



48%

Flow-based

# Malware Bags

## Single-flow bags

5.4k flows  
5 404 bags

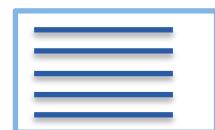


Percent of  
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## Static bags

140k flows  
2 526 bags



17%

Flow-based

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7 400 bags



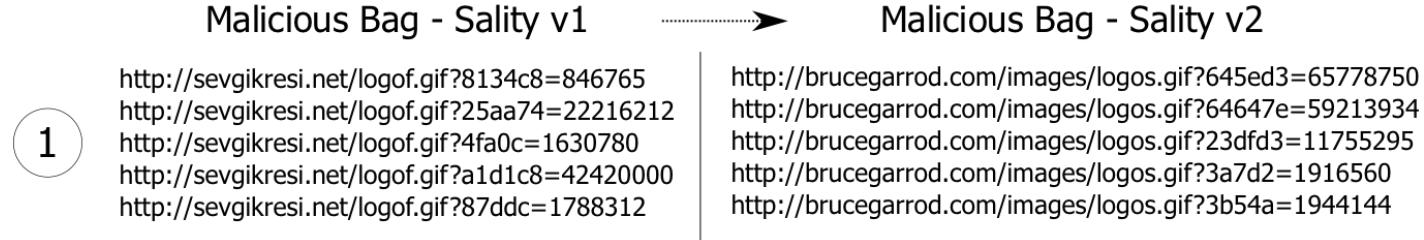
48%

~~Flow-based~~

**Bag-based Features**

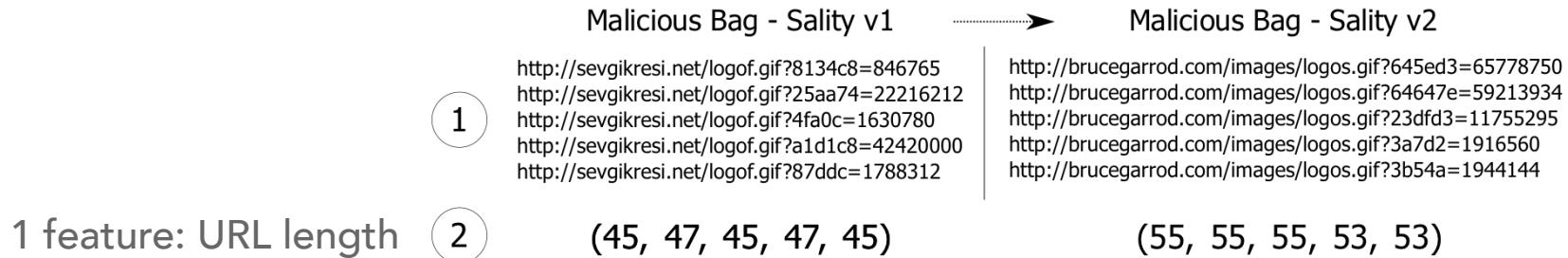
# Malware Changes – Example

Network traffic of two malware bags of the same type

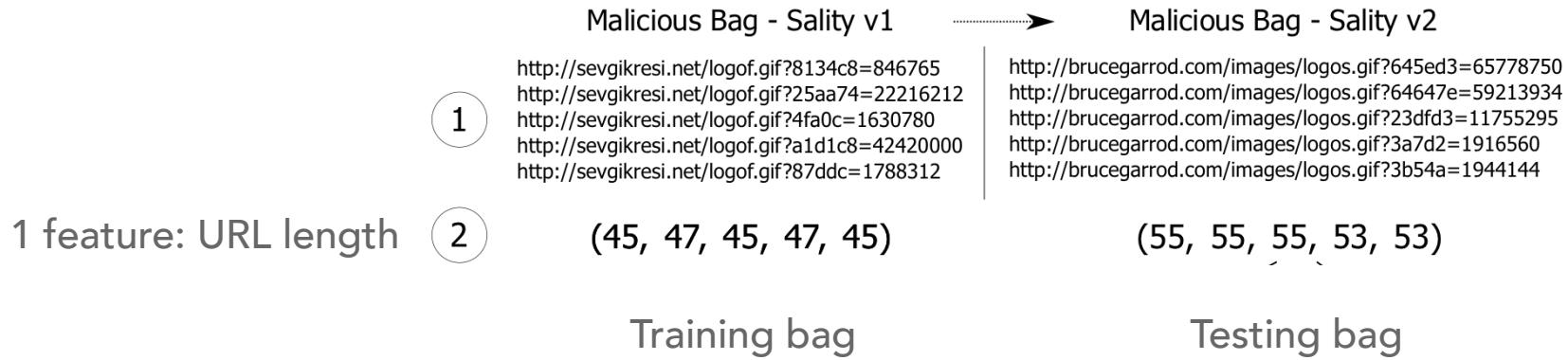


Similar or different?

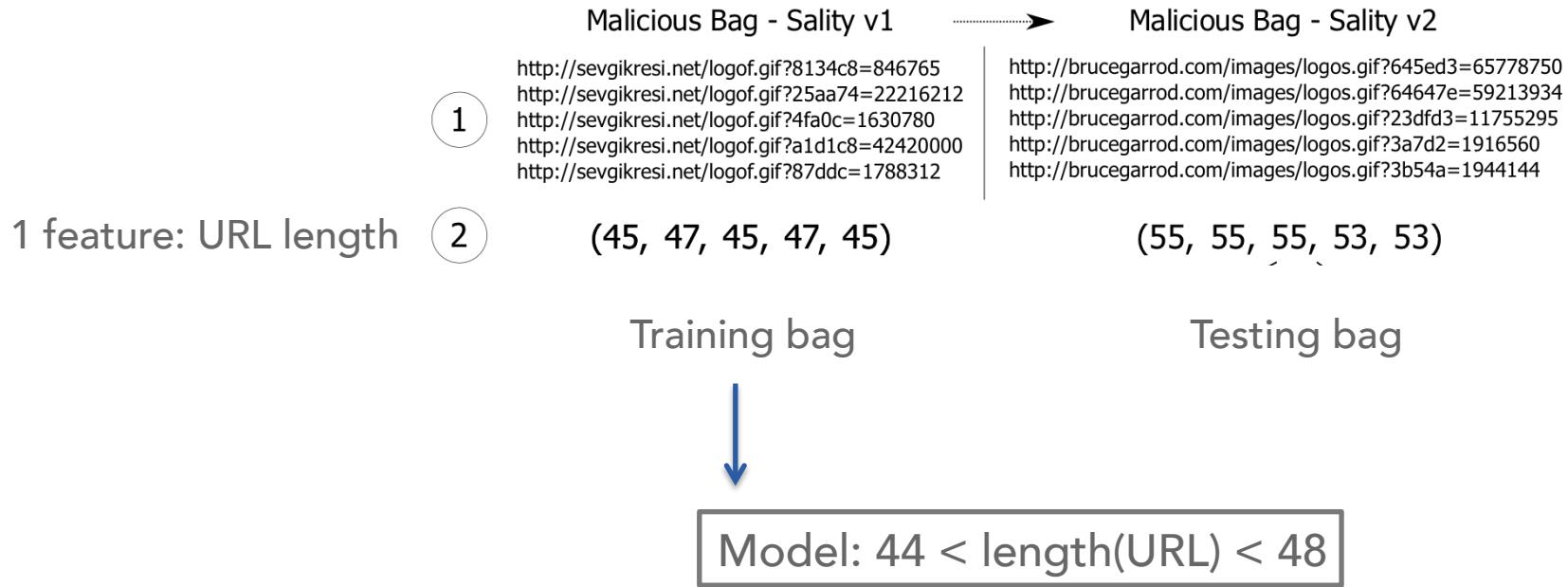
# Flow-based features



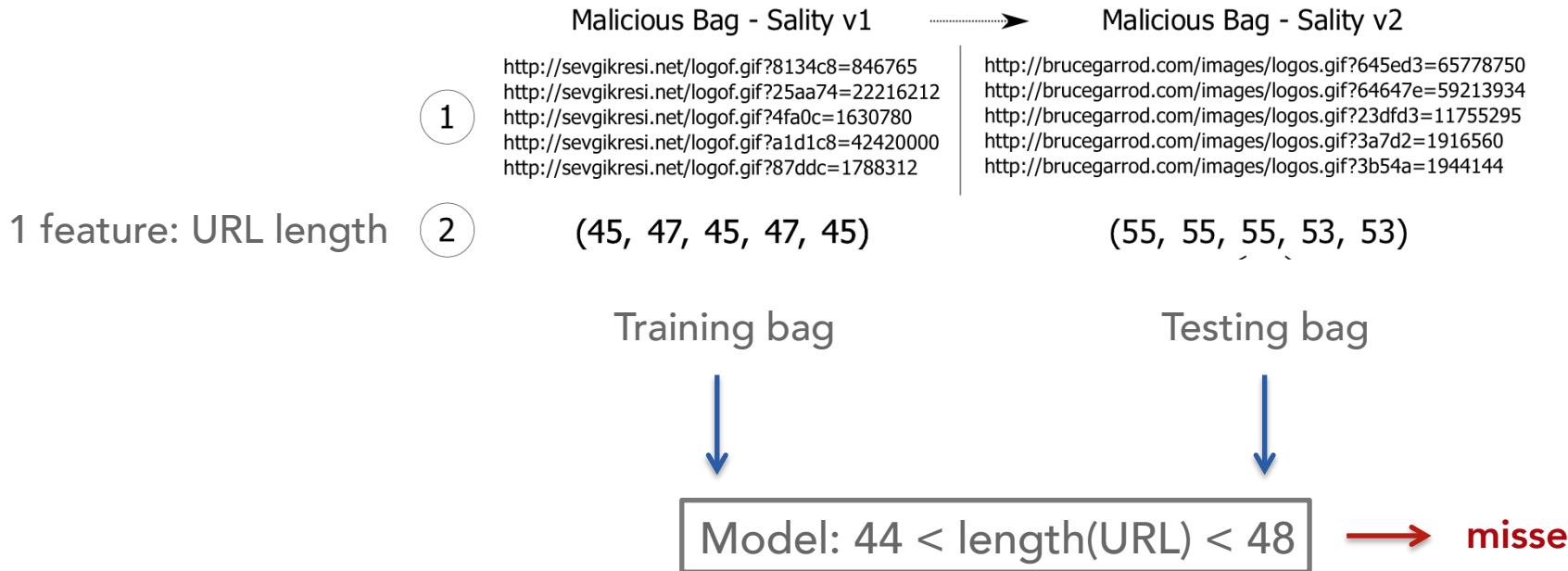
# Flow-based features



# Flow-based features



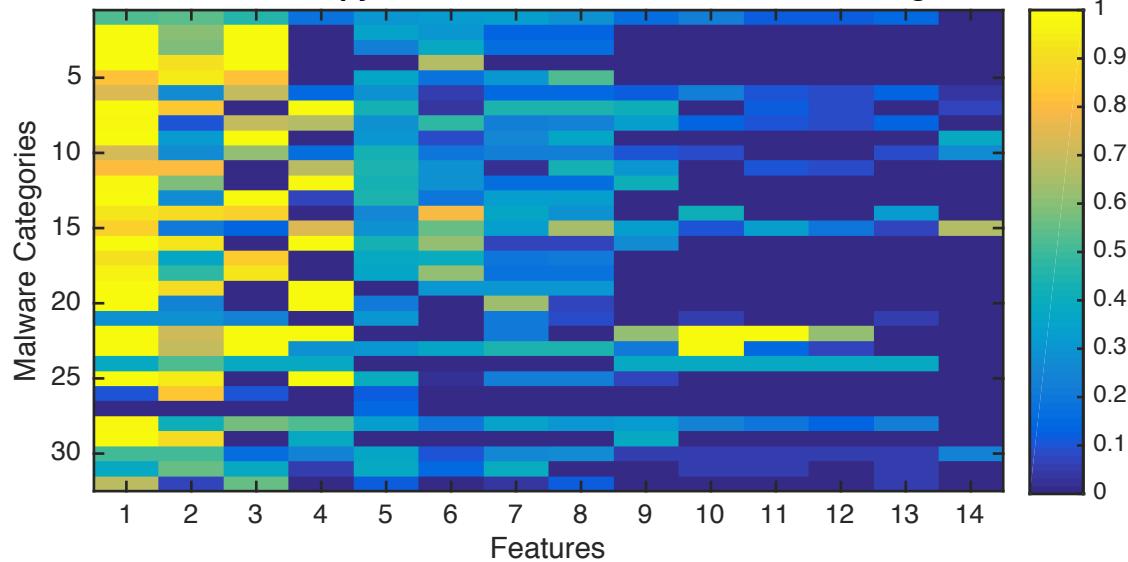
# Flow-based features



Intuition: Flow-based features are not suitable for dynamic bags.

# High Variability of Flow-based Features

Normalized Entropy of Feature Values for 32 Malware Categories



Yellow = high variability

Features:

- 1 – URL string
- 2 – Thinking time
- 3 – URL query values
- 4 – URL path
- 5 – Number of flows
- 6 – SC bytes
- 7 – Server IP
- 8 – Hostname
- 9 – URL path length
- 10 – URL query names
- 11 – URL filename
- 12 – URL filename length
- 13 – Number of URL query params
- 14 – Cs bytes

Categories: Asterope, Bedep, Dridex, Gamarue, InstallCore,  
Mudrop, MultiPlug, Poweliks, Rerdom, Ramnit,  
Rovnix, Sality, Tempedreve, Upatre, Vawtrak, Wowlik, ...

# Histogram

Malicious Bag - Sality v1



Malicious Bag - Sality v2

1

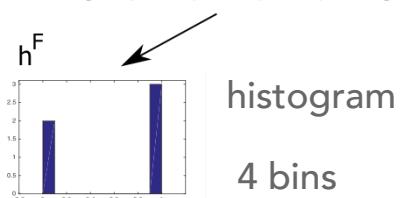
<http://sevgikresi.net/logof.gif?8134c8=846765>  
<http://sevgikresi.net/logof.gif?25aa74=22216212>  
<http://sevgikresi.net/logof.gif?4fa0c=1630780>  
<http://sevgikresi.net/logof.gif?a1d1c8=42420000>  
<http://sevgikresi.net/logof.gif?87ddc=1788312>

2

(45, 47, 45, 47, 45)

(55, 55, 55, 53, 53)

3



4

(0.4, 0, 0, 0.6)

4 feature values

# Histogram

Malicious Bag - Sality v1

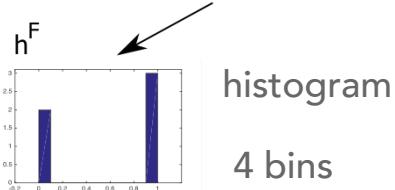
1

<http://sevgikresi.net/logof.gif?8134c8=846765>  
<http://sevgikresi.net/logof.gif?25aa74=22216212>  
<http://sevgikresi.net/logof.gif?4fa0c=1630780>  
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<http://sevgikresi.net/logof.gif?87ddc=1788312>

2

(45, 47, 45, 47, 45)

3



4

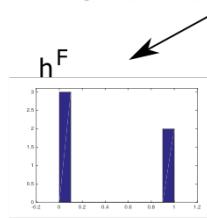
(0.4, 0, 0, 0.6)

4 feature values

Malicious Bag - Sality v2

<http://brucegarrod.com/images/logos.gif?645ed3=65778750>  
<http://brucegarrod.com/images/logos.gif?64647e=59213934>  
<http://brucegarrod.com/images/logos.gif?23dfd3=11755295>  
<http://brucegarrod.com/images/logos.gif?3a7d2=1916560>  
<http://brucegarrod.com/images/logos.gif?3b54a=1944144>

(55, 55, 55, 53, 53)



(0.6, 0, 0, 0.4)      missed?

# What is better?

	Malicious Bag - Sality v1	→	Malicious Bag - Sality v2
1	<a href="http://sevgikresi.net/logof.gif?8134c8=846765">http://sevgikresi.net/logof.gif?8134c8=846765</a> <a href="http://sevgikresi.net/logof.gif?25aa74=22216212">http://sevgikresi.net/logof.gif?25aa74=22216212</a> <a href="http://sevgikresi.net/logof.gif?4fa0c=1630780">http://sevgikresi.net/logof.gif?4fa0c=1630780</a> <a href="http://sevgikresi.net/logof.gif?a1d1c8=42420000">http://sevgikresi.net/logof.gif?a1d1c8=42420000</a> <a href="http://sevgikresi.net/logof.gif?87ddc=1788312">http://sevgikresi.net/logof.gif?87ddc=1788312</a>		<a href="http://brucegarrod.com/images/logos.gif?645ed3=65778750">http://brucegarrod.com/images/logos.gif?645ed3=65778750</a> <a href="http://brucegarrod.com/images/logos.gif?64647e=59213934">http://brucegarrod.com/images/logos.gif?64647e=59213934</a> <a href="http://brucegarrod.com/images/logos.gif?23dfd3=11755295">http://brucegarrod.com/images/logos.gif?23dfd3=11755295</a> <a href="http://brucegarrod.com/images/logos.gif?3a7d2=1916560">http://brucegarrod.com/images/logos.gif?3a7d2=1916560</a> <a href="http://brucegarrod.com/images/logos.gif?3b54a=1944144">http://brucegarrod.com/images/logos.gif?3b54a=1944144</a>
2	(45, 47, 45, 47, 45)		(55, 55, 55, 53, 53)

What do we want to represent?

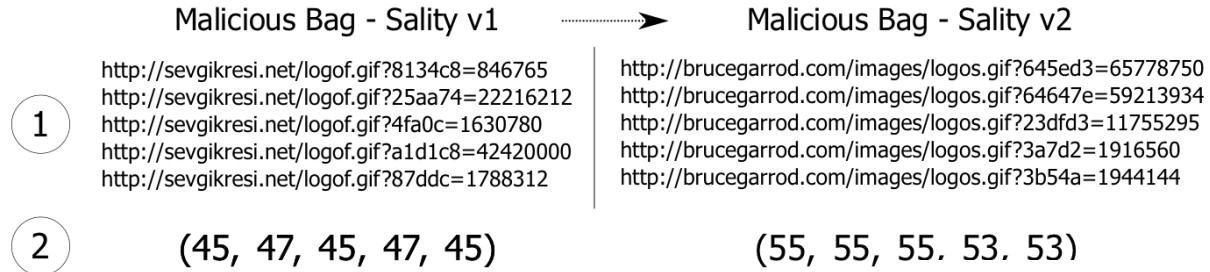
What is common across malware categories?

<b>Asterope</b>	<pre>hxwp://194.165.16.146:8080/pgt/?ver=1.3.3398&amp;id=126&amp;r=12739868&amp;os=6.1—2—8.0.7601.18571&amp;res=4—1921—466&amp;f=1 hxwp://194.165.16.146:8080/pgt/?ver=1.3.3398&amp;id=126&amp;r=15425581&amp;os=6.1—2—8.0.7601.18571&amp;res=4—1921—516&amp;f=1 hxwp://194.165.16.146:8080/pgt/?ver=1.3.3398&amp;id=126&amp;r=27423103&amp;os=6.1—2—8.0.7601.18571&amp;res=4—1921—342&amp;f=1 hxwp://194.165.16.146:8080/pgt/?ver=1.3.3753&amp;id=126&amp;r=8955018&amp;os=6.1—2—8.0.7601.18571&amp;res=4—1921—319&amp;f=1 hxwp://194.165.16.146:8080/pgt/?ver=1.3.3398&amp;id=126&amp;r=31957678&amp;os=6.1—2—8.0.7601.18571&amp;res=4—1921—223&amp;f=1</pre>
<b>Click-fraud, malvertising-related botnet</b>	<pre>hxwp://directcashfunds.com/opntrk.php?tkey=024f9730e23f8553c3e5342568a70300&amp;Email=name.surname@company.com hxwp://directcashfunds.com/opntrk.php?tkey=c1b6e3d50632d4f5c0ae13a52d3c4d8d&amp;Email=name.surname@company.com hxwp://directcashfunds.com/opntrk.php?tkey=7c9a843ce18126900c46dbe4b3b6425&amp;Email=name.surname@company.com hxwp://directcashfunds.com/opntrk.php?tkey=c1b6e3d50632d4f5c0ae13a52d3c4d8d&amp;Email=name.surname@company.com hxwp://directcashfunds.com/opntrk.php?tkey=bfba7d7023220c59d06e76f0085d6573&amp;Email=name.surname@company.com</pre>
<b>DGA</b>	<pre>hxwp://uvyqifymelapuvoh.biz/s531ka.ji5 hxwp://uvyqifymelapuvoh.biz/r159c281.x19 hxwp://uvyqifymelapuvoh.biz/seibpn6.2m0 hxwp://uvyqifymelapuvoh.biz/3854f.u17 hxwp://uvyqifymelapuvoh.biz/06hk3j.449</pre>
<b>Drindex</b>	<pre>hxwp://27.54.174.181/8qV578&amp;So@HU6Q6S/gz\$J0l=iTTH 28%2CM/we20%3D hxwp://27.54.174.181/C4GyRx%7E@RY6x /M&amp;N=sq/bW_ra4OTJ hxwp://27.54.174.181/gPvh+GO/9RPPfk0%2CzXOYU%20/Vq8Ww/+a_m%7Ez hxwp://27.54.174.181/qE0my4K1z48Cf3H8wG%7Evpz=J%26fqMI%24m/46oElP=Gjww%3D%26lB+Ar.y3 iu%2D1E/ss0 hxwp://27.54.174.181/kv7ig2s1vslfv&amp;i_&amp;/s&amp;no%2Ds83%7E%2B+s5%2D%3F%+20&amp;l/kjx%26e8x=\$.pfir@s3j66%2D</pre>
<b>InstallCore</b>	<b>Monetization</b>
<pre>hxwp://rp.any-file-opener.org/?pcrc=1559319553&amp;v=2.0 hxwp://rp.any-file-opener.org/?pcrc=1132521307&amp;v=2.0 hxwp://rp.any-file-opener.org/?pcrc=1123945956&amp;v=2.0 hxwp://rp.any-file-opener.org/?pcrc=1075608192&amp;v=2.0 hxwp://rp.any-file-opener.org/?pcrc=995719244&amp;v=2.0</pre>	<pre>hxwp://utouring.net/search/q/conducing hxwp://utouring.net/go/u/1/r/1647 hxwp://utouring.net/go/u/0/r/2675 hxwp://utouring.net/search/f/1/q/refiles hxwp://utouring.net/search/f/1/q/refiles</pre>
<b>Poweliks</b>	<pre>hxwp://31.184.194.39/query?version=1.7&amp;sid=793&amp;builddate=114&amp;q=nitric+oxide+side+effects&amp;ua=Mozilla%2F5 ... &amp;lr=7&amp;ls=0 hxwp://31.184.194.39/query?version=1.7&amp;sid=793&amp;builddate=114&amp;q=weight+loss+success+stories&amp;ua=Mozilla%2F5 ... &amp;lr=0&amp;ls=0 hxwp://31.184.194.39/query?version=1.7&amp;sid=793&amp;builddate=114&amp;q=shoulder+pain&amp;ua=Mozilla%2F5 ... &amp;lr=7&amp;ls=2 hxwp://31.184.194.39/query?version=1.7&amp;sid=793&amp;builddate=114&amp;q=cheap+car+insurance&amp;ua=Mozilla%2F5 ... &amp;lr=7&amp;ls=2 hxwp://31.184.194.39/query?version=1.7&amp;sid=793&amp;builddate=114&amp;q=natural+testosterone+boosters&amp;ua=Mozilla%2F5 ... &amp;lr=7&amp;ls=2</pre>
<b>Zeus</b>	<pre>hxwp://130.185.106.28/m/lbQFdXVjiriLva4KHeNpWCmThrJBn3f34HNwsLVVsUmLXtsumSSPe/zzXtu9SzvjI9zKlxde ... 3RqvGzKN5 hxwp://130.185.106.28/m/lbQfUvJzEn4vx4KHeNpWCmThrBn3f34HNwsLVVsUmLfk0Pass+S+zzXtu9SzvjI9zKlxde ... 3vKwmk0oUi hxwp://130.185.106.28/m/lbQfUvJzJwBX4KHeNpWCmThrBn3f34HNwsLVVsUmKH7ue2StVsKzzXtu9SzvjI9zKlxde ... 3vKwmk0oUi hxwp://130.185.106.28/m/lbQntVVji5/7Yp4KHeNpWCmThrJBn3f34HNwsLVVsUmLz4sO6YRvOjzzXtu9SzvjI9zKlxde ... 3zB9057quqv hxwp://130.185.106.28/m/lbQG9VVVjjSnDM94KHeNpWCmThrJBn3f34HNwsLVVsUmLxp/+YRue8zzXtu9SzvjI9zKlxde ... 6iN5mt6Tj3</pre>
<b>Legitimate traffic</b>	<pre>hxwp://www.cnn.com/a/1.73.0/js/vendor/usabilla.min.js hxwp://www.cnn.com/element/ssi/auto/4.0/sect/MAIN/markets_wsod_expansion.html hxwp://www.cnn.com/a/1.73.0/assets/sprite-s1dced3ff2b.png hxwp://www.cnn.com/element/widget/video/videoapi/api/latest/js/CNNVideoBootstrapper.js hxwp://www.cnn.com/jsonp/video/nowPlayingSchedule.json?callback=nowPlayingScheduleCallbackWrapper&amp;_=1422885578476</pre>
<b>Legitimate traffic</b>	<pre>hxwp://ads.adaptv.advertising.com/a/h/7_g_doK40WLPYMHbkD9G2u7HSXjqzIaa7Bqhslod+u7iQi ... &amp;context=fullUrl%3Dpandora.com hxwp://ads.adaptv.advertising.com/crossdomain.xml hxwp://ads.advertising.com/411f1e96-3bde-4d85-b17e-63749e5f0695.js hxwp://ads.advertising.com/ids/411f1e96-3bde-4d85-b17e-63749e5f0695 hxwp://ads.adaptv.advertising.com/applist?placementId=297920&amp;key=&amp;d.vw=1&amp;orgId=8656&amp;hostname=data.rtbfy.com</pre>

MALICIOUS

LEGITIMATE

# What is better?

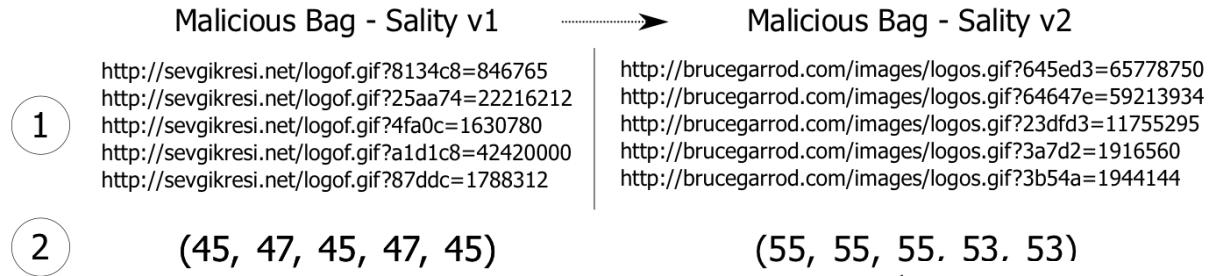


What do we want to represent?

What is common across malware categories?

Usually parameter names and subdomains are not stable, but the **URL structure** usually remains the same.

# Example



Malware dynamics:

It's common for many mw categories and different from most of legitimate traffic

How?

# Example

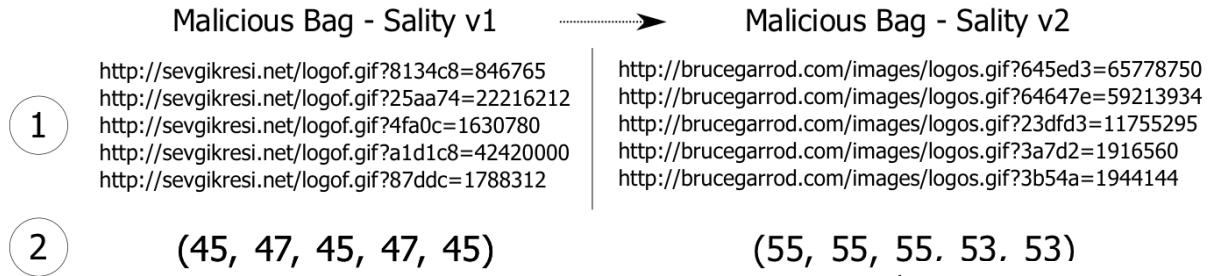
	Malicious Bag - Sality v1	→	Malicious Bag - Sality v2
1	<a href="http://sevgikresi.net/logof.gif?8134c8=846765">http://sevgikresi.net/logof.gif?8134c8=846765</a> <a href="http://sevgikresi.net/logof.gif?25aa74=22216212">http://sevgikresi.net/logof.gif?25aa74=22216212</a> <a href="http://sevgikresi.net/logof.gif?4fa0c=1630780">http://sevgikresi.net/logof.gif?4fa0c=1630780</a> <a href="http://sevgikresi.net/logof.gif?a1d1c8=42420000">http://sevgikresi.net/logof.gif?a1d1c8=42420000</a> <a href="http://sevgikresi.net/logof.gif?87ddc=1788312">http://sevgikresi.net/logof.gif?87ddc=1788312</a>		<a href="http://brucegarrod.com/images/logos.gif?645ed3=65778750">http://brucegarrod.com/images/logos.gif?645ed3=65778750</a> <a href="http://brucegarrod.com/images/logos.gif?64647e=59213934">http://brucegarrod.com/images/logos.gif?64647e=59213934</a> <a href="http://brucegarrod.com/images/logos.gif?23dfd3=11755295">http://brucegarrod.com/images/logos.gif?23dfd3=11755295</a> <a href="http://brucegarrod.com/images/logos.gif?3a7d2=1916560">http://brucegarrod.com/images/logos.gif?3a7d2=1916560</a> <a href="http://brucegarrod.com/images/logos.gif?3b54a=1944144">http://brucegarrod.com/images/logos.gif?3b54a=1944144</a>
2	(45, 47, 45, 47, 45)		(55, 55, 55, 53, 53)

Malware dynamics:  
It's common for many mw categories and different from most of legitimate traffic

Parallel to action recognition:



# Example



## Malware dynamics:

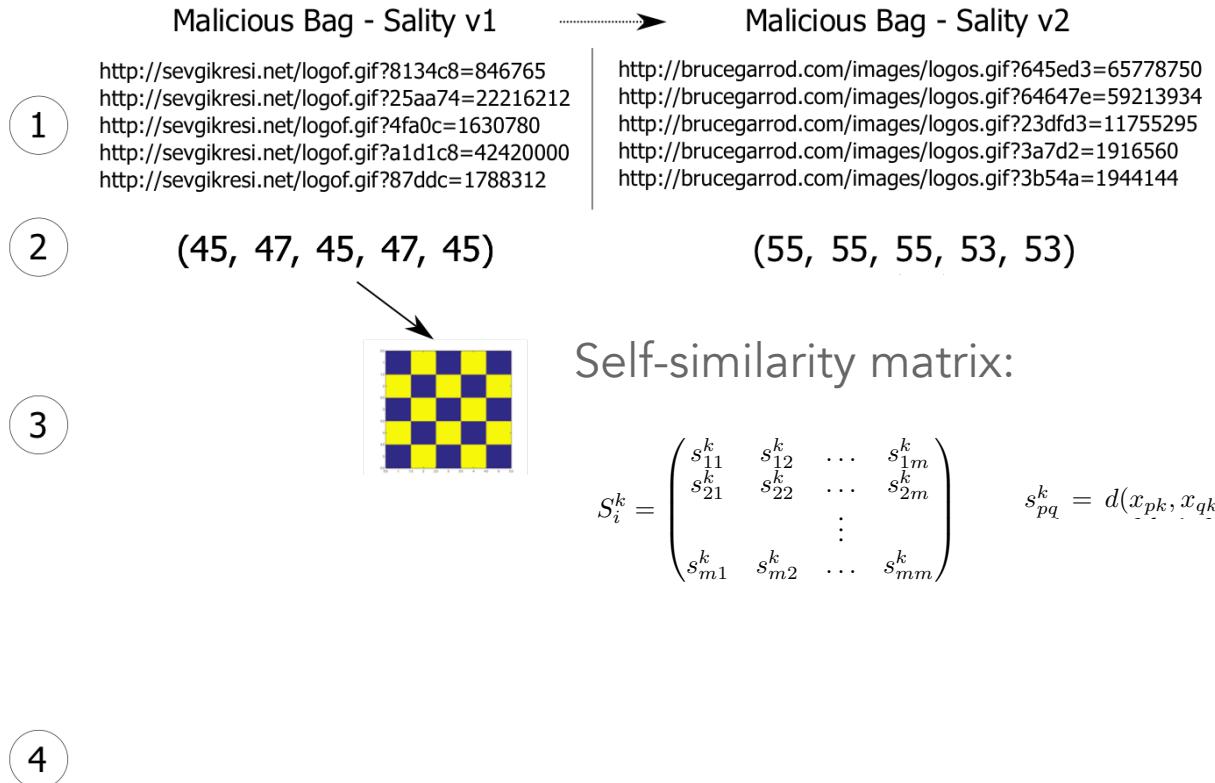
It's common for many mw categories and different from most of legitimate traffic

Parallel to action recognition:  
Each bag (set of mw flows) is an action of mw

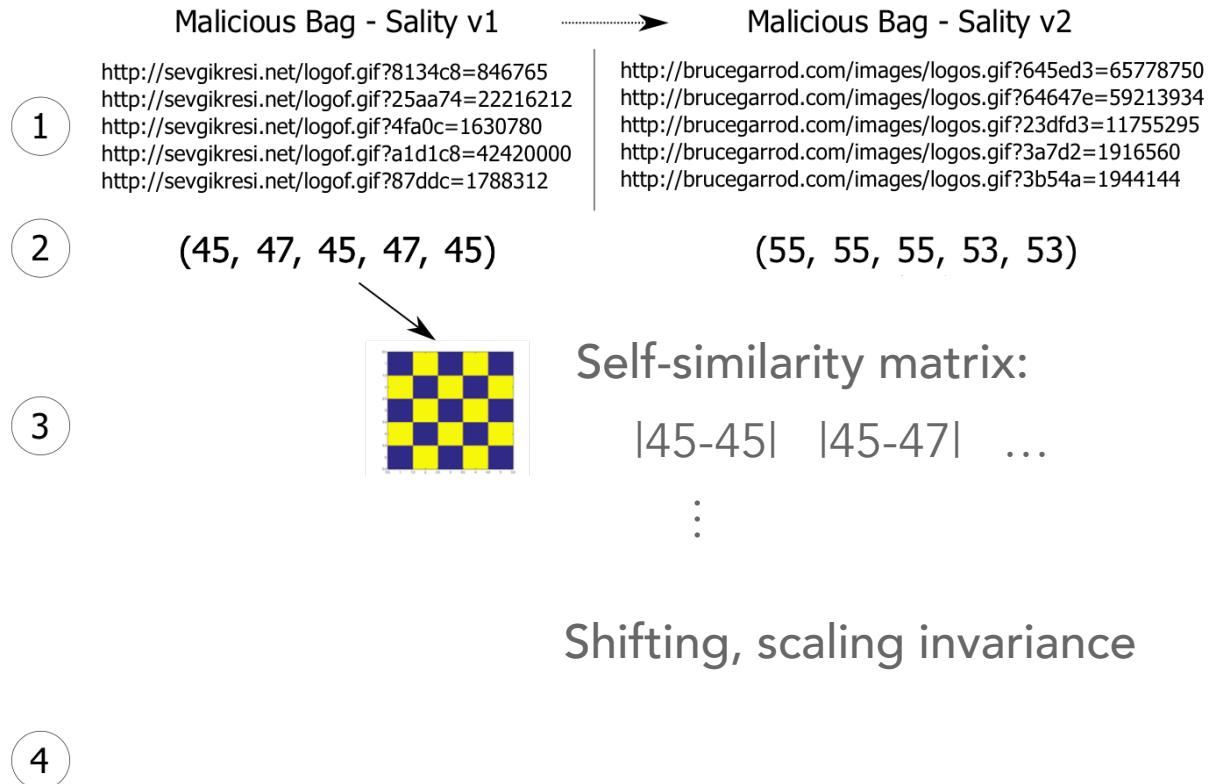
1 image  $\approx$  1 flow

Action recognition can be solved with self-similarity matrix

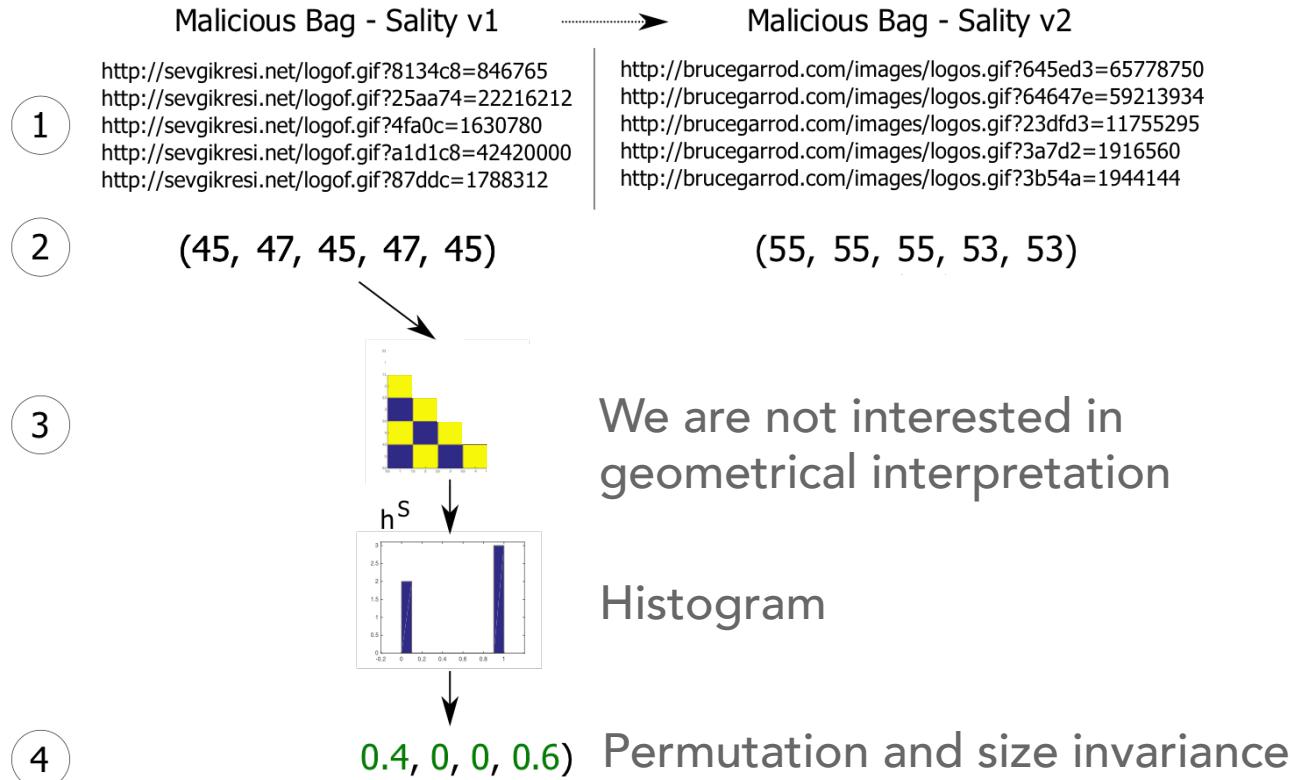
# Example



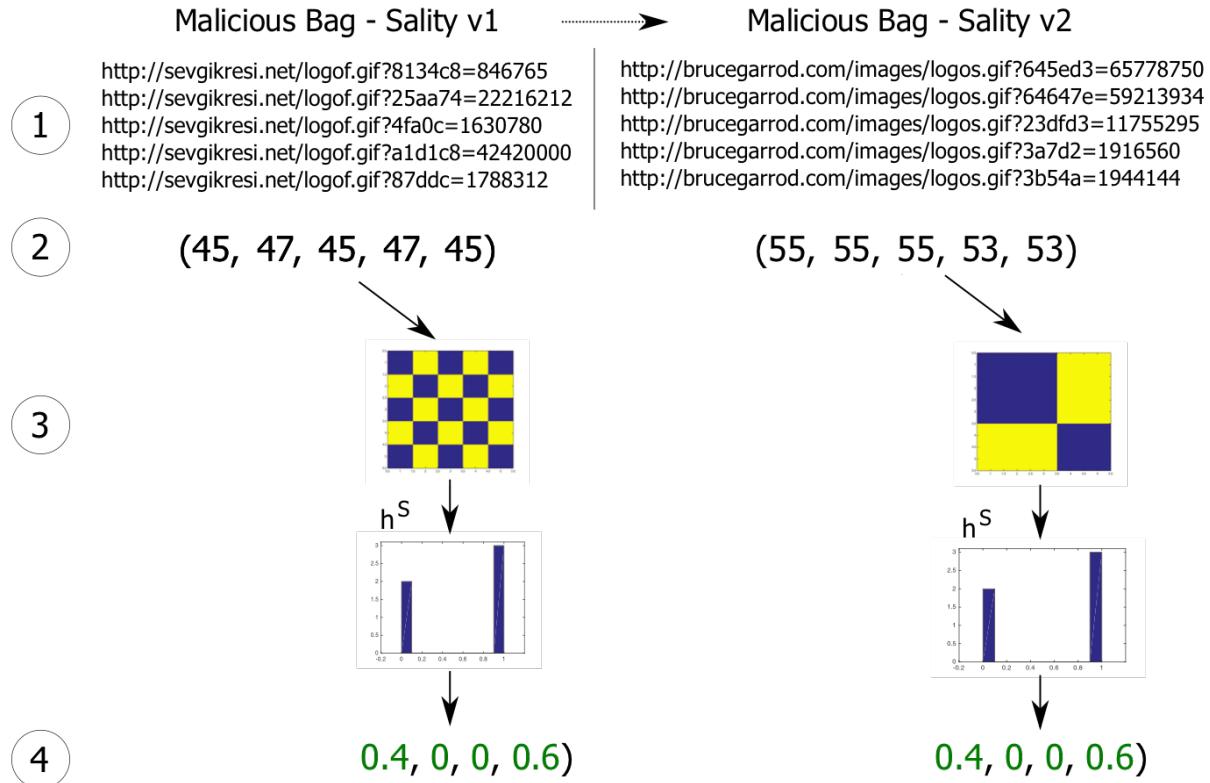
# Example



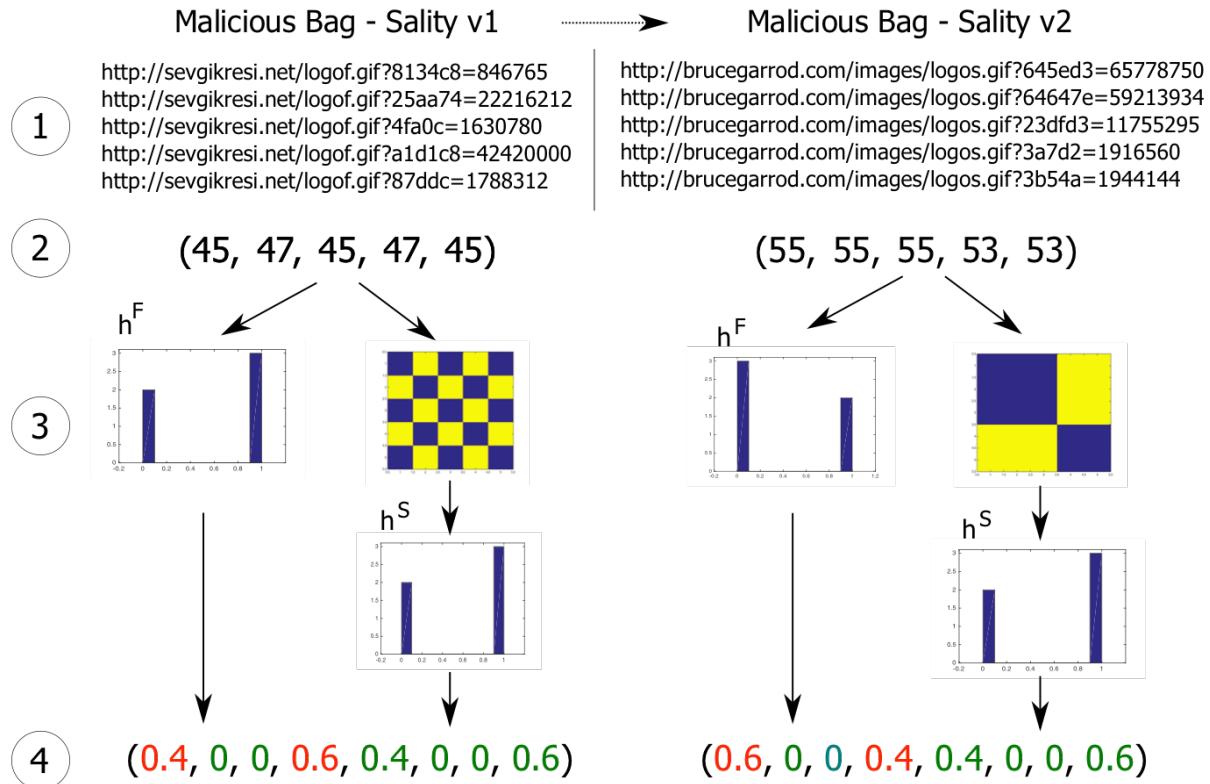
# Example



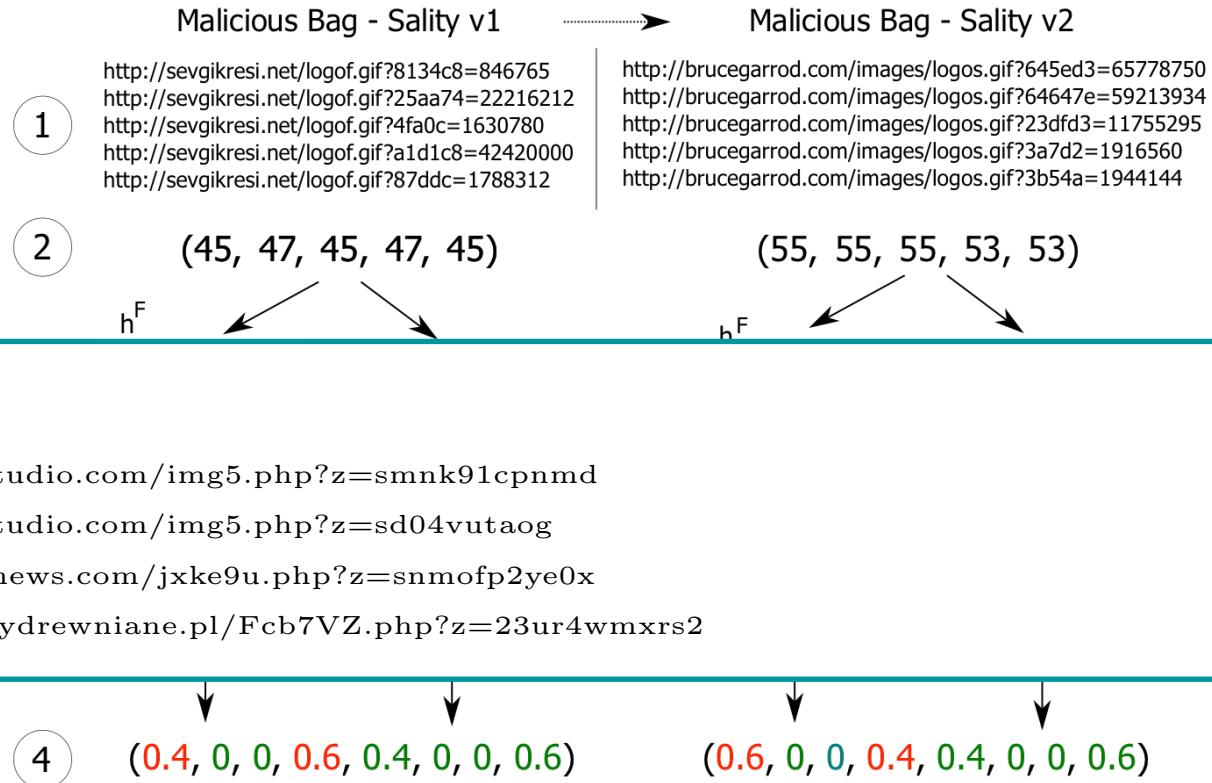
# Example



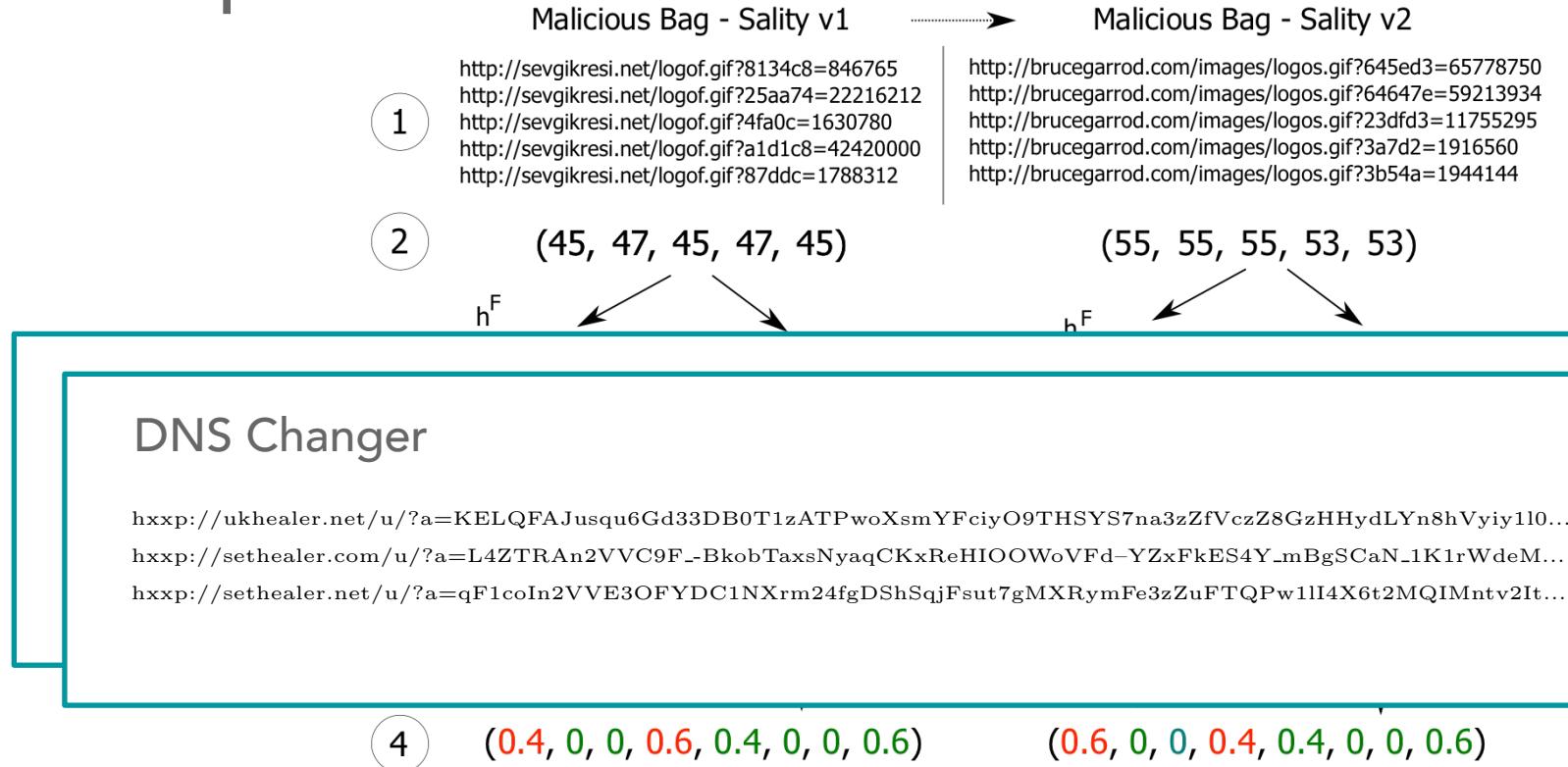
# Example



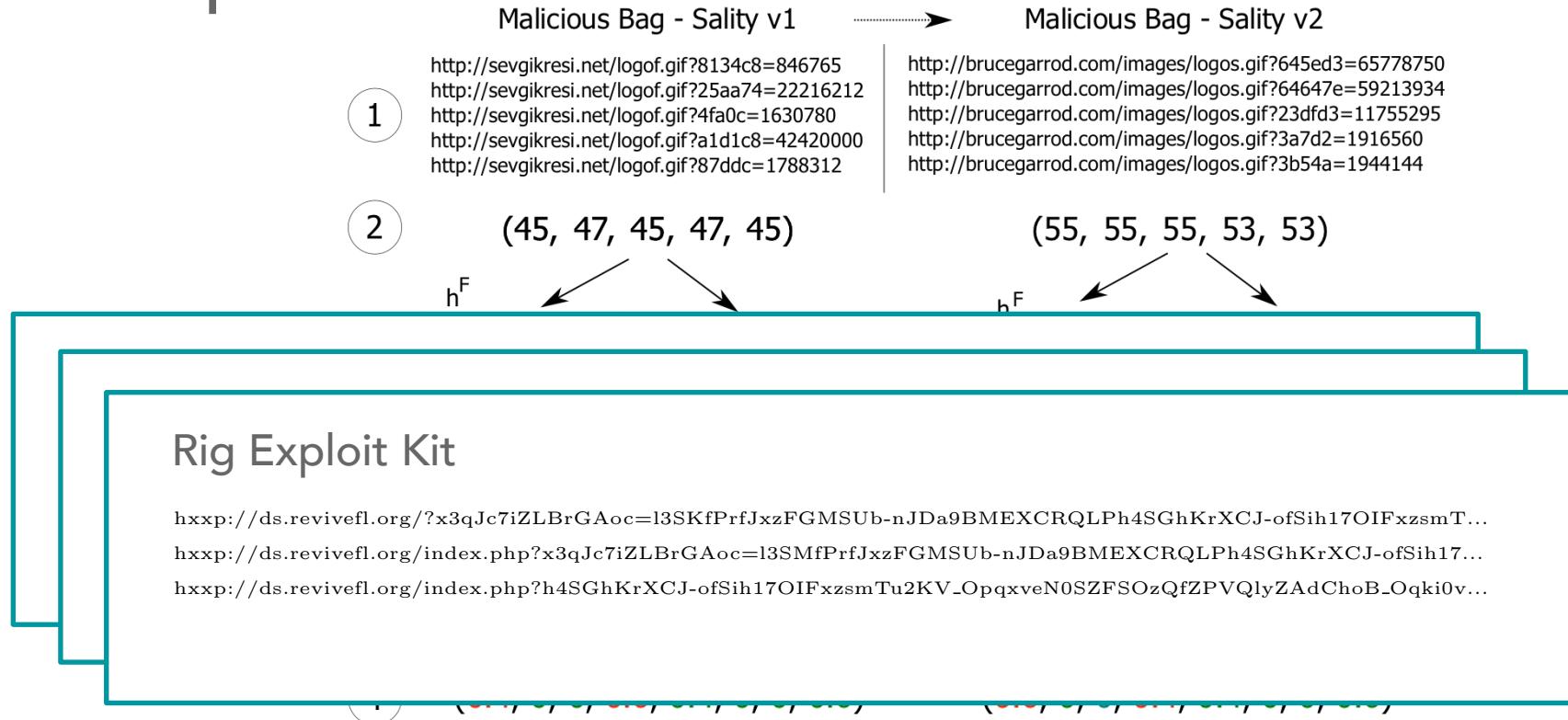
# Example



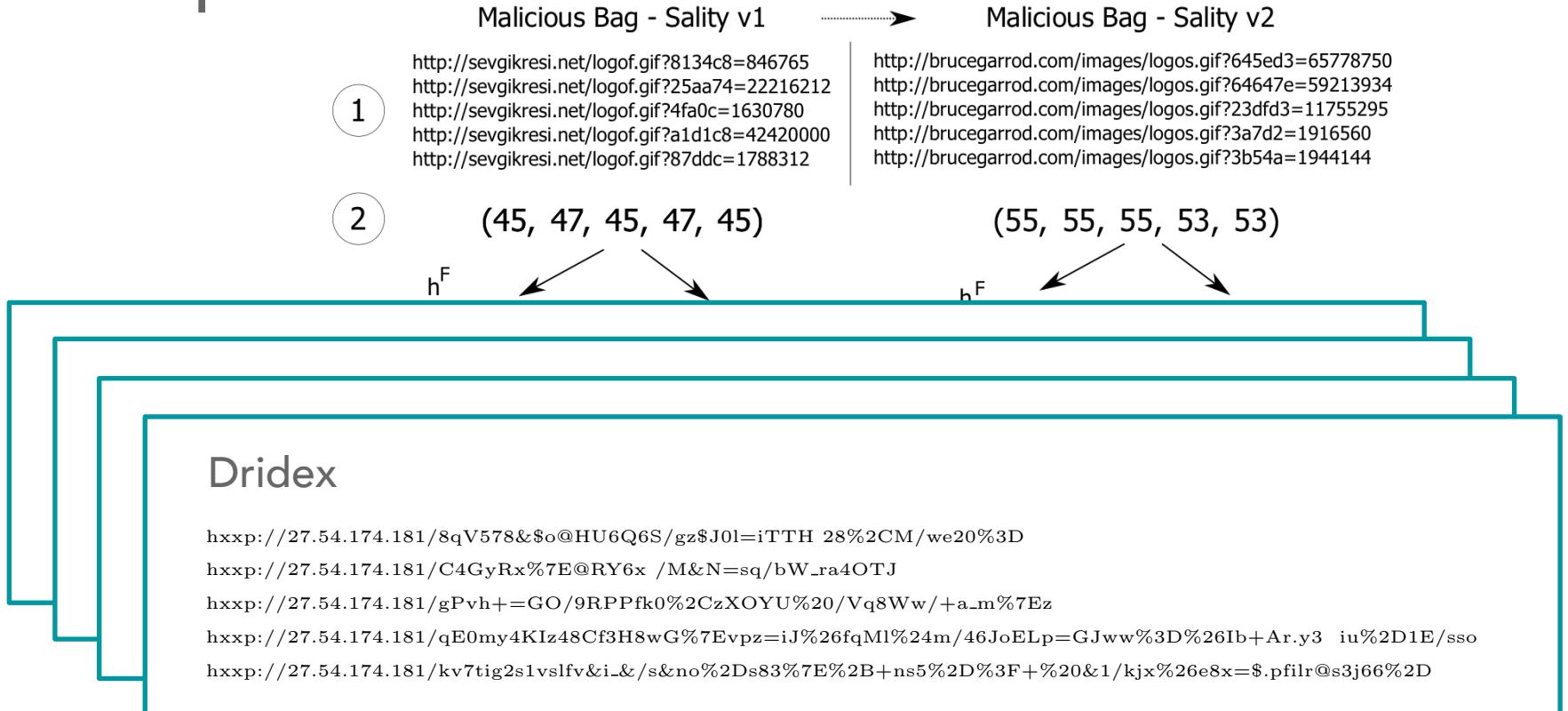
# Example



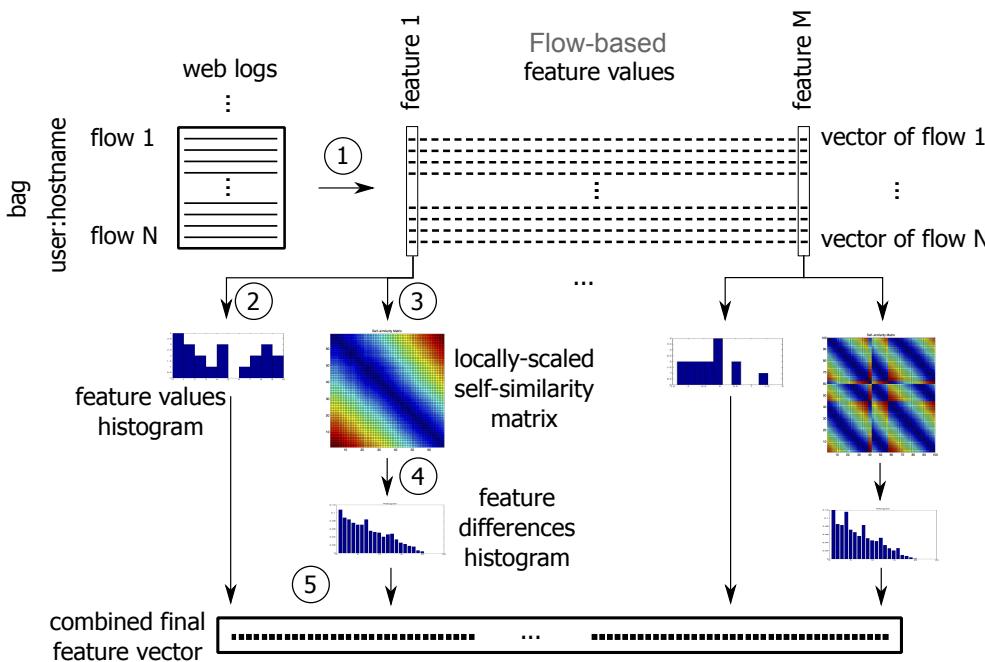
# Example



# Example



# Overview



- 1 – create bag  
+ extract flow-based feature vectors
- 2 – create feature values histogram
- 3 – create self-similarity matrix
- 4 – create feature differences histogram
- 5 – combine into final feature vector

# Invariant to the following changes:

- Malicious code, payload, obfuscation
- Server or hostname
- URL path or filename
- Names, values, or number of URL parameters
- Encoded URL content
- Number of flows
- Thinking time
- Ordering of flows
- Size of flows

## InstallCore

```
hxxp://rp.any-file-opener.org/?pcrc=1559319553&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=1132521307&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=1123945956&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=1075608192&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=995719244&v=2.0
```



## Asterope

```
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3398&id=126&r=12739868&os=6.1—2—8.0.7601.18571&res=4—1921—466&f=1  
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3398&id=126&r=15425581&os=6.1—2—8.0.7601.18571&res=4—1921—516&f=1  
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3398&id=126&r=274223103&os=6.1—2—8.0.7601.18571&res=4—1921—342&f=1  
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3753&id=126&r=8955018&os=6.1—2—8.0.7601.18571&res=4—1921—319&f=1  
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3398&id=126&r=31957678&os=6.1—2—8.0.7601.18571&res=4—1921—223&f=1
```

# Not invariant to the following changes:

- Static behavior is not considered
- Multiple behaviors in a bag
- Encrypted HTTPS traffic
- Real-time changes and fast evolution

## InstallCore

```
hxxp://rp.any-file-opener.org/?pcrc=1559319553&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=1132521307&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=1123945956&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=1075608192&v=2.0  
hxxp://rp.any-file-opener.org/?pcrc=995719244&v=2.0
```



?

```
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3398&id=126&r=12739868&os=6.1—2—8.0.7601.18571&res=4—1921—466&f=1  
hxxp://27.54.174.181/C4GyRx%7E@RY6x /M&N=sq/bW_ra4OTJ  
hxxp://130.185.106.28/m/lbQJFUVjgZn4vx4KHeNpWCmThrJBn3f34HNwsLVVsUmLfkoPaSS+S+zzXtIu9SzwjI9zKlxsdE ... 3vKwmk0oUi  
hxxp://uvyqifymelapuvoh.biz/rL59c281.x19  
hxxp://194.165.16.146:8080/pgt/?ver=1.3.3398&id=126&r=31957678&os=6.1—2—8.0.7601.18571&res=4—1921—223&f=1
```

# Parameters of the Representation

Number of bins, all of them are equidistant

→ major impact on the results



How to choose the correct number?

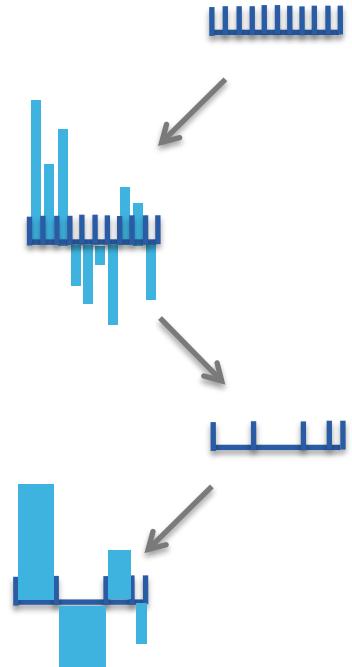
We want to learn the parameters automatically from the data

# Proposed Optimization Algorithm

- 1) Define the initial number of mini-bins (256)
- 2) Find a set of weights by solving:

$$\min_{\mathbf{w} \in \mathbb{R}^{b \cdot p}, w_0 \in \mathbb{R}} \left[ \underbrace{\gamma \sum_{i=1}^n \sum_{j=1}^{b-1} |w_{i,j} - w_{i,j+1}|}_{\text{merging}} + \underbrace{\frac{1}{m} \sum_{i=1}^m \max \{0, 1 - y^i \langle \phi(\mathbf{z}^i; \boldsymbol{\theta}), \mathbf{w} \rangle\}}_{\text{hinge loss}} \right]$$

- 3) Create new bins (by merging mini-bins)
- 4) Solve standard SVM with new bins



# Experiments – Dataset Description

Training data:

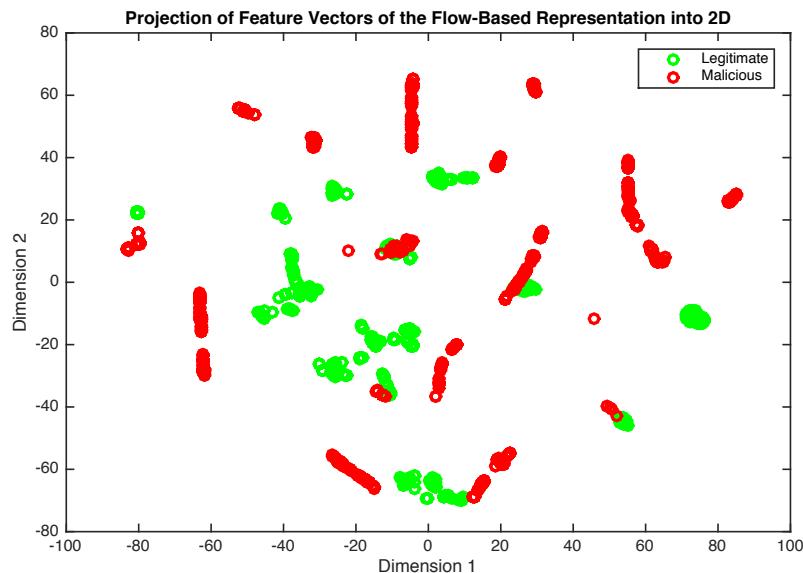
positives: 8 mw categories  
negatives: company A

Testing data:

positives: 24 unseen mw categories  
negatives: company B

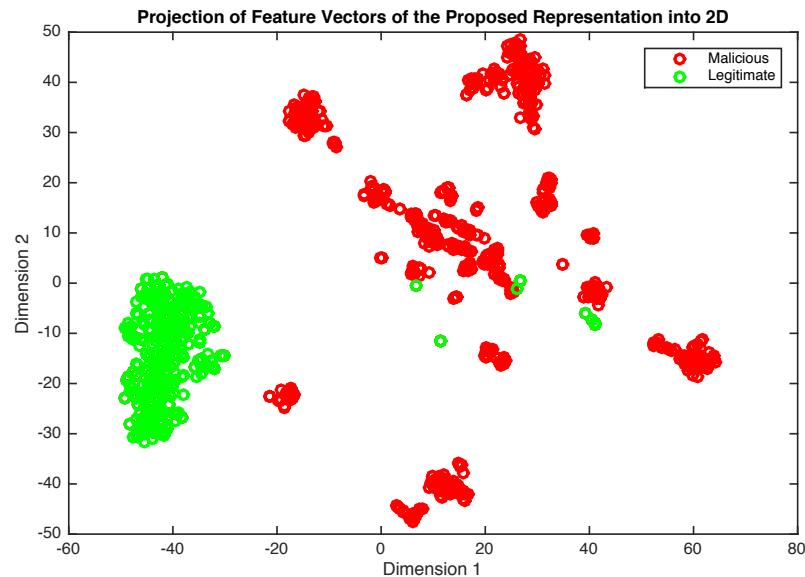
Category	Samples	
	Flows	Bags
Training Positives	132,756	5,011
Click-fraud mw	12,091	819
DGA malware	8,629	397
Dridex	8,402	264
IntallCore	17,317	1,332
Monetization	3,107	135
Mudrop	37,142	701
Poweliks	11,648	132
Zeus	34,420	1,275
Testing Positives	43,380	2,090
Training Negatives	862,478	26,825
Testing Negatives	15,379,466	240,549

# Experiments – 2D projection (t-SNE)



Flow-based features

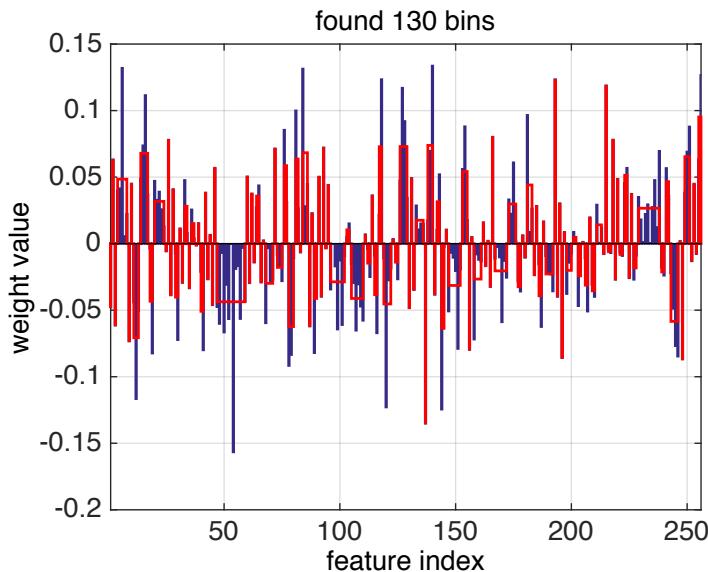
Good for individual malware families



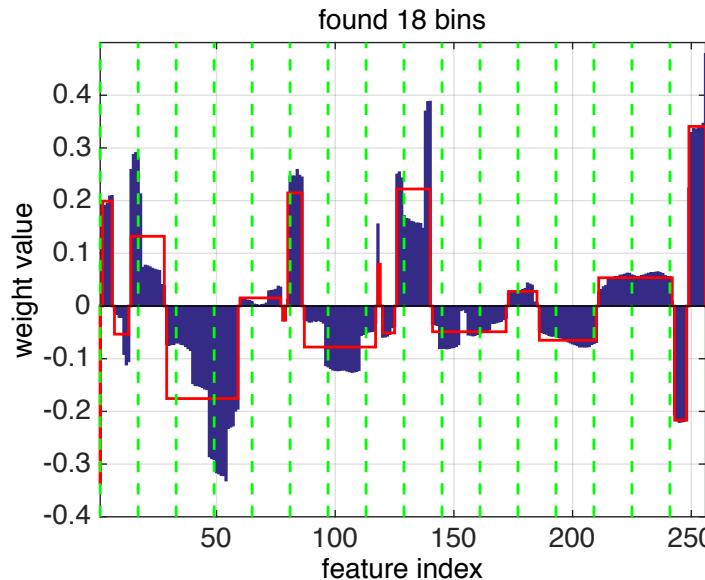
Bag Invariant Features

Good for general malware

# Learning bins from 256 equidistant mini-bins



Standard SVM

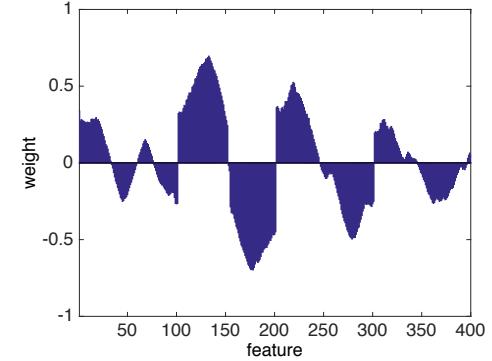
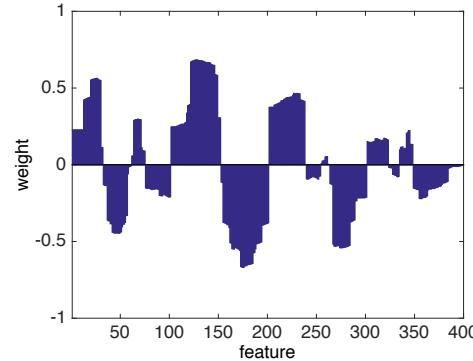
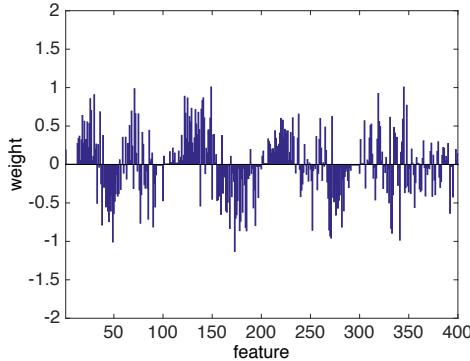
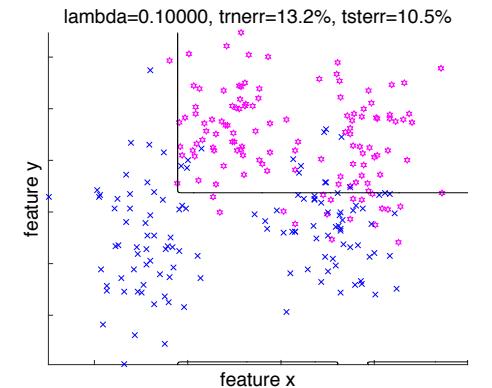
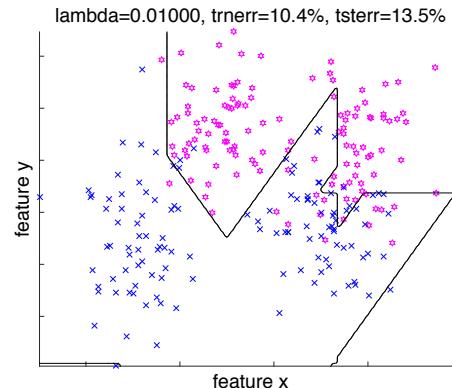
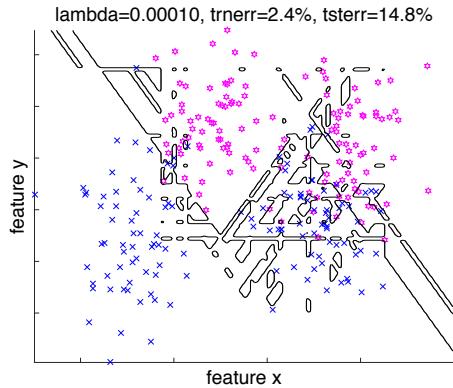


Modified SVM with merging

Blue bars... weights  
Red lines... bins

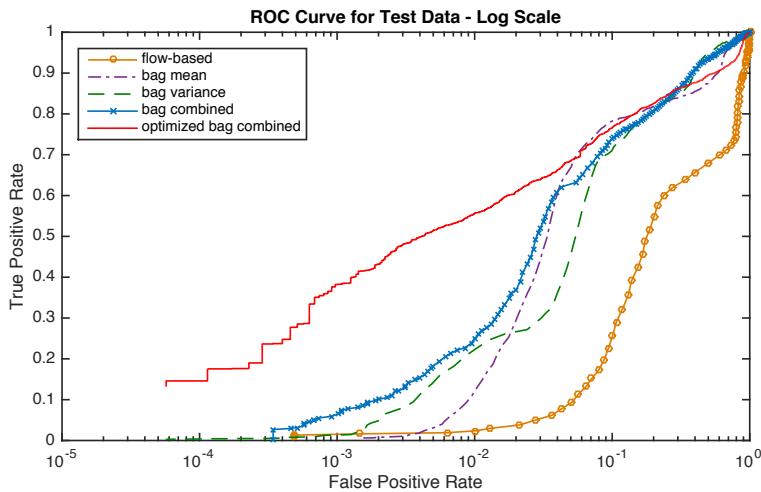
All mini-bins with  
the same weight  
sign create new  
bin.

# Optimizing Decision Boundary

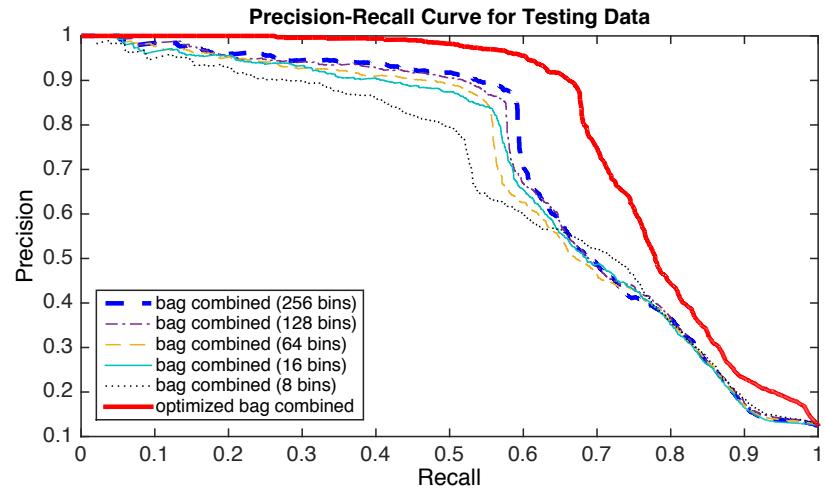


# Efficacy Results – Unseen Malware

ROC Curve – log scale



Precision – Recall Curve



90% precision, 67% recall

# Conclusion and Future Work

- Flaws of flow-based representation
- New representation based on the dynamics of malware bags
- New optimization method that learns the parameters of the representation automatically from the data
- In progress:
  - Modified version for HTTPS

Thank you

Q&A ?