

On the Use of Data Mining Techniques for the Clustering of **URLs Extracted from Network**based Malware Traces



A ML-based 2-stage URL Clustering Framework

Partner

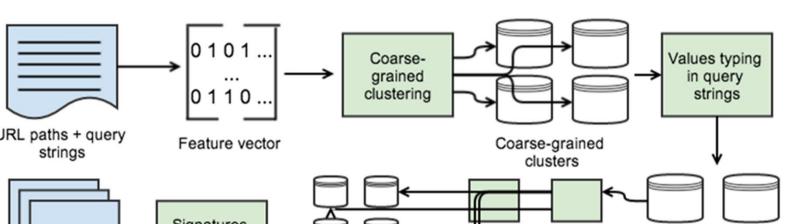
Auteurs

Orange Labs

Overview

keys/attributes http://ieeexplore.ieee.org/stamp/stamp.jsp?tp=2&arnumber=1111&tag=11 scheme://domain:port/path?query_string#fragment_id

■ **Goal:** identify families of malware by grouping issued URLs



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Dataset

Collection

Provided by our partner, URLs are extracted from pcap files generated by executing malware in a sandbox. Legitimate communications are filtered out to reduce the noise. The original malware samples were collected by a third party

• Assumption: common patterns hint at variants or code reuse

Dataset collection: URLs extracted from network capture of sandboxed malware communications with possible peers and C&C Proposed architecture with two steps clustering and typing after the first clustering algorithm

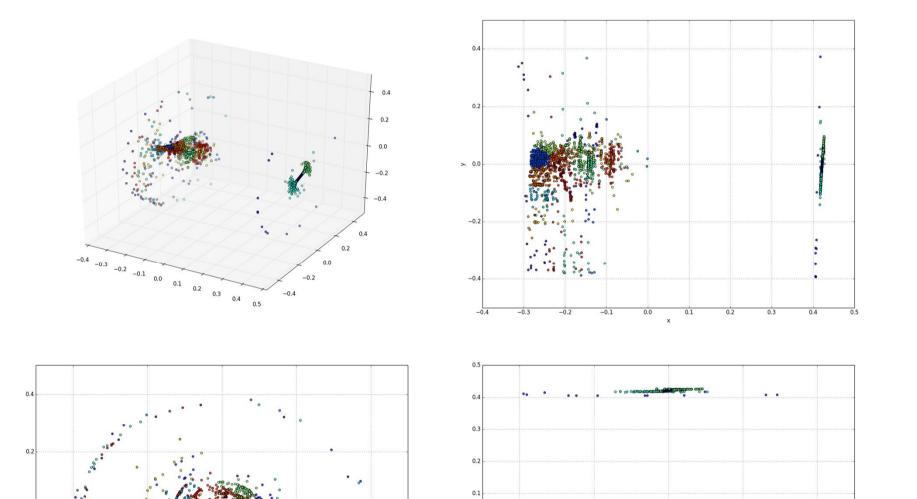
Contributions

- Signatures generation Signatures < Quality evaluation 2D and 3D visualization
- New framework with a typing step and a DBSCAN step to create clusters from a dataset of 1.2M URLs
- Features and distances based only on paths and query strings, not on domains or HTTP headers
- Centralized web platform to monitor and execute machine learning experiments
- Development of generic tools to visualize and navigate through huge numbers of points in 2D and 3D

• Similarity matrix of a coarse-

clustering

grained cluster before DBSCAN



Coarse-grained Clustering

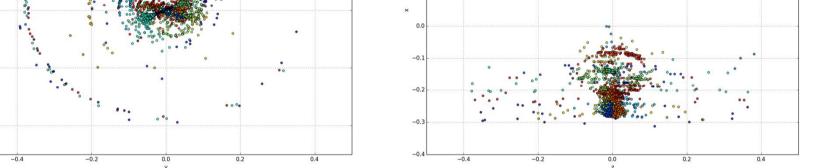
- **Goal:** reduce performance overhead of fine-grained clustering by providing smaller input
- Method: k-means (with k=30) based on ASCII character frequency
- Advantages: unsupervised learning to automate malware discovery, low complexity and ability to specify number of clusters

Pre-processing

The original dataset contains more than 3.5M samples. Removing duplicates leaves a little less than 2M samples. Domain names are then removed to thwart obfuscation, and only GET requests are retained for privacy concerns, leaving around 1.2M samples

Typing

Intermediate step between the two



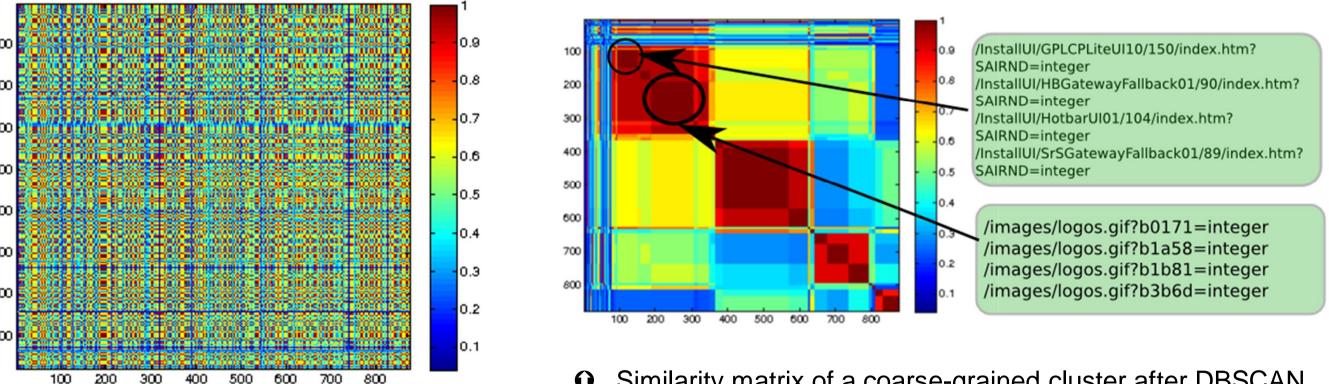
Fine-grained Clustering

- Method: Unsupervised density-based clustering with DBSCAN
- *Distance function:* 1) path distance using the longest common substring algorithm; 2) key/value pair distance based on Jaccard distance on sets of keys associated with a value type

Quality

- Visualizing the density of a similarity matrix gives on the quality of a clustering algorithm
- Dunn index was also used to assess the quality of coarse-grained clusters

C 3D visualization of coarse-grained clusters using projection. One color is associated with each fine-grained cluster



• Similarity matrix of a coarse-grained cluster after DBSCAN clustering

Visualization

- To confirm that a density-based clustering algorithm fits well with the dataset by visualizing the shapes of clusters
- Using multidimensional scaling on the cluster distance matrix, it is possible to compute the main contributing axes on which will be

clustering stages, typing allows to replace values in query string by types. Such abstraction offers better performance during fine-grained clustering. More than 70% of values match with one of the 13 types we defined

Future Work

- URL signature generation for a family of malware
- Signature-matching-based incremental DBSCAN
- Improve first stage through early typing or Canopy clustering
- Apply typing to paths and possibly keys, try refined typing using length of values

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based the 2D and 3D visualizations

• Scre	enshot of the web interface to launch	n experiments and analyze results
Malwurl In Progress	New experiment Datasets Experiments Settings About	
Overview File Browser	Experiment #6	Download data Semove Live
Console	 Dataset: 100k-2 Started: Thu Jan 09 2014 01:42:43 GMT+0100 (CET) 	 Coarse-Grained Clustering kmeans
Fine-Grained Clustering -	 Finished: Thu Jan 09 2014 01:42:53 GMT+0100 (CET) Description: 100k-2, distance v1 	 k = 30 distance: lettersFrequency Fine-Grained Clustering
2D Visualization 3D Visualization Quality		 DBScan eps = 0.5 k = 3 distance: v1-KeyMat
		 Quality Dunn Index: true Visualization Enable: true
		Dimensions Reduction Algorithm: MDSSignatures: false

Site web www.necoma-project.eu

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