



Security Holdings

The Rule for Wild Mal-Gopher Families.

NTT Security Japan

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 2. Creating YARA Module
 3. Clustering Evaluation
 4. Applying to "Wild" Binaries
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1. Introduction

The Rule for Wild Mal-Gopher Families.

Kazuya Nomura

SOC analyst at NTT Security Japan. His main work is alert monitoring with IPS/IDS/EDR. He has contributed articles on malware analysis and data visualization in NTT Security Japan. He is a recipient of the MWS2020 paper award and an outstanding alumnus of SecHack2020.

Sachito Hirao

SOC analyst at NTT Security Japan. Formerly an infrastructure engineer in the financial sector.

At SOC, he was in charge of NW/EDR alert monitoring as well as malware analysis.

- **Golang malware family grows year after year**

- Complexity of analysis due to characteristic structure
- Buildable for multiple platforms



Advantages for the attacker



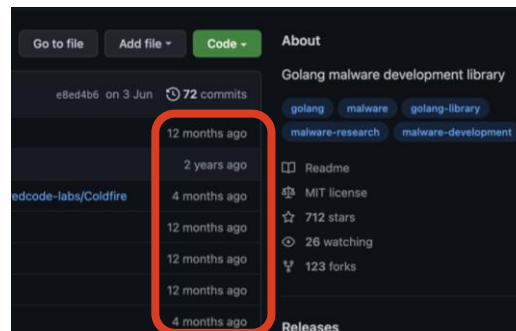
- The number of diverse samples will continue to increase

- Increased efficiency of classification
- Increased efficiency of analysis by attributing to previously analyzed samples

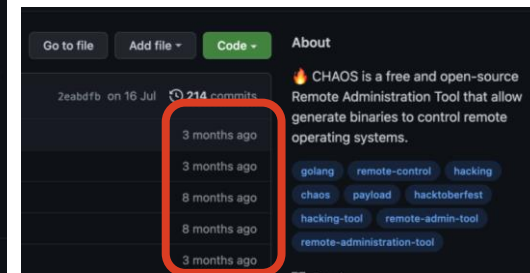
Improved efficiency of Golang malware classification and analysis

Golang Malware

- Various malware creation frameworks exist for Golang
 - Coldfire
 - CHAOS
 - EGESPLOIT
 - ARCANUS
- Many frameworks in active development



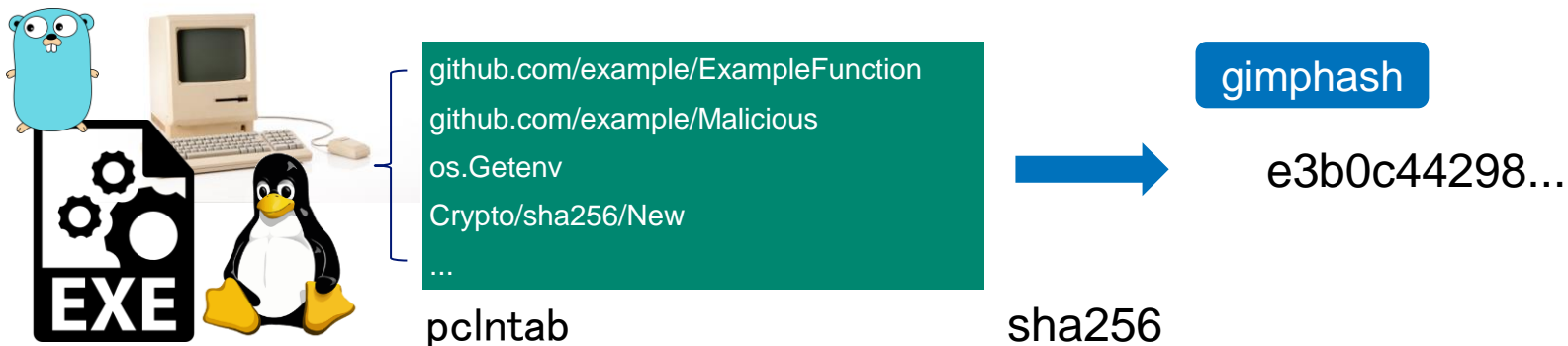
<https://github.com/redcode-labs/Coldfire>



<https://github.com/tiagorlampert/CHAOS>

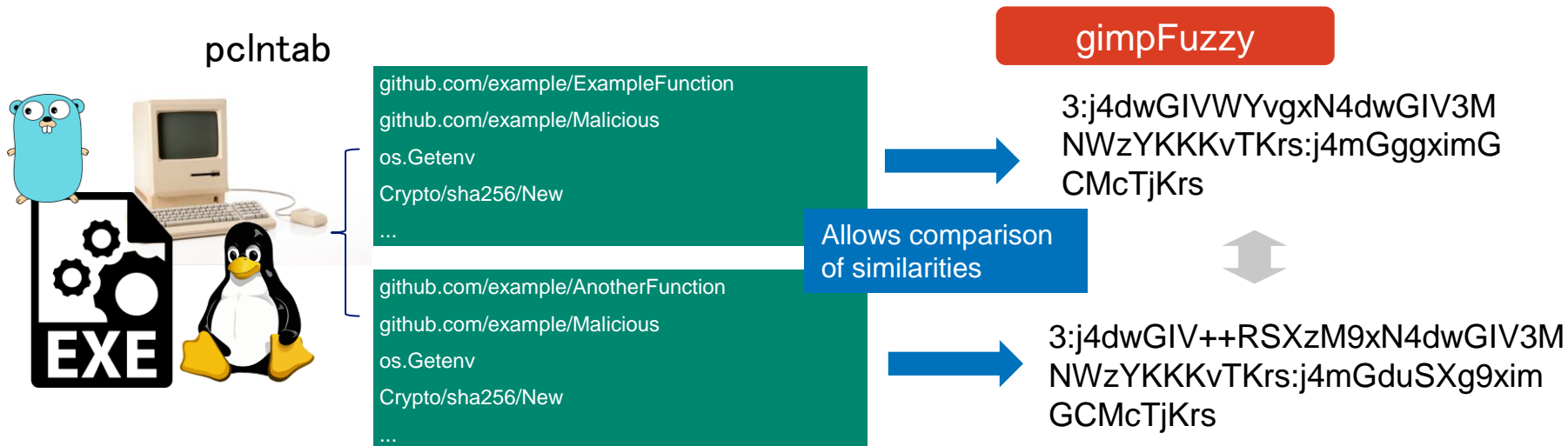
- **Golang binary version's imphash**

- Golang binaries have a platform-independent structure called pcIntab
- Dependent package names, function names, etc. can be restored
- gimphash is a partial SHA256 hash of the recovered package/function name
- Uniquely capable of representing the functionality on which malware depends, but **similarity comparisons of different hashes are not possible**



- **gimphash to fuzzy hash**

- SHA256 output varies greatly if input differs by even 1 bit
- Fuzzy Hash computes a “rough” hash that returns similar values for similar inputs
- gimpfuzzy uses ssdeep. Similarity between samples can be measured.



1. YARA module implementation

- Enables fast and easy classification of large sample groups

2. Accuracy evaluation using analyzed samples

- Consider optimal parameters for family classification

3. Applied to samples submitted to VirusTotal

- Application to unanalyzed “wild” samples
- Discussion of the latest Golang malware applications

- **CODE BLUE 2022**

- **“Who is the Mal-Gopher? – Implementation and Evaluation of “gimpfuzzy” for Go Malware Classification”**

- First to propose a method for applying Fuzzy Hash to gimphash
 - Analyzed samples are classified by gimphash and evaluated the accuracy.

- **JSAC2023 “The Rule for Wild Mal-Gopher Families.”**

- **Implementation and evaluation with a focus on application in actual operations and analysis**

- Creation of a YARA module that allows classification of samples for implementation.
 - Application and evaluation of “wild” unanalyzed samples submitted to VT

Creating YARA Module

The Rule for Wild Mal-Gopher Families.

- Toolkit for malware classification being developed by VT^[1]
 - By writing classification rules, only samples that satisfy the rules can be searched
 - High speed because it is implemented with C
- Various modules exist depending on the file structure
- **The following modules do not exist**
 - Module for handling Golang binaries
 - Module for Fuzzy Hashing and string similarity calculation
- To make it easier to classify samples by gimpfuzzy,
YARA module was newly implemented



[1] <https://github.com/VirusTotal/yara>

Implement the following two

- **go module : analyzes PE binaries made by go lang**
 - `go.gimpfuzzy()` : gimpfuzzy calculate from extracted function name
 - `go.function_names` : sort strings of extracted function names
- **fuzzy module : calculate similarity of Fuzzy Hash**
 - `fuzzy.fuzzy()` : fuzzy hash calculation from argument string
 - `fuzzy.score()` : computes score based on the edit distance between two argument strings

- Example of YARA rules
 - Enables searches based on similarity of samples based on gimpfuzzy

```
import "go"
import "fuzzy"
import "pe"

rule GoFuzzyTest
{
  CONDITIONS:
    pe.is_pe and
    fuzzy.score(go.gimpfuzzy(), "96:05iaa8UdGAq27F92...") > 80
}
```

Similarity
Score
Calculation

gimpfuzzy
calculation
results for
samples

Compare
Fuzzy Hash

The similarity is over 80.
Search for samples

- Enables sample search based on gimpfuzzy similarity

```
root@1f06b9d1f716:/malwares# yara /test.yara -r .
GoFuzzyTest . /Valhalla_hkctl_htran_golang/4550635143c9997d5499d1d4a4c860126ee9299311fed0f85df9bb304dca81ff
GoFuzzyTest . /Valhalla_hkctl_htran_golang/645622a85906da6304315ae9444046f2310609da933f53e87b54fbb206b53e3e
GoFuzzyTest . /Valhalla_hkctl_htran_golang/4e5468e36dc7bc5601384f22c032f990f2e8454d27f6b11e8e897fb0c6c5e0e5
GoFuzzyTest . /Valhalla_hkctl_htran_golang/65cfa86dec6f19cdbf5f9641ab835af023d34fa23b0e31a9f9b66c93a221d7a2
GoFuzzyTest . /Valhalla_hkctl_htran_golang/72549bdc9e857162603f3ce90f1bfc8eb761e7e9f399a24a2bba47468b6edfe3
GoFuzzyTest . /Valhalla_hkctl_htran_golang/91bce99e792db5c3da42da3f01f50a1021f9538b78f70544bedc9ca7508ce54e
GoFuzzyTest . /Valhalla_hkctl_htran_golang/d45a6f12d5956f0fb8ad17727c717b621e3be06fabf9ff27058cb86f8f108b7d
GoFuzzyTest . /Valhalla_hkctl_htran_golang/e70e0c8fb2727b35b65596a6e2838abd0b5f7351cdd4031b9971b91c22f5d15c
```

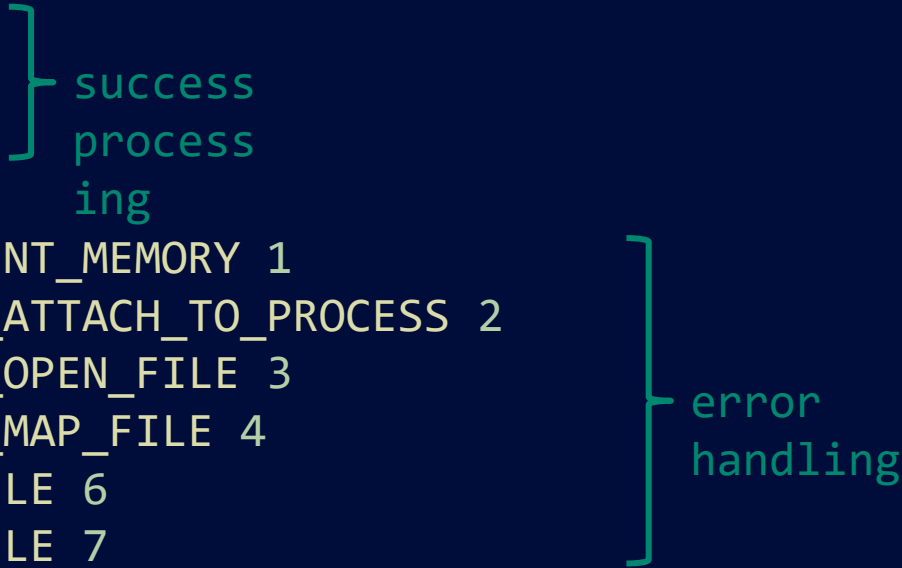
- Implement the following functions

Main imple ment ation	function (e.g. math, programming, programing)	
	module_initialize	Initialization process for YARA module
	module_finalize	YARA module termination process
	module_load	Processing when the module reads a file Implement the actual parsing logic for the file
	module_unload	Post-processing when the module reads a file Delete hash tables, open structures, etc.

- It is important that each function returns a corresponding error when an exception occurs.

- ```
#ifndef ERROR_SUCCESS
#define ERROR_SUCCESS 0
#endif

#define ERROR_INSUFFICIENT_MEMORY 1
#define ERROR_COULD_NOT_ATTACH_TO_PROCESS 2
#define ERROR_COULD_NOT_OPEN_FILE 3
#define ERROR_COULD_NOT_MAP_FILE 4
#define ERROR_INVALID_FILE 6
#define ERROR_CORRUPT_FILE 7
```



# **Demonstration : YARA Module**

## **Demonstration**

# **Clustering Evaluation**

## **The Rule for Wild Mal-Gopher Families.**

- Clustering evaluation using actual observed “wild” samples

## Evaluation using analyzed samples

- Classify **samples identified as malware only**
- Evaluate the validity and accuracy of clustering

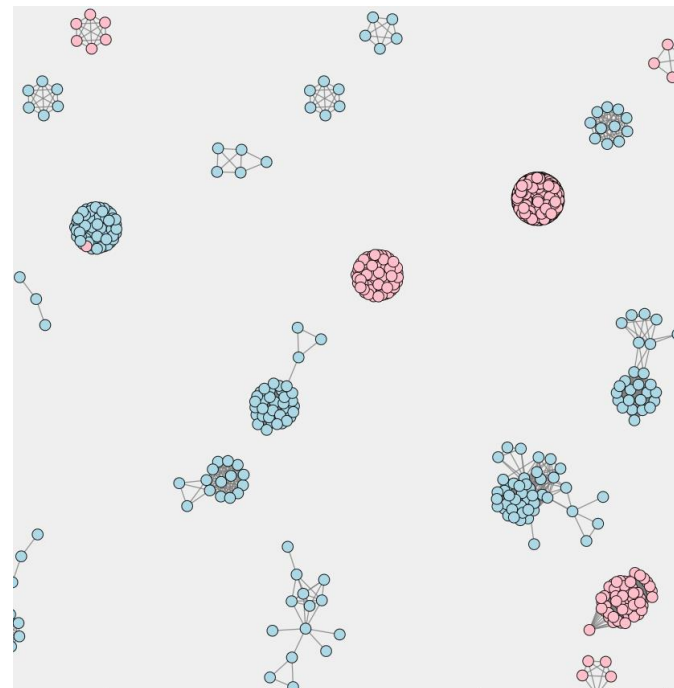
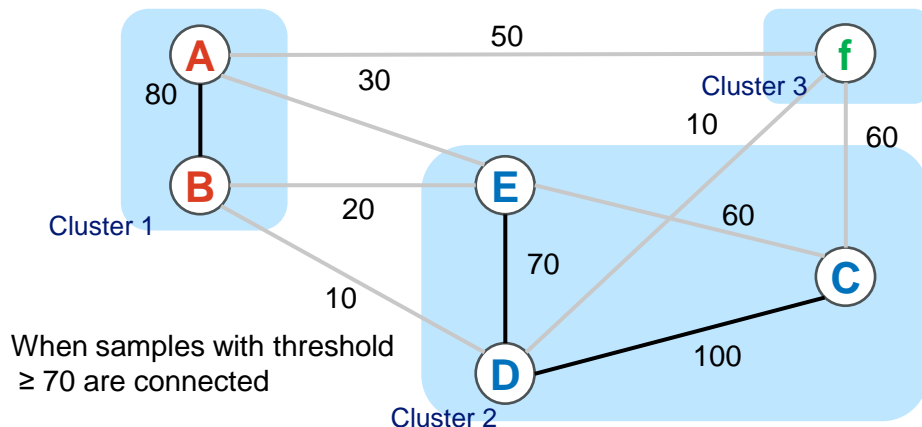


## Evaluation using unanalyzed, up-to-date samples

- Includes **samples not identified as malware**
- Evaluate use in actual operations

# Overview of Clustering Methods

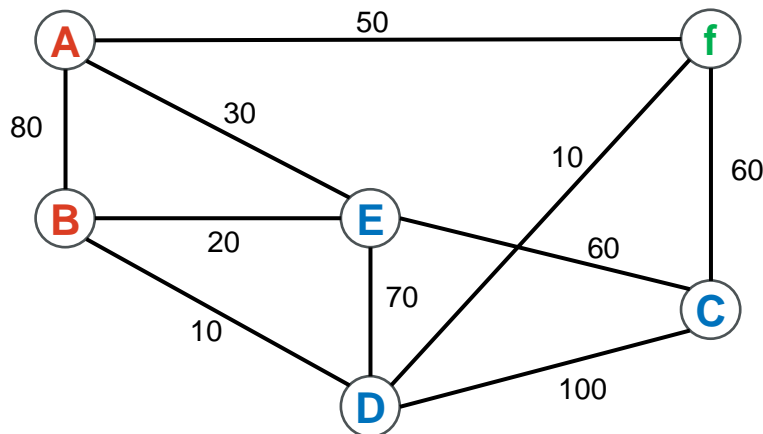
- gimpfuzzy similarity-based clustering
  - Calculate gimpfuzzy of samples for clustering
  - Calculate gimpfuzzy similarity between samples for clustering
  - Edge-connect samples with similarity above a threshold
  - Connected samples are considered as a cluster.



- Calculate scores for all combinations of samples
  - Scoring of string similarity from 0~100 based on edit distance
  - Create an “adjacency matrix” and consider an undirected graph

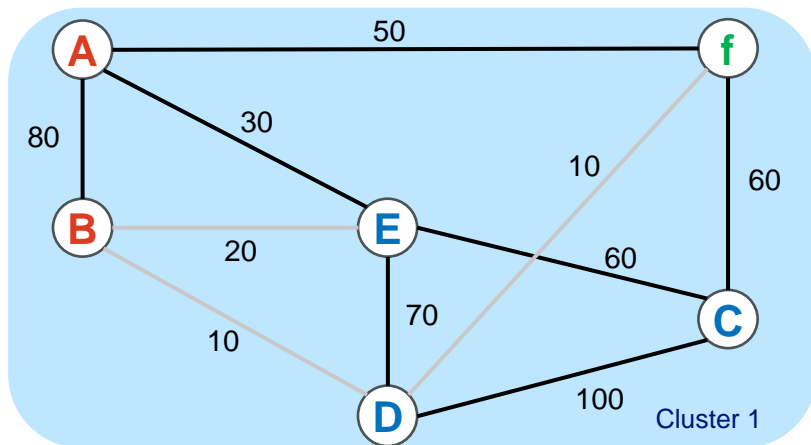
Color corresponds to family

|   | A  | B  | C   | D   | E  | f  |
|---|----|----|-----|-----|----|----|
| A | 0  | 80 | 0   | 0   | 30 | 50 |
| B | 80 | 0  | 0   | 10  | 20 | 0  |
| C | 0  | 0  | 0   | 100 | 60 | 60 |
| D | 0  | 10 | 100 | 0   | 70 | 10 |
| E | 30 | 20 | 60  | 70  | 0  | 0  |
| f | 50 | 00 | 60  | 10  | 0  | 0  |



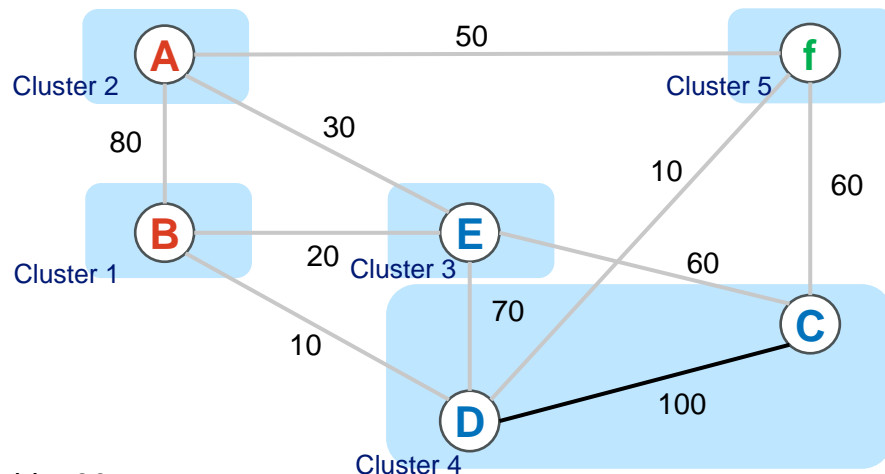
# How to Evaluate Clustering

- How to cut the threshold: what constitutes good clustering?
  - Low threshold case: a small number of large clusters are formed
  - High threshold case: a large number of small clusters are formed



Threshold  $\geq 30$

◎ Many samples can be tied together (**integrity** ↑)  
△ Decrease in classification accuracy within a  $\Delta$  cluster (**homogeneity** ↓)



Threshold  $\geq 90$

◎ High classification accuracy within clusters (**homogeneity** ↑)  
△ Clusters are too separated to be meaningful (**integrity** ↓)

- **Harmonic mean** is considered in trade-off between the two
- Using **V-measure**<sup>[2]</sup> implemented in scikit-learn for evaluation
  - Homogeneity Score  $h$ : The higher the percentage of a single correct answer group in a given cluster, the score is better
  - Integrity score  $c$ : The fewer cluster into which a given group of correct answers is classified, the score is better

$$V_{\beta} = (1 + \beta) \cdot \frac{h \cdot c}{h + c}$$

$$h = \begin{cases} 1 & \text{if } H(C, K) = 0 \\ 1 - \frac{H(C|K)}{H(C)} & \text{else} \end{cases} \quad (1)$$

where

$$H(C|K) = - \sum_{k=1}^{|K|} \sum_{c=1}^{|C|} \frac{a_{ck}}{N} \log \frac{a_{ck}}{\sum_{c=1}^{|C|} a_{ck}}$$
$$H(C) = - \sum_{c=1}^{|C|} \frac{\sum_{k=1}^{|K|} a_{ck}}{n} \log \frac{\sum_{k=1}^{|K|} a_{ck}}{n}$$

$$c = \begin{cases} 1 & \text{if } H(K, C) = 0 \\ 1 - \frac{H(K|C)}{H(K)} & \text{else} \end{cases} \quad (2)$$

where

$$H(K|C) = - \sum_{c=1}^{|C|} \sum_{k=1}^{|K|} \frac{a_{ck}}{N} \log \frac{a_{ck}}{\sum_{k=1}^{|K|} a_{ck}}$$
$$H(K) = - \sum_{k=1}^{|K|} \frac{\sum_{c=1}^{|C|} a_{ck}}{n} \log \frac{\sum_{c=1}^{|C|} a_{ck}}{n}$$

[2] [https://www.researchgate.net/publication/221012656\\_V-Measure\\_A\\_Conditional\\_Entropy-Based\\_External\\_Cluster\\_Evaluation\\_Measure](https://www.researchgate.net/publication/221012656_V-Measure_A_Conditional_Entropy-Based_External_Cluster_Evaluation_Measure)



- **What is the correct classification in the first place?**
  - Minor variants and version differences in malware families
  - malware family
  - Rough malware features
- **Evaluation by paloalto dataset [3]**
  - Analyzed samples with families classified by YARA
  - Exclude samples that did not have a family name
  - Evaluate the classified results and family name with V-measure

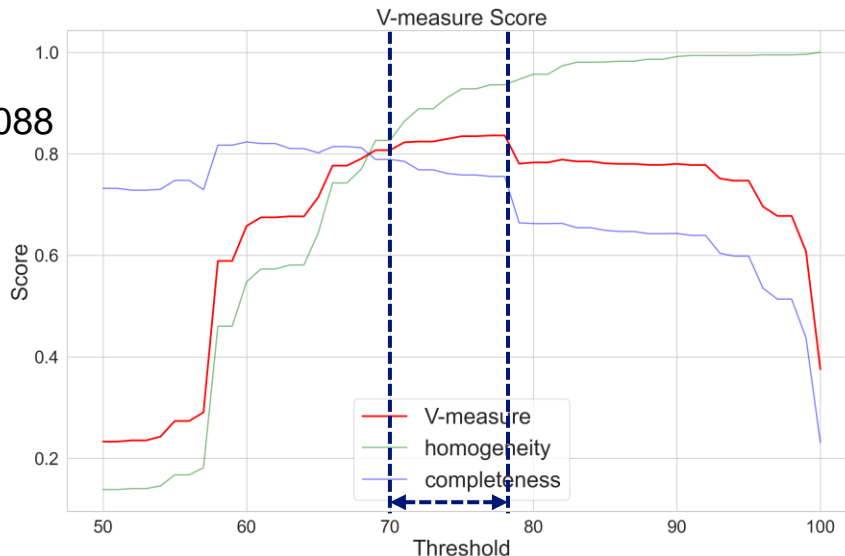
```
1 SHA256,YARA Hit
2 74fc63cfc60f3f9dd3c0d43f59052ba189fa0544ccf79a8f9a90ffdb6b0f0e,trojan_golang_hercules
3 99b89e9580af7fc70d8f6ac079358e6b716f7fd242a6547cf2ca932c4ad9c3df,trojan_golang_veil
4 c6d4fb8c4924863d61678df3aba57fe8efa19946f4c5ea678444ec3d7ada0152,trojan_golang_veil
5 f5798d675289fa5b96635635c94562b3c8ddb99ae12ad5af7b56cccd7c35062,N/A
6 738439ade9ae9e9e6d2f2aff3e63f4161722b3149bf7d02902715c127340c676,trojan_golang_veil
7 f25c859b8f2db7f9a7b40d9234885a1c0a8e2b36e091dbb88041f04f1c46c760,trojan_golang_chaos
8 e49125ac24e15a30619f07fe1ebc2dbce3c8137aabcb86a88b5f1a57a89d03d5f,trojan_golang_infostealer
9 7a0598927921eb15980ee7d512fc2f20dd697642727eb4a38ba638bf4e7ce902,trojan_golang_goBot2
10 57ca3cb685eef7a1fa0f6bb42946adc3a018f8371d4d57204e98601f08d097d,N/A
11 53ded1467133e8c68c47aba33ea242a1751371031d727e8497a60bb9edb2abd3,trojan_golang_gobrut
12 2ad37fa2946780e99f049b8be7980c6a3483c91ccb3b90506e3fdcc629a69039,N/A
13 437e5762c1a814c5934d5d36f1e4a077b14b63be7ffce86999b5503ba34f1aa0,trojan_golang_veil
14 56b110a95c2b16784ba053c69f3ffcd6ffcfef1fd42214f71d61b9e0d59b9a42,N/A
15 55f4f5be742d8557956af3278f01825fb02cb90fa2f2f0c1f5160322c26a1af,trojan_golang_veil
16 e15c2dad9d8e9c788cb394aa04d5e070a50512c25cecb4c5e1e99d69fb52d7ea,trojan_golang_veil
17 722f1c182bac229812107b6ab87853f886ff5c1f96fbdd343dc1847667fb7f79,trojan_golang_veil
18 d69f4caf27097e9a8d7241aa1334fa790d4f5a5708de12a1b8aabd5239724cd8,trojan_golang_veil
19 c680a89e218c74bde438119f9f3112c8725be59956a5f3c53812165bfe556d2,N/A
```

sample  
SHA256

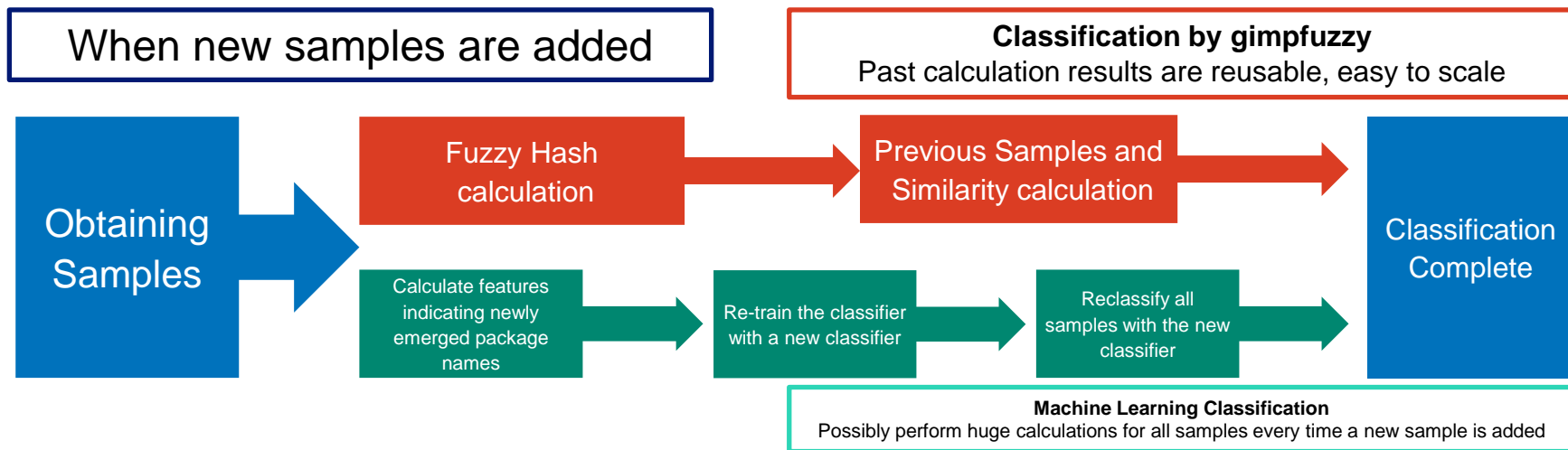
Matches.  
YARA rules

[3][https://github.com/pan-unit42/iocs/blob/master/golang\\_malware\\_results.csv](https://github.com/pan-unit42/iocs/blob/master/golang_malware_results.csv)

- paloalto data set
  - Number of samples: 10,700
  - Number of samples downloadable from VT: 7,088
  - Of which, family name is indicated : 5,808
- Evaluation by V-measure
  - Evaluation of classification assuming that the family is correct
  - Best classification accuracy at a threshold of about 70~80



- Advantages of gimpfuzzy's similarity-based clustering compared to machine learning
  - Low calculational complexity
  - Less susceptible to time variation



# **Applying to “Wild” Binaries**

## **The Rule for Wild Mal-Gopher Families.**

- Clustering evaluation using actual observed "wild" samples

## Evaluation using analyzed samples

- Classify **only samples identified as** malware
- Evaluate the validity and accuracy of clustering



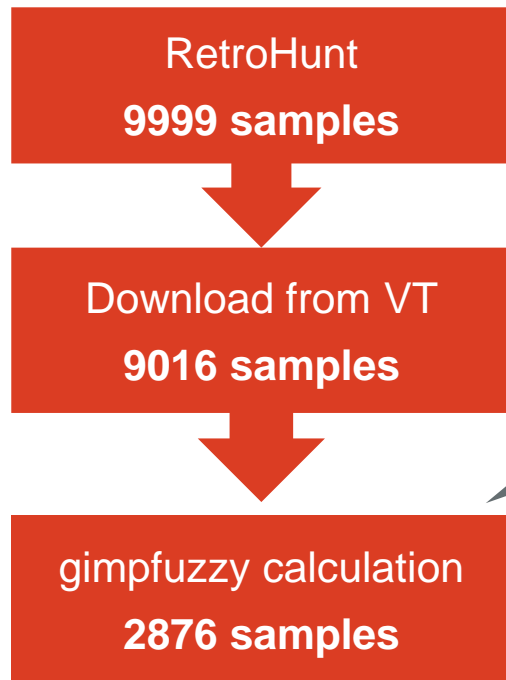
## Evaluation using unanalyzed, up-to-date samples

- Includes **samples not identified as** malware
- Evaluate use in actual operations

- Download the samples that matched the following YARA with VT's Retrohunt
  - Samples that matched Windows binaries made by golang

```
rule go_language_pe
{
 strings:
 $go1 = "go.buildid" ascii wide
 $go2 = "go.buildi%" ascii wide
 $go3 = "Go build ID:" ascii wide
 $go4 = "Go buildinf:"
 $go5 = "runtime.cgo"
 $go6 = "runtime.go"
 $go7 = "GOMAXPRO"
 $str1 = "kernel32.dll" nocase
 CONDITIONS:
 uint16(0) == 0x5A4D and uint32(uint32(0x3C)) == 0x00004550 and 2 of ($go*) and all of ($str*)
}
```

- Sample collection results



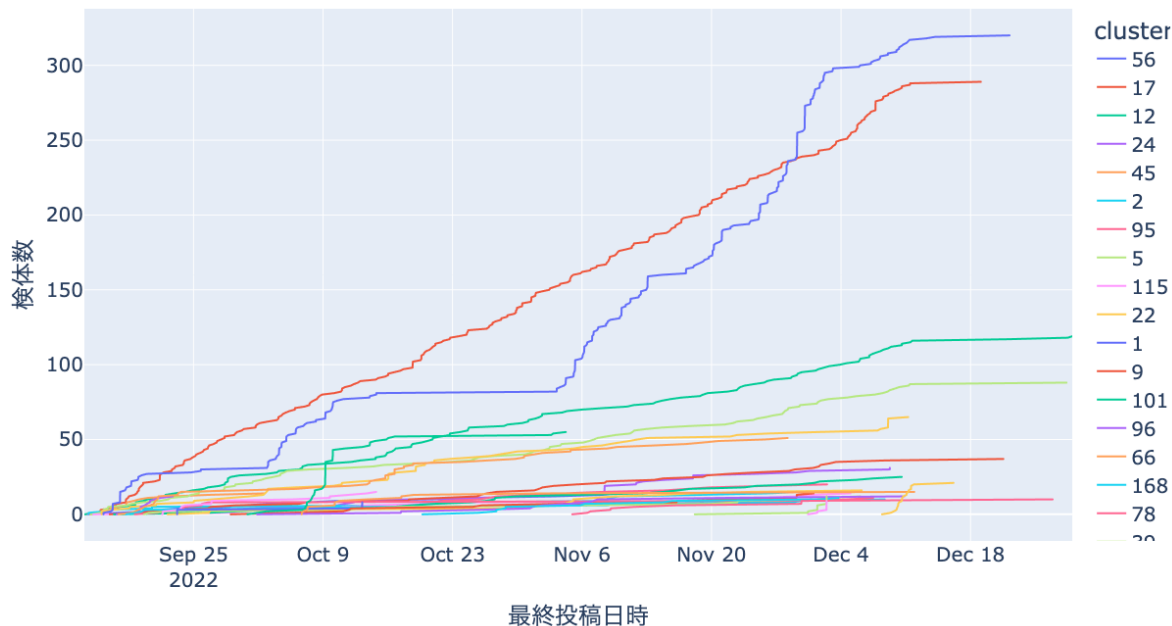
- Samples collected by Retrohunt  
(Approximately the last 3 months from 2022/12)
- Downloaded samples without duplicates  
such as Subfile

In case of insufficient bytes of ssdeep input  
Can be improved by using TLSH, etc.

- Samples for which pcIntab analysis + gimpfuzzy  
calculation was possible
- UPX are unpacked and analyzed

# Clustering Result

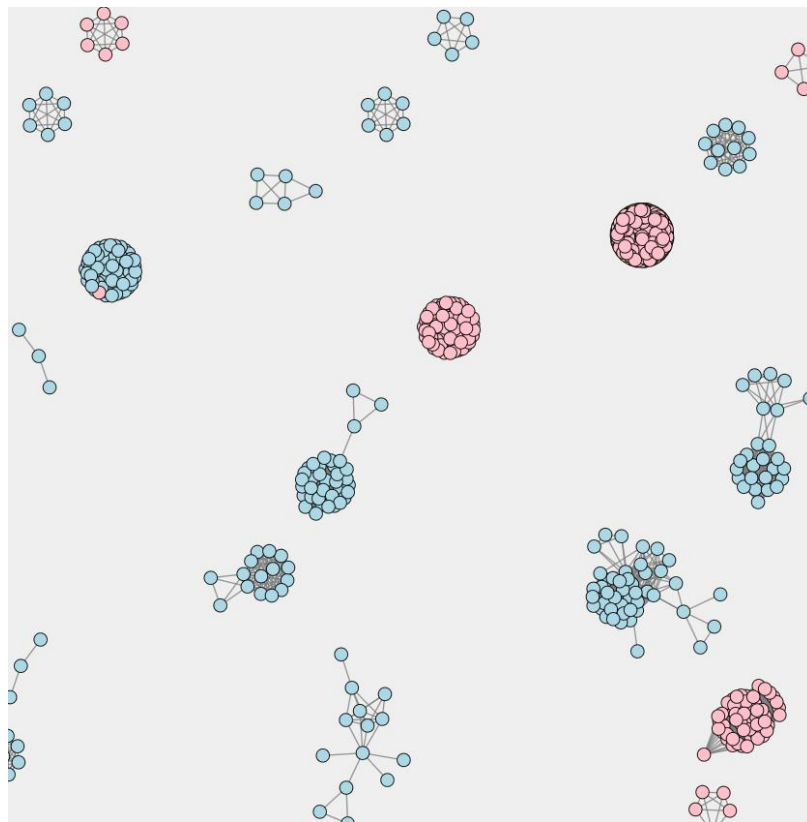
- Created 1093 clusters for 2867 samples





# Clustering Result

- Cluster Visualization
  - Implemented with Python's bokeh.io
  - Create interactive “moveable” graphs
  - Red node have 10 or more malignant determinations by VT



# **Demonstration : Cluster Visualization**

## **Demonstration**

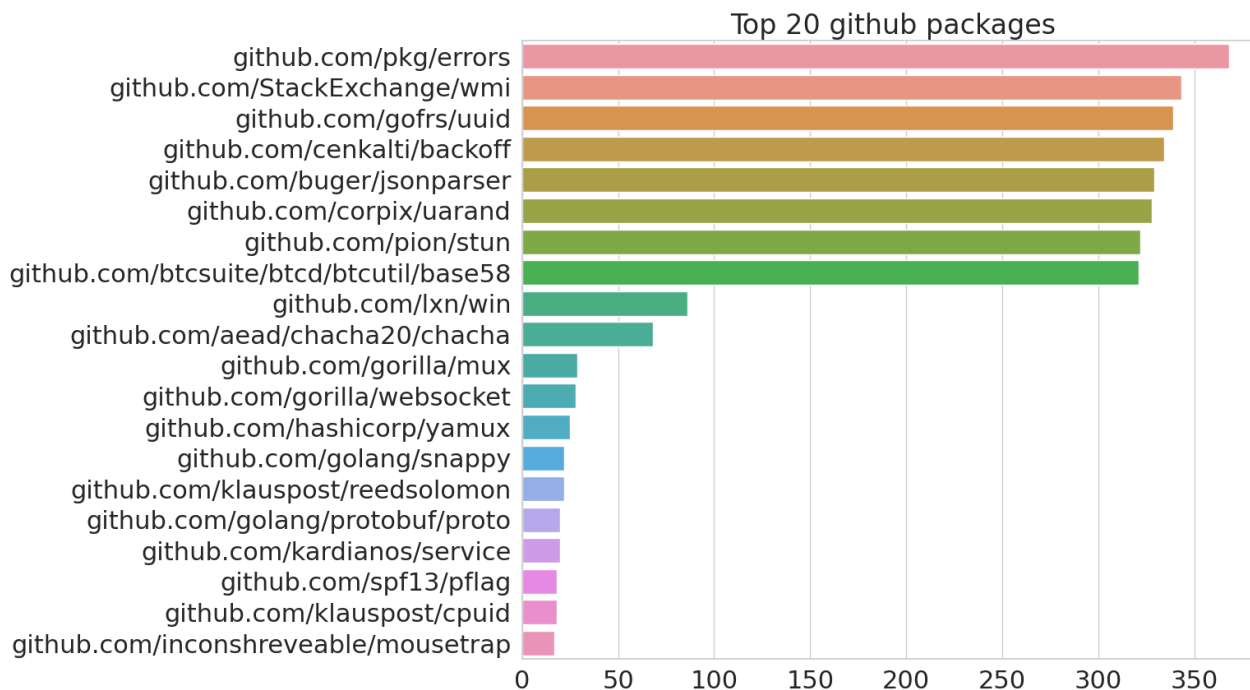
- Golang can specify github repository for packages
- Focused on samples with more than 10 malignant determinations, 705 repository names recovered.
  - Including repositories that are considered private.
- Interesting repository name restored
  - Packages that generate random UAs (corpix/uarand etc.)
  - Packages that conduct Process Invoking (inconshreveable/mousetrap, etc.)
  - Malware creation frameworks (tiagorlampert/CHAOS, etc.)
  - Multi-hop proxies (Dliv3/Venom, etc.)
  - Post-Exploitation framework (Ne0nd0g/merlin, etc.)

# Case Study 1 / GitHub Packages



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## Top 20 github repositories that appeared in malignant samples



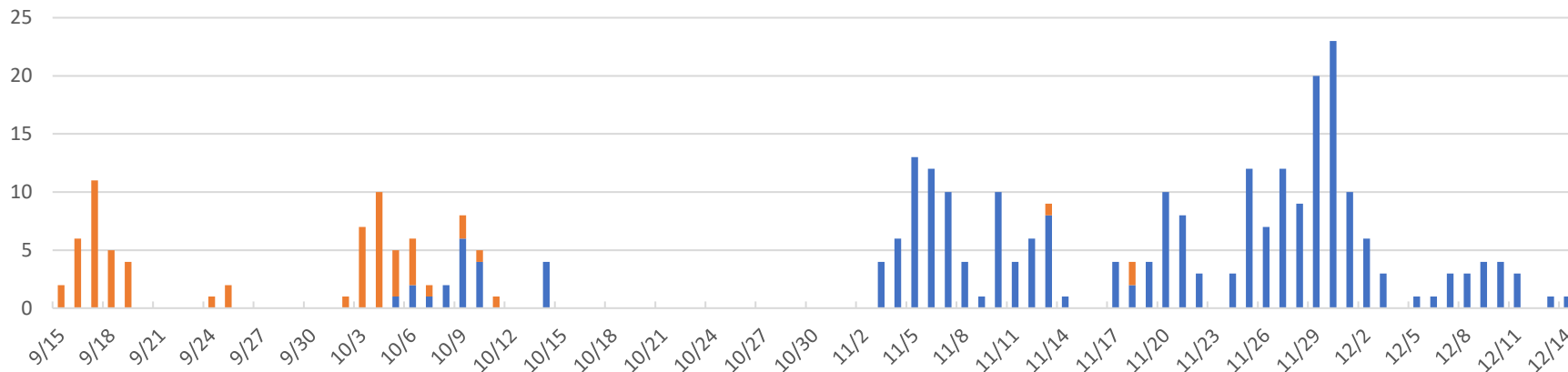
# Case Study 2 : Detecting Additional Malware Features



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- Detect small changes in GimpFuzzy values in clusters of malignant samples and observe temporal changes
  - 768:KZZ99PdnRrLXT3UbhHPBj/RqJgvm+HHHyScP0OhZIXCPINvxtWrX7G/VAmWeEX
  - 768:KZZ99PdnRrLXT3UbhHPBj/RqJgvm+HHHyScP0OhZIXCPINvxtWrX7G/V3mWeEX



# Case Study 2: Detecting malware functionality additions



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- Changed functions also changed logic.
  - Function name that was changed
    - 768:~VAmWeEX  
→ application/pesignaturetest/wincert.GetPostalCode
    - 768:~V3mWeEX  
→ application/pesignaturetest/wincert.Extract



# Case Study 2: Detecting malware functionality additions

- Some functions with unchanged function names have logic changes

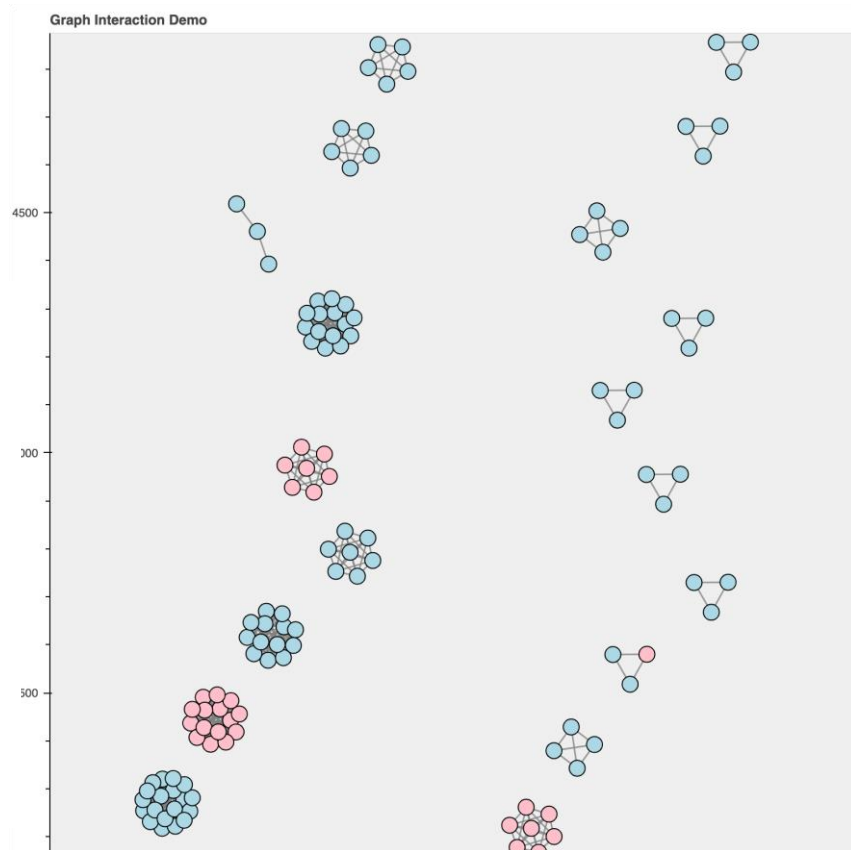
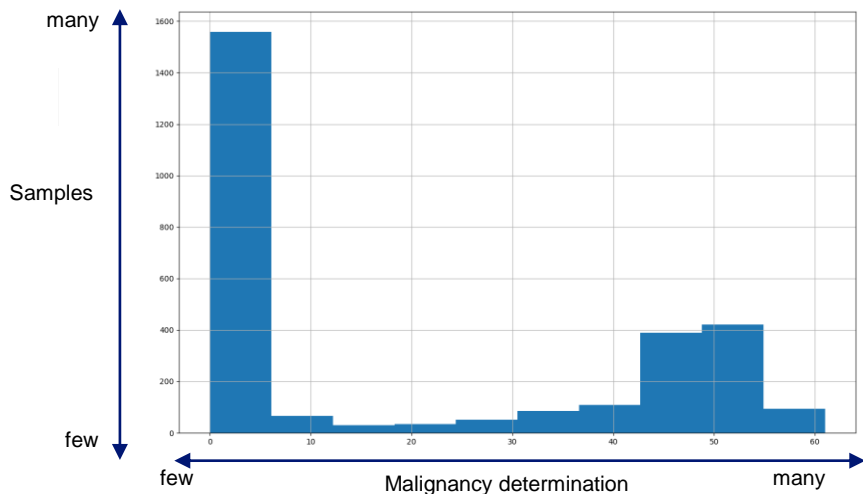
|                                                                                   | Similarity % | Confidence | Address  | Primary Name               | Type   | Address  | Secondary Name             | Type   | Basic Blocks |    |    | Jumps |    |    |
|-----------------------------------------------------------------------------------|--------------|------------|----------|----------------------------|--------|----------|----------------------------|--------|--------------|----|----|-------|----|----|
|  | 0.26         | 0.98       | 006A4080 | main_reportInstallFailure  | No...  | 006A3DC0 | main_reportInstallFailure  | No...  | 0            | 10 | 58 | 4     | 9  | 84 |
|  | 0.45         | 0.97       | 0069DA10 | main_getCampaignID         | No...  | 0069D790 | main_getCampaignID         | No...  | 11           | 3  | 2  | 15    | 3  | 2  |
|  | 0.81         | 0.96       | 0067B610 | application_pesignature... | Normal | 0067B610 | application_pesignature... | Normal | 7            | 39 | 0  | 22    | 32 | 13 |
|  | 0.86         | 0.97       | 006B01B0 | main_extractDistributor... | Normal | 006B05E0 | main_extractDistributor... | Normal | 0            | 10 | 2  | 1     | 11 | 3  |
|  | 0.91         | 0.99       | 006943D0 | main_initializeConfig      | Normal | 006941D0 | main_initializeConfig      | Normal | 0            | 21 | 3  | 2     | 22 | 6  |

- main\_reportInstallFailure is added to the communication functionality.

```
234 v42 = (http_Request *)net_http_NewRequestWithContext(
235 (int)&go_itab_ptr_context_emptyCtx_comma_ptr_context_Context,
236 dword_C76630,
237 (int)"POSTQEMU",
238 4,
239 (int)"https://fulusus.com/api/install-failure",
240 39,
241 v39,
242 v40);
243 if (!v43)
244 {
245 Header = (runtime_hmap *)v42->Header;
246 v56 = net_textproto_CanonicalMIMEHeaderKey((int)"Content-Type", 12);
247 v55 = (_DWORD *)runtime_newobject((int)&RTYPE__1_string);
248 v55[1] = 33;
249 *v55 = "application/x-www-form-urlencoded";
250 }
```

# Case Study ③ : Legitimate Files

- In reality, legitimate files dominate.
  - Same for samples submitted to VT.
  - Is clustering of regular files possible?

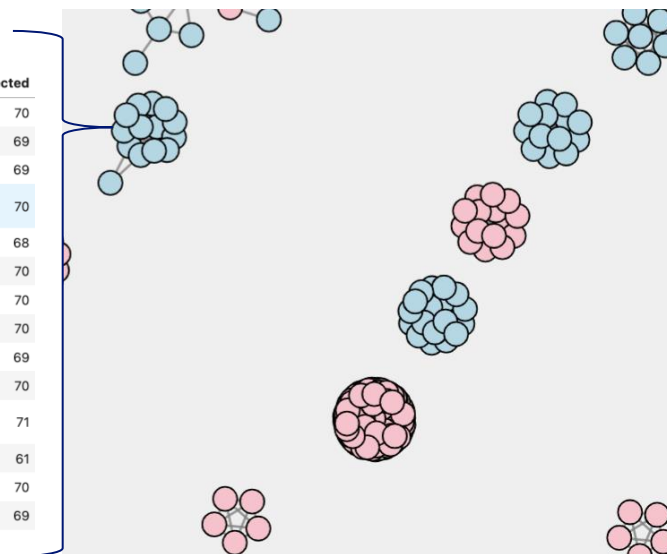




# Case Study ③ : Legitimate Files

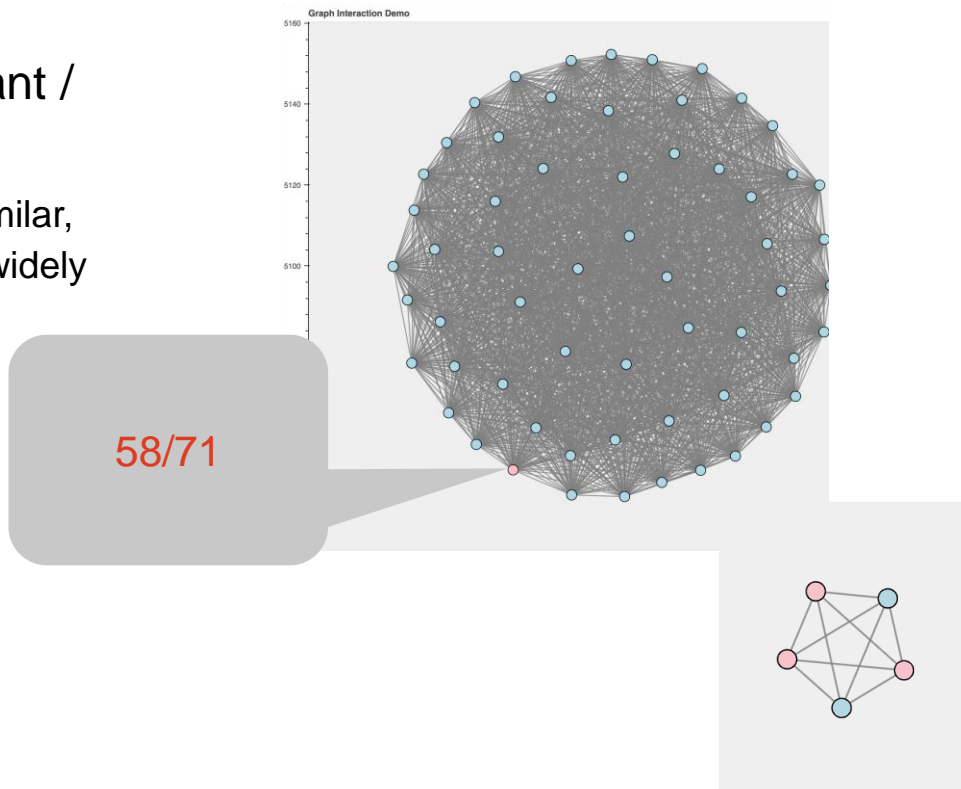
- Even samples that appear to be legitimate files can cluster.
  - samples submitted to VT are not necessarily only malignant files

|                                                   | gimpfuzzy | cluster | last_submission_date | size     | type_description | malicious | undetected |
|---------------------------------------------------|-----------|---------|----------------------|----------|------------------|-----------|------------|
| 3072:ZVZoThQpAM+mBL+5CR61yLR2Mr8tA4ICAluEbXDbK... |           | 106     | 2022-08-23 17:29:28  | 27937584 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+mBL+5CR61yLR2Mr8uA4ICAluybXDbK... |           | 106     | 2022-08-29 10:48:57  | 27937072 | Win32 EXE        | 0         | 69         |
| 3072:ZVZoThQpAM+mBL+5CR61yLR2Mr8uA4ICAluhbXDbK... |           | 106     | 2022-09-02 02:07:33  | 27938096 | Win32 EXE        | 0         | 69         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-05 12:12:53  | 27936048 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-06 05:52:38  | 27913216 | Win32 EXE        | 2         | 68         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-08 11:21:42  | 27940144 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-14 07:23:13  | 27940144 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-16 09:19:15  | 27940136 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-16 13:15:25  | 27940136 | Win32 EXE        | 0         | 69         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-18 12:38:06  | 27936048 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-19 07:44:26  | 27940144 | Win32 EXE        | 0         | 71         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-19 17:28:46  | 27940128 | Win32 EXE        | 0         | 61         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-21 05:35:09  | 27940128 | Win32 EXE        | 0         | 70         |
| 3072:ZVZoThQpAM+zBL+5CR61yLR2Mr8uA4ICAluwbXRb1... |           | 106     | 2022-09-28 16:48:43  | 27917312 | Win32 EXE        | 1         | 69         |



# Case Study ④ : Floxif

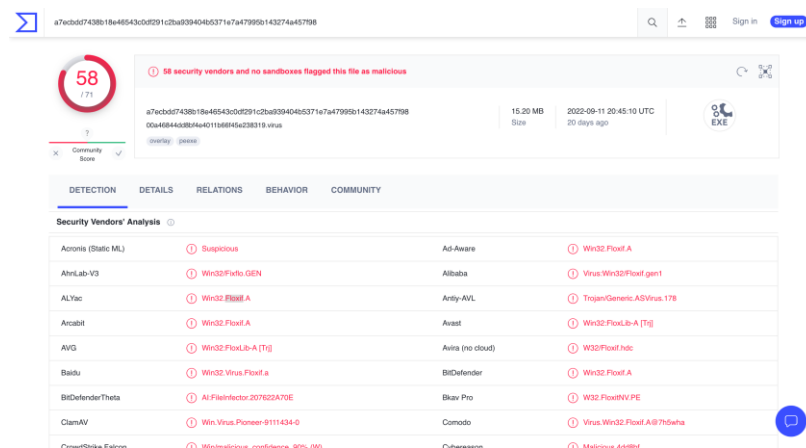
- Mixed clusters of malignant / benign determinations
  - Even though gimpfuzzy is similar, malignancy judgments vary widely within clusters



# Case Study ④ : Floxif

- Highly malicious samples lurking in legitimate file clusters
  - We found a highly malignant Floxif sample that mimicked the following program
    - psiphone-tunnel-core
    - Acronis Cyber Protect
  - It is difficult to determine malignancy/benignity in some samples with unsupervised clustering alone
  - Correct results as sample profiling

|   |                     |          |           |    |    |
|---|---------------------|----------|-----------|----|----|
| 4 | 2022-09-09 20:21:30 | 15857968 | Win32 EXE | 0  | 63 |
| 4 | 2022-09-10 08:11:54 | 15857968 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-10 10:48:23 | 15857968 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-10 13:50:39 | 15936247 | Win32 EXE | 58 | 13 |
| 4 | 2022-09-11 18:54:17 | 15857968 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-12 07:56:33 | 15857968 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-12 14:09:19 | 15857968 | Win32 EXE | 0  | 65 |
| 4 | 2022-09-12 21:44:51 | 16389424 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-13 04:48:15 | 15857968 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-13 19:51:55 | 15857968 | Win32 EXE | 0  | 70 |
| 4 | 2022-09-13 22:13:20 | 15857968 | Win32 EXE | 0  | 70 |



- Existence of samples for which gimpfuzzy cannot be calculated
  - Lower limit of ssdeep input size exists (>4KB). It can be replaced by TLSH, etc.
  - Analysis is interfered by packing, obfuscation, etc.
- Limitations of “unsupervised” classification
  - It is difficult to determine malignant / benign.
  - Separated by clusters to some extent, but some clusters with malignant and benign samples still exist.

# **Conclusions**

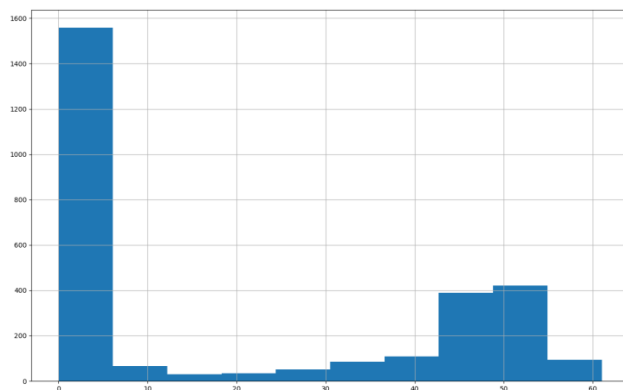
**The Rule for Wild Mal-Gopher Families.**

- Presented on the following topics to apply gimpfuzzy to actual operations
  - YARA module implementation
  - Accuracy evaluation using analyzed samples
  - Application to samples submitted to VirusTotal
- YARA module and visualization scripts are to be released.

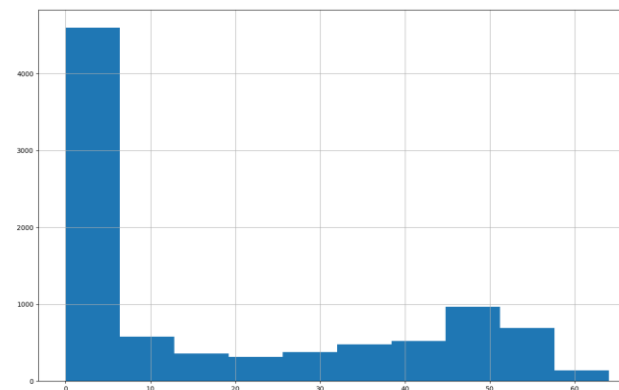
# **Appendix**

## **The Rule for Wild Mal-Gopher Families.**

- Number of malignancy determinations for samples collected in VT
  - Overwhelmingly less malignant files



Analyzed samples  
analyzed : 2835

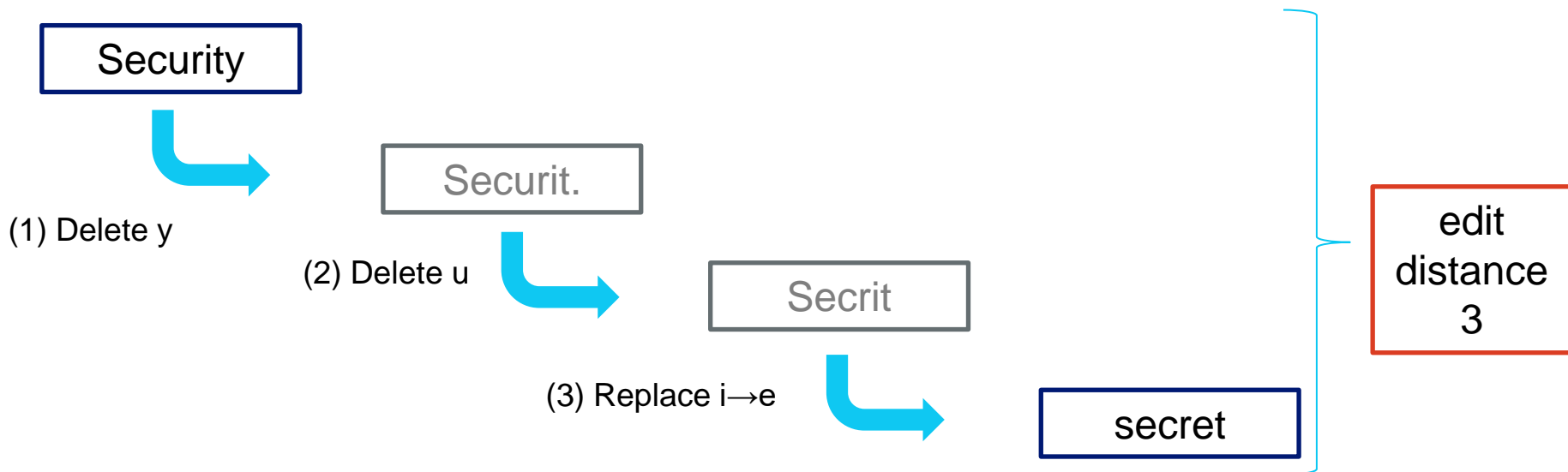


Unanalyzed samples :  
9999 samples



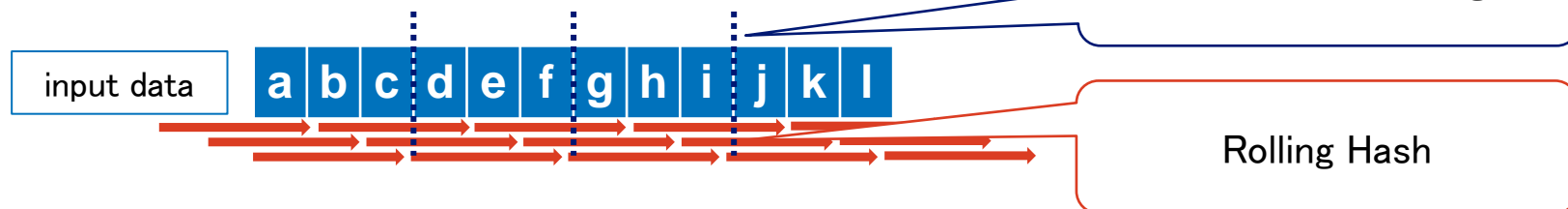
- **Edit Distance (Levenshtein Distance)**

- Classic method of showing string similarity
- Minimum number of times one string can be converted to the other by inserting, deleting, or replacing a single character.



- **ssdeep (Context Triggered Piecewise Hashing)**

- Piecewise Hashing : Hash of divided part of the data
- Rolling Hash : Hash for fixed-length partial data



- When the Rolling Hash reaches a certain value, it is split there and Piecewise Hashing is performed.
- The triggering value is calculated based on the input data length

