

# Malware Classification based on Static Analysis

# Malware Classification based on Static Analysis

- Write the code to implement the malware classification method based on the below paper:

[Nataraj, Lakshmanan, et al. "Malware images: visualization and automatic classification." \*Proceedings of the 8th international symposium on visualization for cyber security\*. 2011.](#)

- The feature extraction methods (GIST,...), and machine learning methods (CNN,...) are not limited. You can try different ones, and make some *comparison*.

[Some Reference]

- [Kumar, Nitish, and Toshnall Meenpal. "Texture-based malware family classification." \*2019 10th International Conference on Computing, Communication and Networking Technologies \(ICCCNT\)\*. IEEE, 2019.](#)
- [Bensaoud, Ahmed, Nawaf Abudawaood, and Jugal Kalita. "Classifying malware images with convolutional neural network models." \*International Journal of Network Security\* 22.6 \(2020\): 1022-1031.](#)
- [Kiger, John, Shen-Shyang Ho, and Vahid Heydari. "Malware Binary Image Classification Using Convolutional Neural Networks." \*International Conference on Cyber Warfare and Security\*. Vol. 17. No. 1. Academic Conferences International Limited, 2022.](#)
- [Vasan, Danish, et al. "Image-Based malware classification using ensemble of CNN architectures \(IMCEC\)." \*Computers & Security\* 92 \(2020\): 101748.](#)

# Hints

1. Malware dataset: Collect samples from [VirusShare](#) or [MOTIF Dataset](#) (with labeled virus hash value)

- **Be Careful. You may need to do it in the Virtual Machine (VMWare, VirtualBox, ...).**
- **Do not use pre-existing transformed datasets, such as the Mallmg dataset. You need to perform the image transformation using your own code.**

Virus Family

A1	A	B	C	D	E	F	G	H	I	J	K
1	108d180ef1366eed83d27a26cdca2741	0e99fa4 0be7de 108d180 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
2	2786c78cf6edc7b85adaf4234e1a4d6e	7efe949 1e9ac4 2786c78 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
3	32a56ca79f17fea432250ee704432dfc	0ba7bc 758b3c 32a56ca 7ev3n				['7ev3n', 'seven',	0 Malwarebytes	5/6/2016		<a href="https://blog.malwarebytes.com/thr">https://blog.malwarebytes.com/thr</a>	
4	483debd567d37a4e78c20e88d2c2c0ee	57af1d7 f6ef4af 483debd 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
5	4a7599b6591fcd643bd435e53b5850b8	41543c f801b2c 4a7599b 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
6	4d6f43e9c2a48e258a2102e9b49d56d6	a39182 75b7a7 4d6f43e 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
7	5acbeb7ddacbf7297fe25ef02f215038	6377f47 7a827fc 5acbeb7 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
8	5b271620663c1a48ac986d412478b5d2	0c0c9a 2a742d 5b27162 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
9	5b7c466ce24ef6359c0006af70d9e4fa	e580dd 022e57 5b7c466 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
10	8434eea972e516a35f4ac59a7f868453	39eff0a 92ac6b 8434eea 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
11	95f18fe1d393e2c671d9afac9590a5a3	ecd47e aa242f 95f18fe 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
12	9f8bc96c96d43ecb69f883388d228754	61ed25 7d373c 9f8bc96c 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
13	a3dfd4a7f7c334cb48c35ca8cd431071	ccc4adc 0132c0 a3dfd4a 7ev3n				['7ev3n', 'seven',	0 Malwarebytes	5/6/2016		<a href="https://blog.malwarebytes.com/thr">https://blog.malwarebytes.com/thr</a>	
14	d3609b3179b164b0af6845226ac05f70	b03a0f5 db7f34f d3609b3 7ev3n				['7ev3n', 'seven',	0 Malwarebytes	5/6/2016		<a href="https://blog.malwarebytes.com/thr">https://blog.malwarebytes.com/thr</a>	
15	f4c66e06eafe74b8343f35a90b194169	0b17d5 17c95a f4c66e0 7ev3n				['7ev3n', 'seven',	0 Proofpoint	2/19/2016		<a href="https://www.proo">https://www.proo</a>	<a href="https://www.proo">https://www.proo</a>
16	e4709fb8bc86334096093f3c6a181caa	00a46a 24e397 24e3975 abaddonpos				['abaddonpos']	1 Proofpoint	5/10/2016		<a href="https://www.proofpoint.com/us/thr">https://www.proofpoint.com/us/thr</a>	
17	ed06bf280c1694d4d41a23d6a5240b2a	15e8c2 631156 631156f abaddonpos				['abaddonpos']	1 Proofpoint	12/8/2016		<a href="https://www.proofpoint.com/us/thr">https://www.proofpoint.com/us/thr</a>	
18	8e2a899d404d33e0789ee0fdbbb340af	aa1b4ff eb30a1 eb30a18 acidbox				['acidbox']	2 Palo Alto Networ	6/17/2020		<a href="https://unit42.paloaltonetworks.co">https://unit42.paloaltonetworks.co</a>	
19	f5ad74379f28147858881888de96a8fc	a7d89b 3ef071e 3ef071e acidbox				['acidbox']	2 Palo Alto Networ	6/17/2020		<a href="https://unit42.paloaltonetworks.co">https://unit42.paloaltonetworks.co</a>	
20	fc73d8b63a78aac1f44b9eb4494e25e	b6631b 003669 0036697 acidbox				['acidbox']	2 Palo Alto Networ	6/17/2020		<a href="https://unit42.paloaltonetworks.co">https://unit42.paloaltonetworks.co</a>	
21	ae829244df44b3735944707157db0885	27efeeb 12cb02 12cb029 adchiate				['adchiate']	3 G DATA	9/29/2017		<a href="https://www.gdat">https://www.gdat</a>	<a href="https://www.gdat">https://www.gdat</a>
22	4d7471711185364b8d9c8a19bc6ff3d8	ea29bb 9dd12d 9dd12d3 advisorsbot				['advisorsbot']	4 Proofpoint	8/23/2018		<a href="https://www.proofpoint.com/us/thr">https://www.proofpoint.com/us/thr</a>	

# Hints

## 2. Transform each malware to a grey image.

- A given malware binary is read as a vector of 8 bit unsigned integers and then organized into a 2D array. This can be visualized as a gray scale image in the range [0,255] (0: black, 255: white). The width of the image is fixed and the height is allowed to vary depending on the file size (Fig. 1). Tab. 1 gives some recommended image widths for different file sizes based on empirical observations.

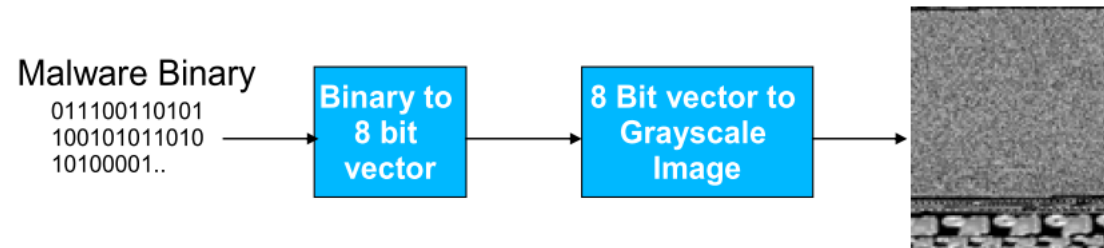


Fig.1 Visualizing Malware as an Image

Tab. 1: Image Width for Various File Sizes

File Size Range	Image Width
<10 kB	32
10 kB – 30 kB	64
30 kB – 60 kB	128
60 kB – 100 kB	256
100 kB – 200 kB	384
200 kB – 500 kB	512
500 kB – 1000 kB	768
>1000 kB	1024

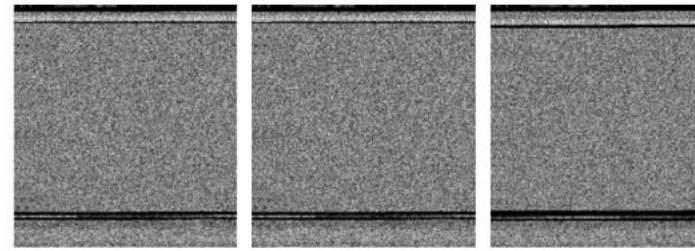
- There are several feature extraction methods: GIST, HOG, LBP, ... (Some methods even try to eat the image of the whole file)
- There are many machine learning methods: CNN, SVM, ...
- You could make some comparison, if possible.

## 3. Training and testing by machine learning model for detecting malware (grey image).

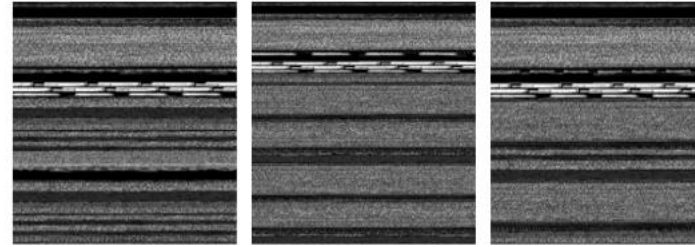
■ Programming Language is not limited.

# Malware Images

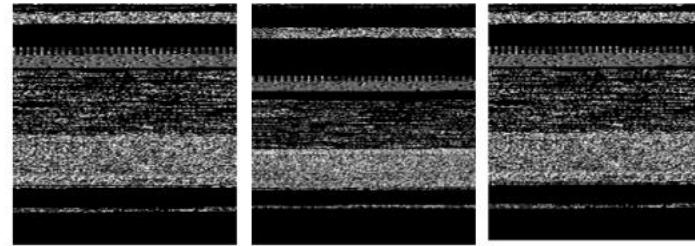
Instantaccess



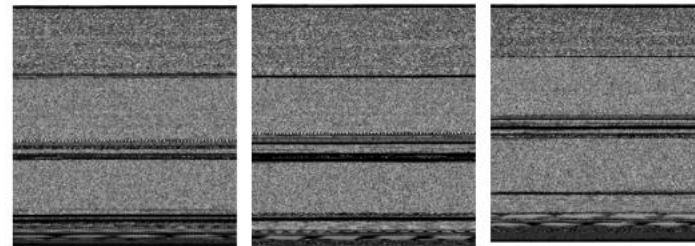
VB.AT



Dontovo.A



C2Lop.P



Allaple.L

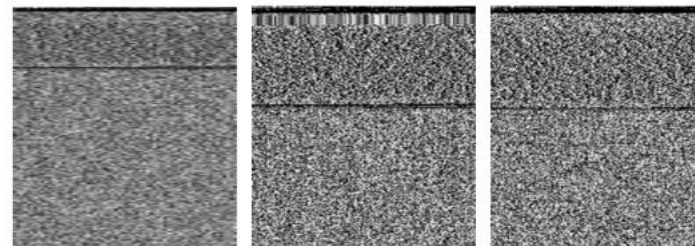


Fig. 2. Malware images of 5 different families. Each row having 3 variants of malware of same family [10]

# Some Feature Extraction Methods

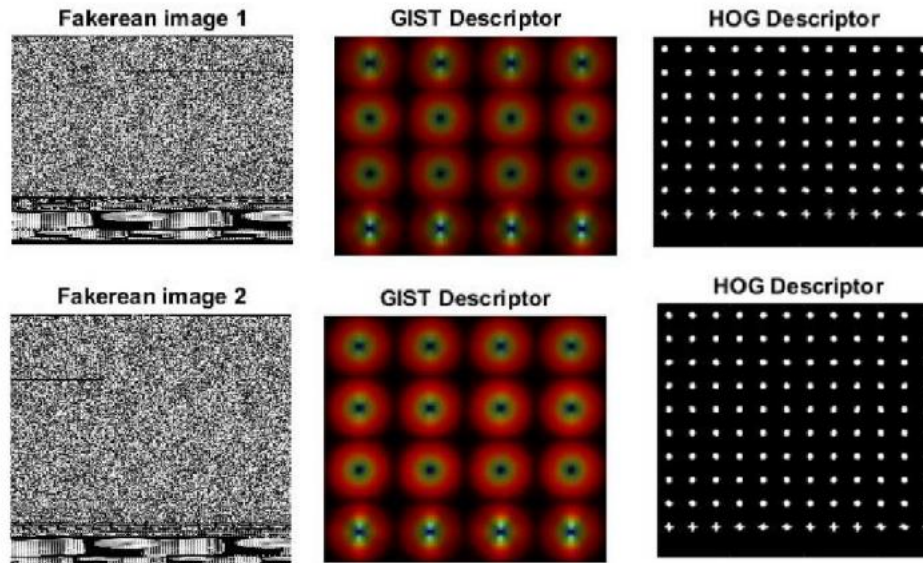


Fig. 3. GIST and HOG descriptor visualization of 2 variants Fakerean malware family.

# Upload the result to Moodle

- Upload your code and the ***detailed*** reports to Moodle.
  - **Code:** You must add some comments in your code for easy understanding.
  - **Report Files:** Document (Word) and Presentation (PPT).
    - Including the table outlining team member responsibilities.
  - Zipped to a file. The file name should be “Student ID\_HW3.zip”