

Malware
Analysis
Project

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### Introduction

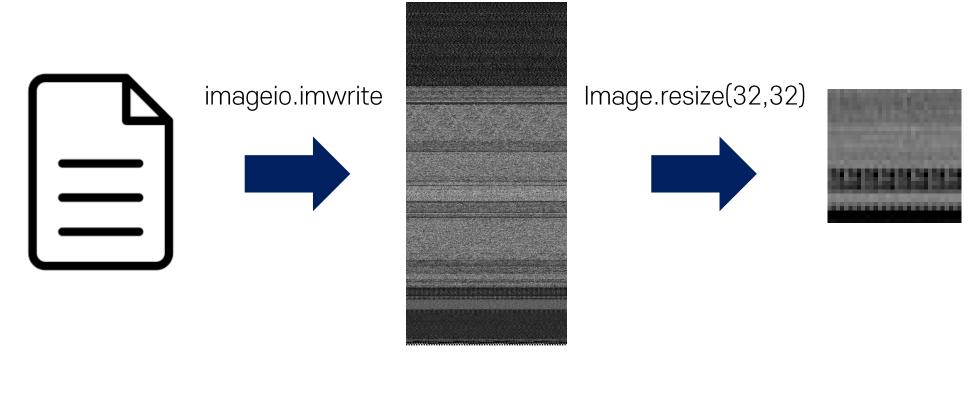
이번 프로젝트에서는 악성코드 파일을 이미지로 변환한 후, 이미지를 기반으로 악성 여부를 판단하고, 악성코드 계통(malware family)을 분류하는 인공지능 모델을 설계하고 학습하였습니다.

# Previous works

지난 발표 시기까지의 프로젝트 진행 요약



# Malware Image Visualization



.vir file .png file .png file

# Binary classification training using VGGNet

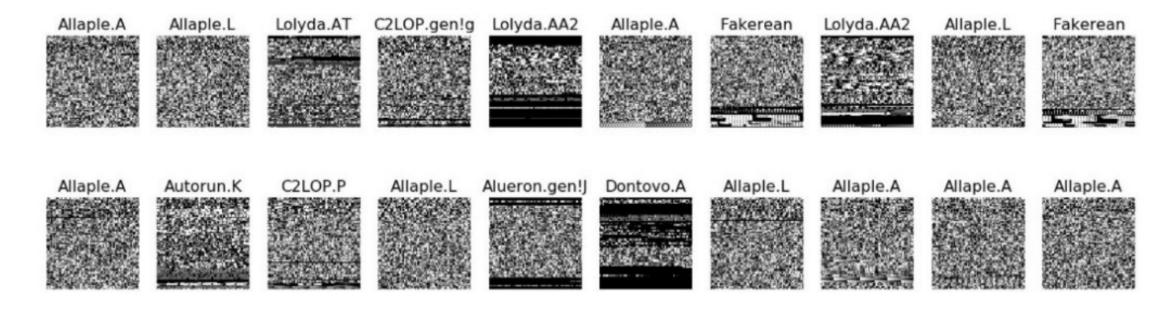
malware: 5007 non-malware: 5007 श

Layer(type)	Output Shape	# of params
con2d(Conv2D)	(None, 32, 32, 3)	84
activation (Activation)	(None, 32, 32, 3)	0
vgg16 (Functional)	(None, None, None, 512)	14714688
flatten (Flatten)	(None, 512)	0
dense (Dense)	(None, 32)	16416
dense_1 (Dense)	(None, 1)	33

Performance	
Train loss	loss: 0.0291
Train acc	accuracy: 97.87
Test loss	loss: 0.2332
Test acc	accuracy: 69.50



#### Malware Image Transformation



dataset	# of malware data points	# of non-malware data points
train set	7,000	3,000
test set1	5,000	5,000
test set2	5,000	5,000



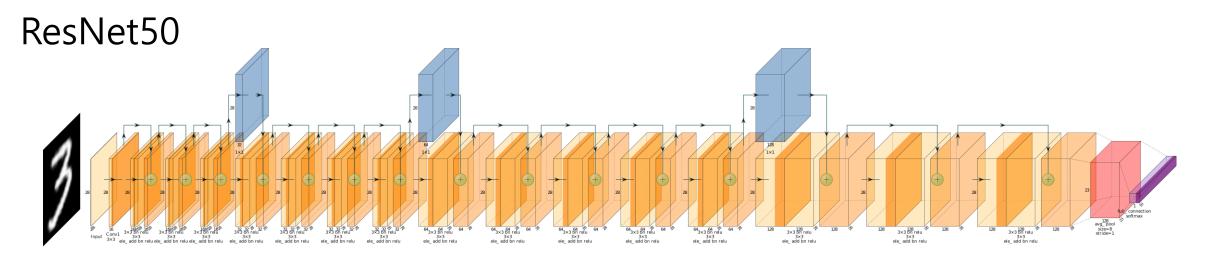


A model that receives a transformed 32 x 32 size image as input and dichotomously determines whether it is malware or not(non-malware)

dataset	# of malware data points	# of non-malware data points
train set	7,000	3,000
test set1	5,000	5,000
test set2	5,000	5,000



dataset	# of malware data points	# of non-malware data points
train set	5,000	5,000
test set	5,000	5,000

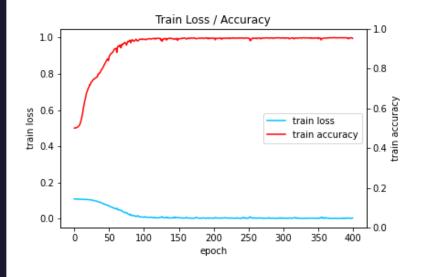


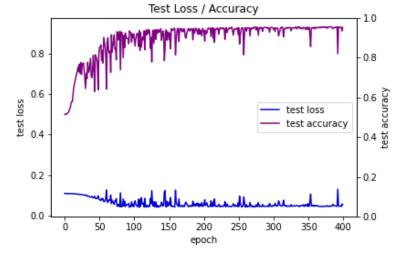
Malware or Non-malware?
Binary Classification

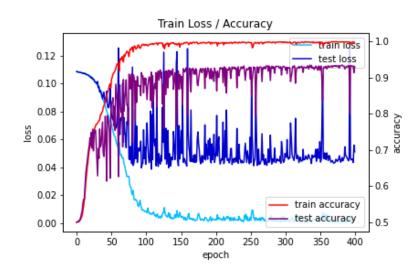


```
self.resnet_50.fc = nn.Sequential(
    nn.Linear(2048, 1000),
    nn.ReLU(inplace=True),
    nn.Linear(1000, 256),
    nn.ReLU(inplace=True),
    nn.Linear(256, 64),
    nn.ReLU(inplace=True),
    nn.ReLU(inplace=True),
    nn.Linear(64, 1),
    nn.Sigmoid()
)
```

#### Model Performance



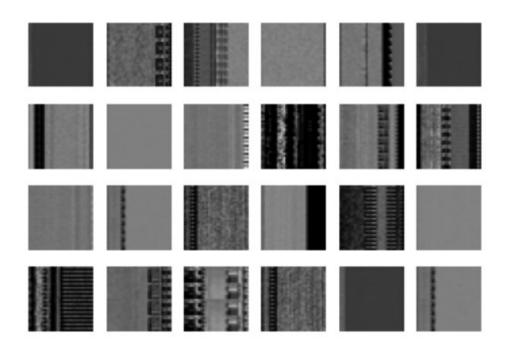




- Train Accuracy: 99.5%
- Test Accuracy: 93.0%
- Epoch: 400

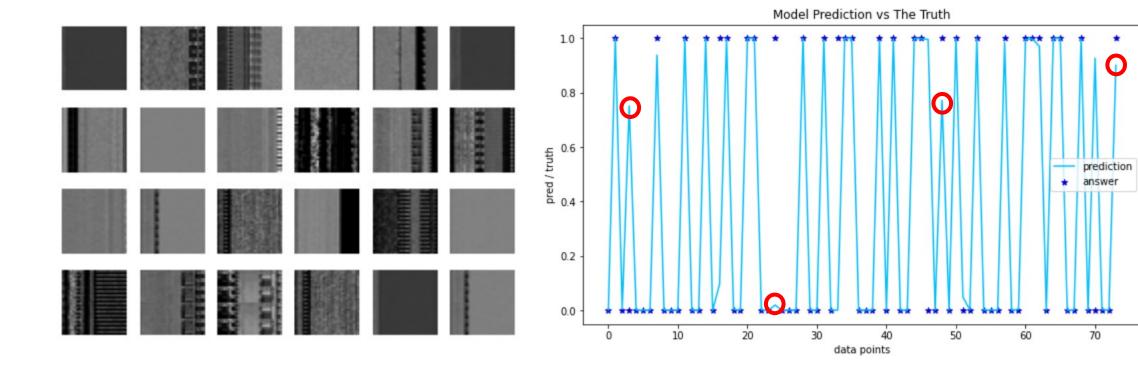
- Learning rate: 0.001
- Loss: nn.BCELoss
- Optimizer: SGD

#### Model Performance

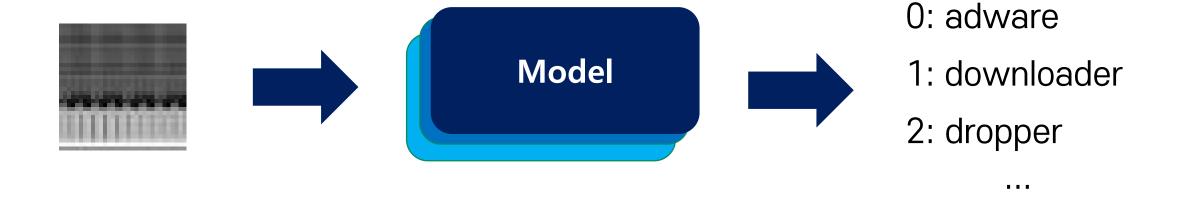


```
pred=0.0014788481639698148 | ans=0
image 0 |
image 1 |
         pred=0.9997912049293518 |
         pred=0.005231290124356747 |
image 2 |
image 3 l
         pred=0.7512921690940857 | ans=0
         pred=0.0017309949034824967 | ans=0
image 4 |
image 5 |
         pred=0.001742606982588768 |
                                      ans=0
         pred=0.000600828614551574 |
image 6 l
                                      ans=0
         pred=0.9383810758590698 |
image 7 |
         pred=0.0016865450888872147 | ans=0
image 8 |
image 9 l
         pred=0.0007257546531036496
                                       ans=0
image 10 | pred=0.0006229666178114712 | ans=0
image 11 | pred=0.9989845156669617 | ans=1
          pred=0.001240861602127552 |
image 13 | pred=0.0006093949778005481 |
image 14 | pred=0.9991174340248108 |
image 15 | pred=0.00038419407792389393 | ans=0
image 16 | pred=0.0970044657588005 |
                                     ans=1
          pred=0.9984909296035767
                                     ans=1
image 18 | pred=0.00386615376919508
                                      ans=0
image 19 | pred=0.0015628642868250608 | ans=0
image 20 | pred=0.9997197985649109
                                     ans=1
image 21 | pred=0.9999942779541016
                                     ans=1
image 22 | pred=0.002477513626217842 |
                                       ans=0
image 23 | pred=0.0005884646088816226
                                        ans=()
```

#### Model Performance







A model that receives a transformed 32 x 32 size image as input and determines its malware family such as adware, downloader, etc.

dataset	# of malware data points	# of non-malware data points
train set	7,000	3,000
test set1	5,000	5,000
test set2	5,000	5,000



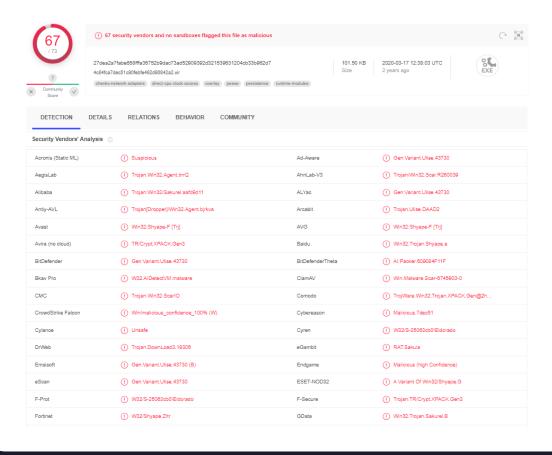
dataset	# of malware data points
train set	7,000
test set1	5,000
test set2	5,000

dataset	# of malware data points
train set	7,000
test set1	5,000
test set2	5,000



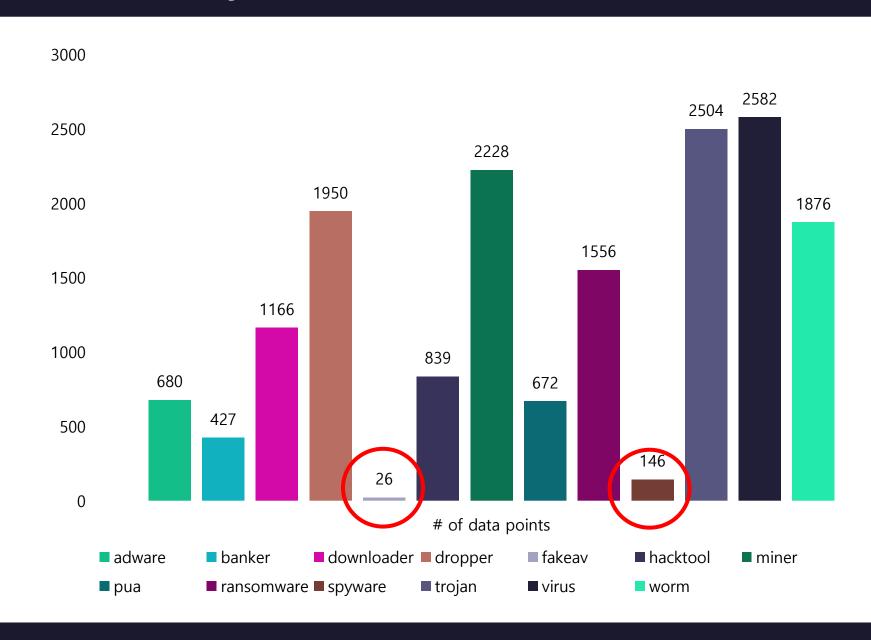
dataset	# of malware data points
ALL	17,000

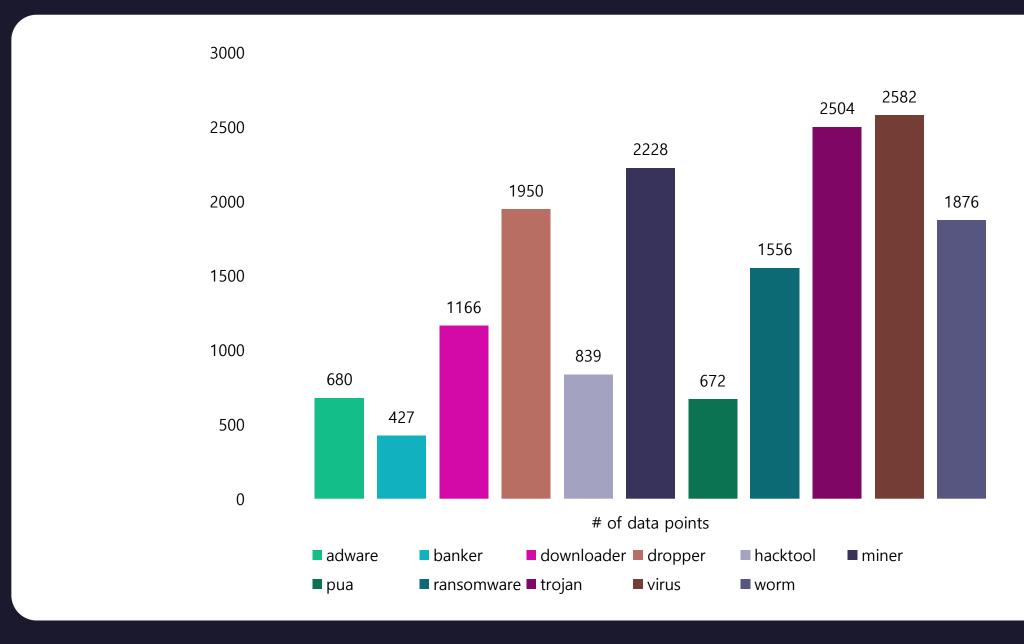
dataset	# of malware data points
ALL	17,000

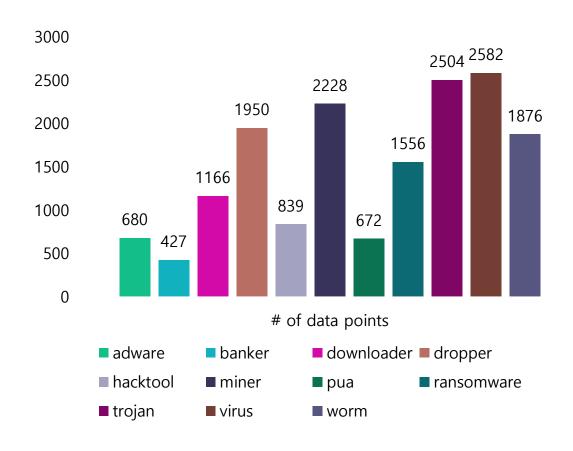


Extracts labels from examination data based on VirusTotal.

- 1. majority rule
- 2. exceptions for trojan
  - 1) only case -> trojan
  - 2) else -> get the 2<sup>nd</sup> decision





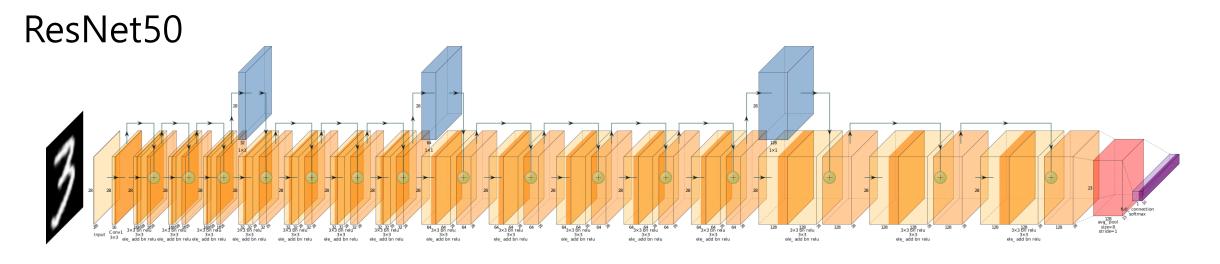


We performed over-sampling

With imblearn.over sampling.RandomOverSampler

Object to over-sample the minority class(es) by picking samples at random with replacement.

The bootstrap can be generated in a smoothed manner.



Which malware family does this

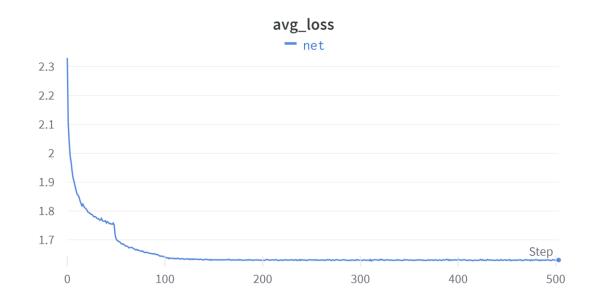
virus belong to?

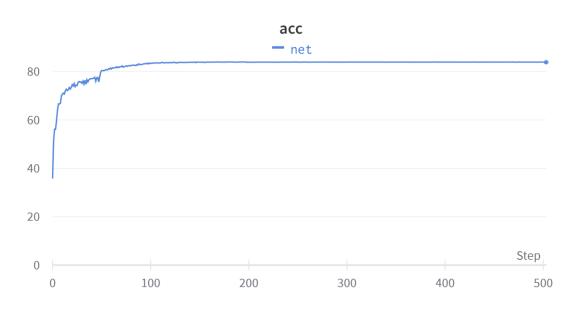
multi-class classification



```
self.resnet_50.fc = nn.Sequential(
    nn.Linear(2048, 1000),
    nn.ReLU(inplace=True),
    nn.ReLU(inplace=True),
    nn.Linear(256, 64),
    nn.ReLU(inplace=True),
    nn.ReLU(inplace=True),
    nn.Linear(64, 11),
    nn.Softmax()
)
```

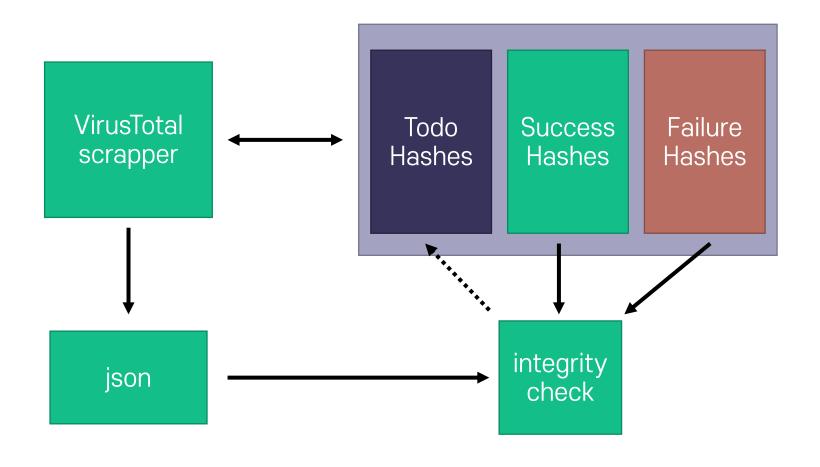
#### Performance

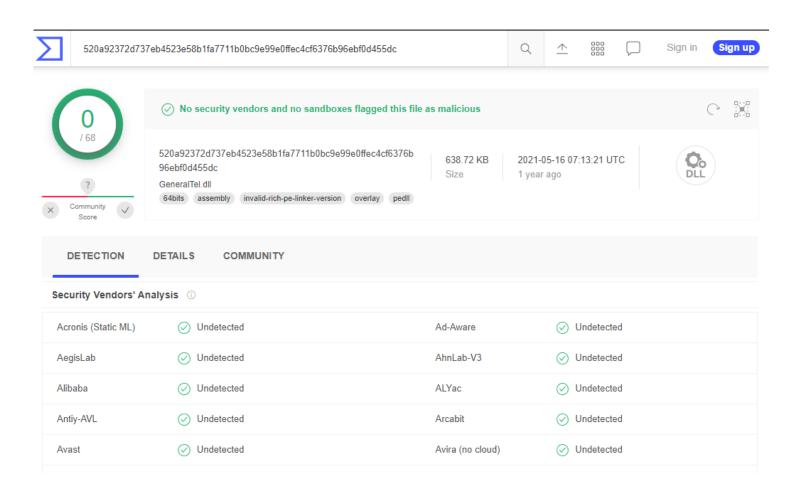




test set accuracy: 84%







- Folder: 40,000
- Files: 414,784
- Capacity: 2.84GB
- base.json behaviours.json bundled\_files.json contacted\_domains.json contacted\_ips.json contacted\_urls.json dropped\_files.json execution\_parents.json pe\_resource\_children.json pe\_resource\_parents.json

```
"data": {
    "attributes": {
        "vhash": "064046551d1500f1z14z211b5z3dz17z45z",
        "trid": [
```

#### Type description

- Win32 DLL
- Win32 EXE
- JavaScript
- ...

#### TrID

- Win32 Executable MS Visual C++ (generic)
- OS/2 Executable (generic)
- Microsoft Visual C++ compiled executable (generic)
- ...

```
'popular_threat_classification": {
   "popular_threat_category": [
           "value": "trojan"
           "value": "dropper"
   "popular threat name": [
           "count" 12
           "value": "ddos"
           "value": "rincux"
```

#### Popular threat category/name

- Trojan
- Worm
- Virus
- Dropper
- DDOS
- •

```
"packers": {
    "PEiD": "Microsoft Visual C++"
},
```

#### **Packers**

- PEiD
  - UPX
  - Microsoft Visual C++ v6.0 DLL
  - .NET executable
  - •

```
"entropy": 2.0,
   "chi2" 252.0
   "filetype": "ASCII text",
   "md5": "2f5f7485fd0c9653a9338251cad59e8b",
   "size": 4
"entry point": 9844
"machine type": 332,
"sections": [
       "virtual size": 6298.
       "entropy": 5.13,
       "raw size" 8192
       "md5": "6dfbc6899241990b3bc003eef552e520"
       "virtual size": 2712,
       "entropy": 4.11,
       "virtual_address": 12288,
       "md5": "29aa83f4db54de6e04f0b3f5c0e545df"
```

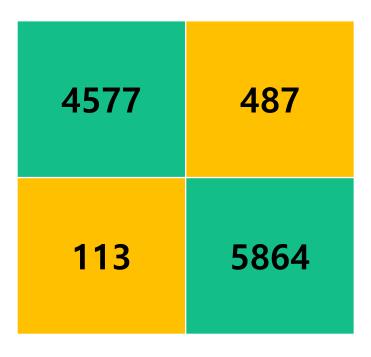
#### Sections

- Overlay
  - Entropy
  - Chi2
- .text, .rdata, .rsrc ···
  - Entropy

type_des	best_trid_	best_trid	type_exte	has_signa	packer	type_tag	overlay_e	overlay_c	overlay_f	dropped_	beha
Win32 Dl	Win32 Exe	33.7	dll	0	unknown	pedll	2	252	ASCII text	0	
Win32 DI	Win64 Exe	61.7	dll	0	unknown	pedll	2	252	ASCII text	0	
Win32 EX	OS/2 Exec	25.2	exe	1	unknown	peexe	0.004528	2.58E+09	Data	0	
Win32 DI	Windows	83.7	dll	1	unknown	pedll	7.398872	22471.63	Data	0	
Win32 EX	Win32 Exe	42.7	exe	0	unknown	peexe	2	252	ASCII text	0	
Win32 DI	Win32 Exe	33.7	dll	1	Microsoft	pedll	2	252	ASCII text	0	
Win32 DI	Win32 EX	60.7	dll	0	UPX	pedll	2	252	ASCII text	0	
Win32 EX	Microsoft	33.5	exe	0	unknown	peexe	2	252	ASCII text	0	
Win32 EX	Win32 Exe	41	exe	0	Microsoft	peexe	4.818212	96692040	Data	0	
Win32 EX	Win32 Exe	58.7	exe	1	Microsoft	peexe	7.406668	7600.766	Data	0	
Win32 DI	Win64 Exe	82	dll	0	unknown	pedll	1.212236	52465.13	ASCII text	0	
Win32 EX	Win64 Exe	42.3	exe	1	UPX v0.89	реехе	3.649286	5206.675	Data	0	
Win32 EX	Win64 Exe	82	exe	1	unknown	peexe	7.999897	268.3149	Data	0	
Wirss Di	Caparia N	650	االہ	1	Microsoft	nadli	1	רור	ACCII +out	۸	

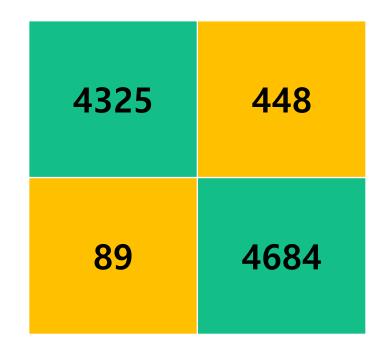
hash type\_description best trid type best\_trid\_probability type\_extension has\_signature packer type\_tag overlay\_entropy overlay chi2 overlay\_filetype dropped\_files\_count behaviours count contacted\_domains\_count contacted\_ips\_count contacted\_urls\_count .text\_entropy .bss\_entropy .rdata\_entropy .data\_entropy .xdata\_entropy .idata\_entropy .pdata\_entropy .rsrc entropy .reloc entropy .CRT entropy

Test set 1

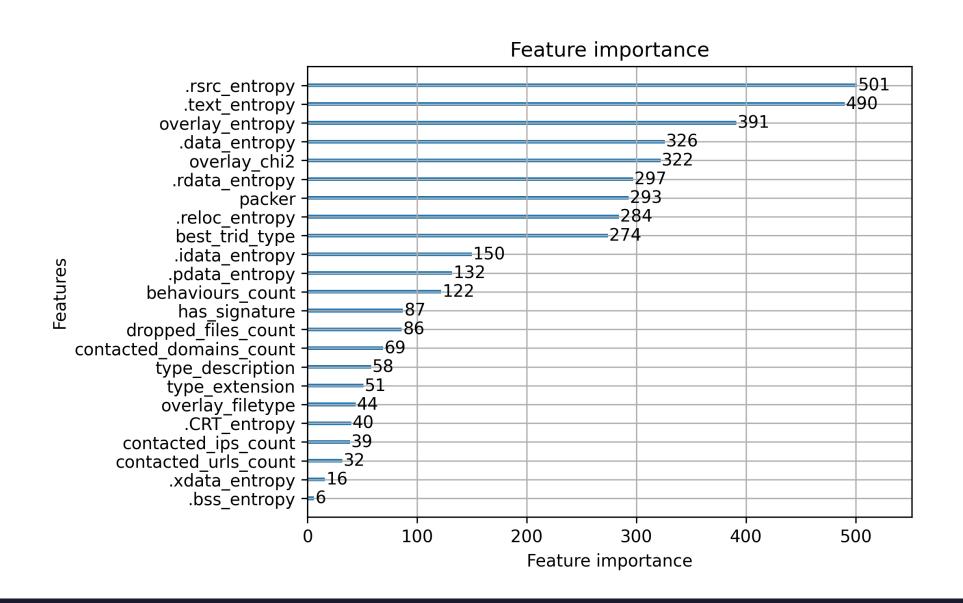


Accuracy: 96.15% Precision: 93.99%

Test set 2



Accuracy: 94.37% Precision: 91.27%



### Thank You ©

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