

Team: Binary Beasts

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Introduction

We have performed dynamic malware analysis using different machine learning algorithms.

Project Outline:

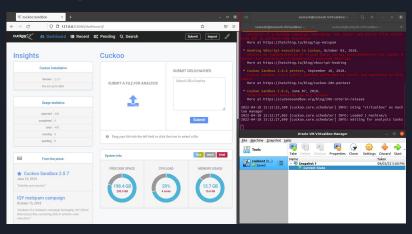
- Collected Portable Executable (PE) files from C3i center, IIT Kanpur.
- Automated installation and configuration of Cuckoo sandbox.
- Then we obtained an analytics report of each PE file using cuckoo sandbox in JSON format.
- Parsed JSON files to collect significant features.
- Pre-processed the data and performed dimensionality reduction
- Applied various ML models and compared their accuracies.

Cuckoo Sandbox Installation

- We had configured Cuckoo Sandbox in a virtual environment.
- Cuckoo host is in Ubuntu 20.04 & Cuckoo guest is in Windows 7 Ultimate.
- As Cuckoo is still based on Python2.7 which is outdated as of now, we had faced errors several times during the setup.
- To overcome various errors, we had used various scripts available on the internet but none of them make installation without any errors.
- So, we have corrected those collection of scripts and used these scripts to install cuckoo on virtual machine and exported that virtual machine in .OVA format.
- Now anyone can install cuckoo by using corrected script or can use cuckoo without installing it by just importing our virtual machine image in their virtualbox.

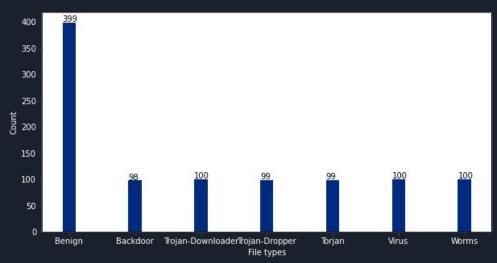
Cuckoo Sandbox Configuration

- The Cuckoo we had used for our analysis is of version 2.0.6
- We had used Cuckoo guest in "Host-only" network environment.
- For our analysis we configured *cuckoo.conf* for both static & behavioural analysis.
- We had disabled memory dump option as it was not needed for our analysis and also it was taking very huge memory space.
- Also, we enabled IP forwarding. So, the internet connection gets routed from host machine to cuckoo guest VM.



Dataset Creation

We have collected JSON files for 995 classified PE files.



Cuckoo generated Analytic Report for these above file.

JSON file created by Cuckoo sandbox contains lots of information like PE information, signatures, process information, etc.

Features Extracted

Collected the following information as features:

- API Calls
- Files written
- Files deleted
- Regkey written
- Regkey deleted
- DLL loaded
- Duration

```
"info": {
    "added": 1648160439.192953.
    "started": 1648168925.14669.
    "duration": 67,
    "ended": 1648168992.926998,
    "owner": null,
    "score": 1.2.
    "id": 72,
    "category": "file",
        "fetch head": "13cbe0d9e457be3673304533043e992ead1ea9b2"
    "monitor": "2deb9ccd75d5a7a3fe05b2625b03a8639d6ee36b",
    "package": "exe",
    "route": "none",
    "custom": null.
    "machine": {
        "status": "stopped",
        "name": "cuckoo2",
        "label": "cuckoo2",
        "manager": "VirtualBox",
        "started on": "2022-03-25 00:42:06",
        "shutdown on": "2022-03-25 00:43:12"
    "platform": "windows",
    "version": "2.0.7",
    "options": "procmemdump=yes, route=none"
"procmemory": [
        "regions": [
                "protect": "rw",
                "end": "0x00020000",
                "addr": "0x00010000",
                "state": 4096,
                "offset": 24,
                "type": 262144,
                "size": 65536
```

Dataset Details

- We treated each api calls, file names and regkey names as features.
- For api calls each cell stores the number of time that call is made.
- For DLLs, files and regkey, each cell contains binary data.
- Duration represents the time duration for which PE is executed.
- Score provided by Cuckoo sandbox is not taken as a feature.

Dataset details after pre-processing

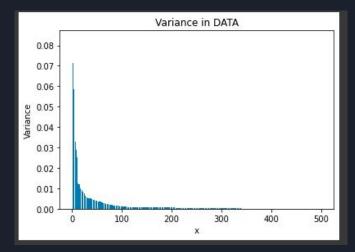
- Dataset details:
 - 14820 columns; 995 rows
- Classification Labels:
 - Normal
 - o Trojan-Dropper
 - Worm
 - o Trojan-Downloader
 - Virus
 - o Backdoor
 - Trojan

Stages in Machine Learning:

• Feature Scaling: Normalised the data before implementing our model.

• Feature Reduction: Implemented PCA to reduce from 14k+ features (original) to top K

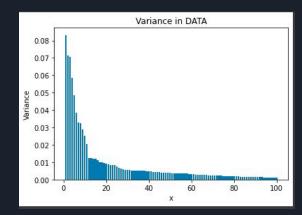
features.



 Implementing and comparing various ML models proven to perform well in classification problems

Stages in Machine Learning:

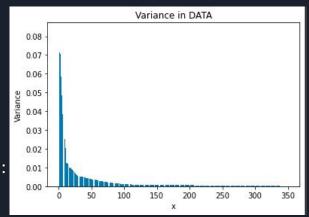
• By Seeing, Variance plot, We took only Top 100 Features :



Model (With 100 features)	Accuracy
SVM	72.36
Decision Tree	89.44
Random Forest	93.46
Random Forest K-Nearest neighbour	93.46 89.94

Stages in Machine Learning:

• By Seeing, Variance plot, We took only Top 350 Features :



Model (With 350 features)	Accuracy
SVM	80.40
Decision Tree	91.95
Random Forest	93.46
K-Nearest neighbour	80.40
Deep Neural Network	87.84

ML Classification Models (with Top-500 features):

We tested various ML models with various parameter combinations. We experimented with the parameters of each model and presenting the best performing models as follows:

1. Multiclass SVM (best accuracy: 91.46),

Kernel	Accuracy
Poly	41.21
RBF	67.34
Linear	91.46

ML Classification Models:

2. **Decision Tree** (best accuracy: 91.45)

Criteria	Accuracy
Gini	88.00
Entropy	91.45

3. Random Forest (best accuracy: 93.46)

Estimators	Accuracy
1000	93.46
500	92.46
100	92.46

ML Classification Models:

4. **KNN** (best accuracy: 78.89)

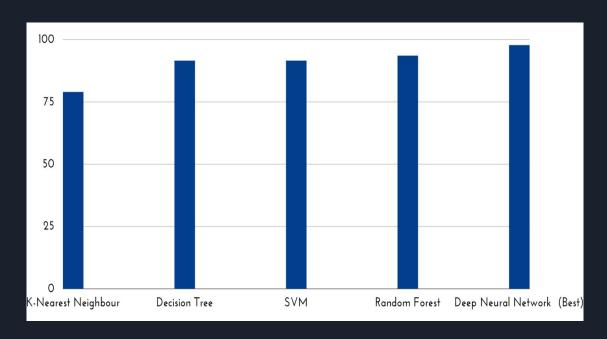
N-Neighbours	Accuracy
15	74.87
7	73.86
5	76.88
3	78.9

5. Deep Neural Network:

Layers	Accuracy
3	94.87
4	97.69

Results:

<u>Models</u>	Best Accuracy
K-Nearest Neighbour	78.89 %
Decision Tree	91.45 %
SVM	91.46 %
Random Forest	93.46 %
Deep Neural Network (Best)	97.69 %



Limitations

- Small dataset: Extraction of Analytics Report from Cuckoo took lots of time.
- Limited significant features: Taking all fields from JSON files created large feature set.
- Windows Malware: Our analysis only covers windows malwares.

Future work

- Extend our analysis for larger dataset.
- Integrate all the pieces to create final malware detection tool.
- Deploy our tool for detecting malicious PE files.

Thank you!