

CSM501: CLASSIFICATION OF MALWARE USING IMAGE REPRESENTATION

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INTRODUCTION



The Internet: risk of getting attacked.



Malware detection and classification: one of the most crucial problems in the field of cyber security.



Malware analysis is the process of dissecting a binary file to understand its working and then devising methods to identify it and other similar files



Signature-based technique fail to cope with code obfuscation and fail to effectively identify newly arrived threats.

LITERATURE SURVEY

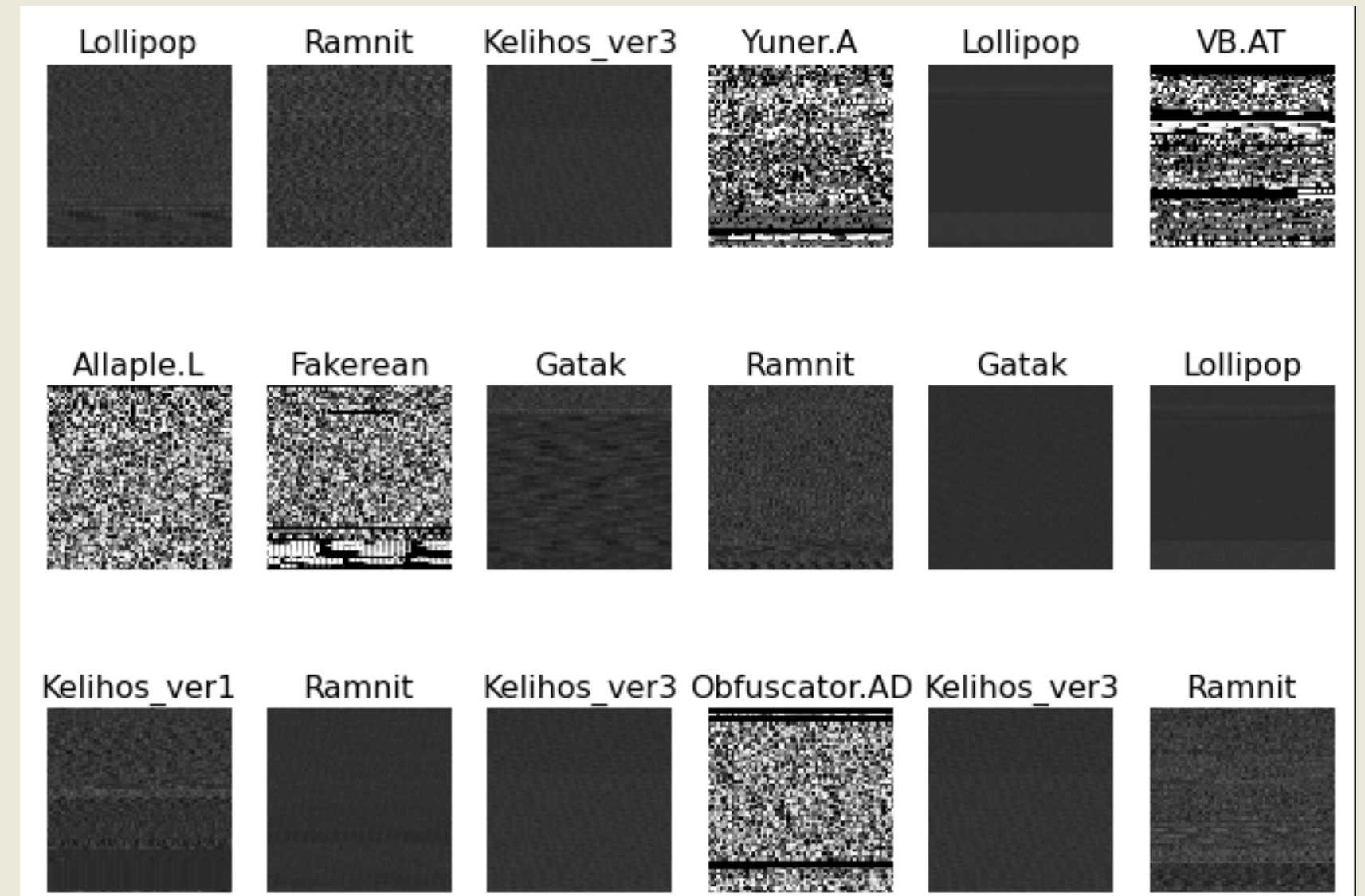
| Sr No. | Paper Title, Year | Author | Dataset/ Method/ Algorithm Used | Contribution/ Advantages | Improvements | Conclusion/ Comments |
|--------|--|--|--|--|--|--|
| 1 | Semi-Supervised Learning for Unknown Malware Detection, 2011 | Santos, I., Nieves, J., and Bringas, P. | Signature based method | Fast and highly accurate method for signatures known | Database has to be prepared manually, time-consuming | Although the parameter & method used are accurate, tedious to add new data |
| 2 | Survey on malware detection methods, 2009 | Vinod, P., Jaipur, R., Laxmi, V., and Gaur, M. | Anomaly based approach | Able to identify new unseen malware samples | False alarm rate very high | High rate of incorrect classification makes it not suitable |
| 3 | Data Mining Methods for Detection of New Malicious Executables, 2001 | Schultz, M. G., Eskin, E., and Zadok, F. | Multinomial Naive Bayes algorithm to classify a malware dataset of 3265 malicious and 1001 benign sample | First to try performing malware analysis using data mining techniques. | Require large datasets which are difficult to obtain | Although method gives good accuracy, not large enough dataset limits it |
| 4 | Limits of Static Analysis for Malware Detection, 2007 | Moser, A., Kruegel, C., and Kirda, E. | Scheme based obfuscation technique to explore the drawbacks of static analysis approaches | Observations showed static analysis can be easily evaded if the malware is obfuscated or packed. | Dynamic analysis approach was needed | Only static analysis is not enough to find best method to classify malware |

| Sr No. | Paper Title, Year | Author | Dataset/ Method/ Algorithm Used | Contribution/ Advantages | Improvements | Conclusion/ Comments |
|--------|---|---|---|--|--|--|
| 1 | Dotplot patterns: A literal look at pattern languages, 1995 | Helfman,J. | Dotplot data visualisation technique to software programs | Useful for design of software systems through Successive Abstraction | Requires manual analysis, time consuming | Although the method is novel, tedious to add new data |
| 2 | Automated Mapping of Large Binary Objects Using Primitive Fragment Type Classification,2010 | Conti, G., Bratus, S., Sangster, B., Ragsdale, R., Supan, M., Lichtenberg, A., PerezAlemany, R., and Shubina, A | Automated binary mapping technique using Byte plot visualisation, automated the problem of finding the start and stop offsets of each distinct region within a binary | The Byte plot visualization helped in finding distinctive patterns, even transformations such as encoding, encryption are applied. | Work was limited only to identify primitive patterns in a binary file | Limitation in data that can be accepted |
| 3 | Malware Images: Visualization and Automatic Classification, 2011 | Nataraj, L., Karthikeyan, S., Jacob, G., and Manjunath, B. | Converted all the malware sample to grayscale byte plot representations and extracted texture based features from the malware image | Showed that image processing based malware classifying techniques can classify malware more quickly than existing dynamic approaches | Their approach has a huge computational overhead of calculating texture based feature | Although method gives good accuracy, has large computational costs |
| 4 | Malware Analysis and Classification using Artificial Neural Network,2015 | Makandar, A. and Patrot, A. | Converted malware into a 2-dimensional grayscale image and classified the samples using texture based features | Reported an accuracy of 96.35%. | Based on Mahenhur dataset for experiments, which is comprised of 3131 binaries samples from 24 unique malware families | Current accuracy based on given dataset |

PROPOSED SYSTEM

PROPOSED SYSTEM

- A novel approach is to use image representation of binaries for classification.
- Machine learning algorithms can be used to classify the files on the basis of features of image



- 34 malware families + non malware
- A comprehensive solution is a website that allows for easy classification of malware

Malware Classification

Home

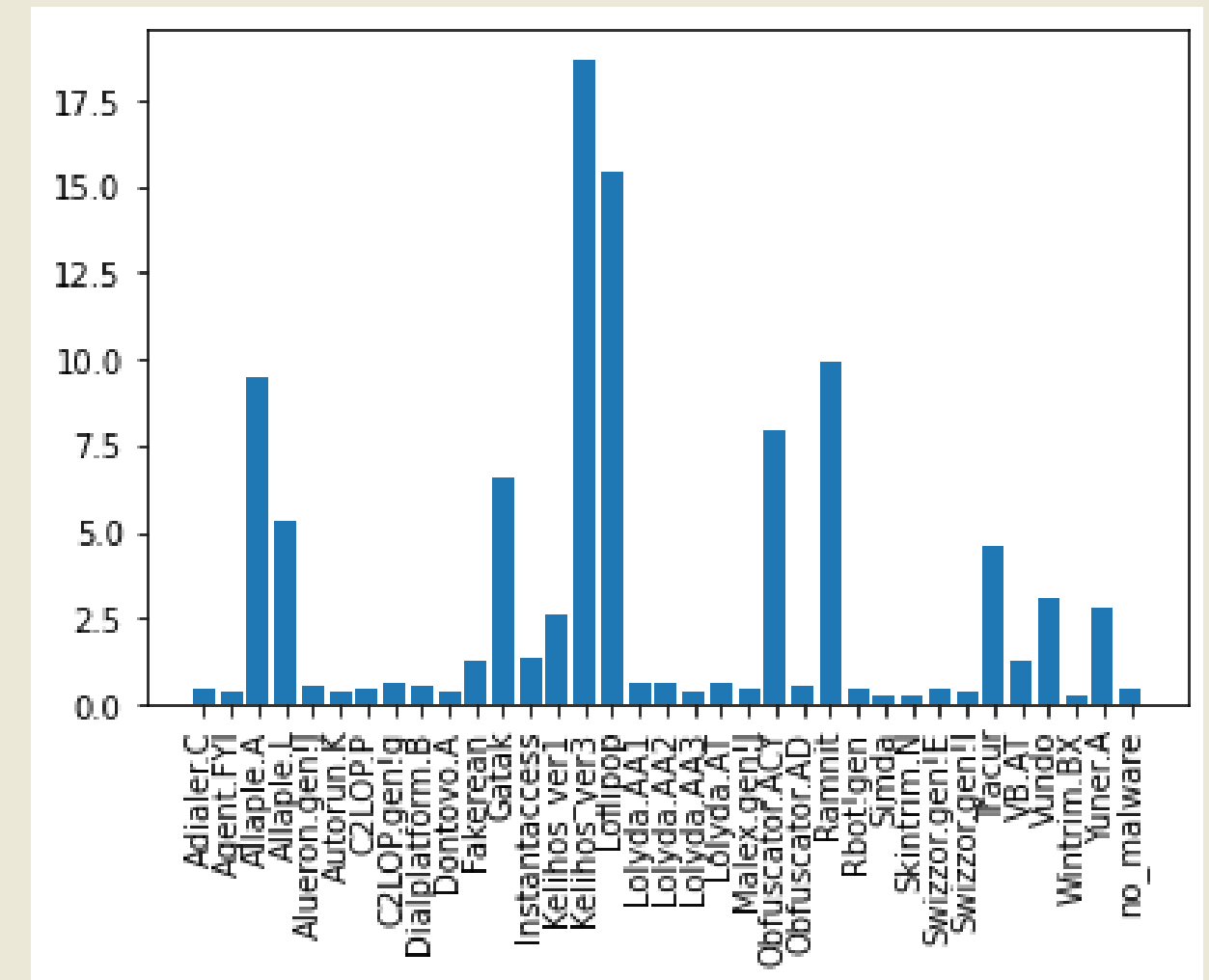
Check the Type of Malware

Document:

Choose File

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Upload



METHODOLOGY

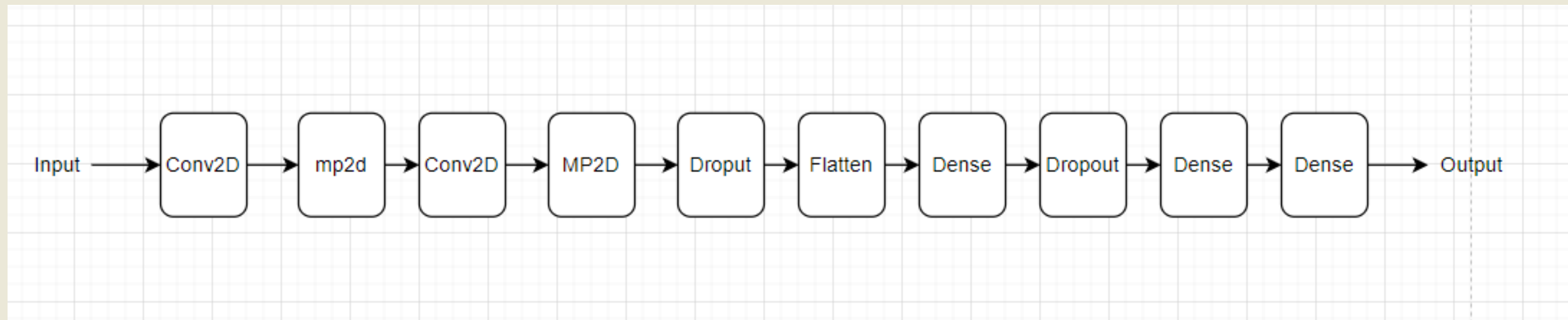
Epochs: 70

Kernel Size: 3 x 3

Accuracy: 86%

Activation Function: 3 Relu, 1 Softmax

No. of layers: 10



OUR CONTRIBUTION

- Added images for non malware files
- Added new malware families for it to adapt to the ever increasing newly arrived threats
- New dataset comprising of malware and non malware files to be uploaded on Kaggle

THANK YOU