



Rajshahi University of Engineering & Technology
Department of Computer Science & Engineering

Title: Vision Based Malware Classification Framework Based On Neural Network

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Table of contents

01 Introduction

02 Motivation

03 Objectives

04 Literature Review

05 Dataset

06 Proposed Method &
Workflow

07 Implementation

08 Result Analysis

09 Model Comparison

10 Future Work

11 Conclusion

12 Reference

Introduction

Malware, or malicious software, refers to any type of program or code designed to harm computer systems or steal data.

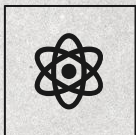
There are a varieties of malwares present today , some of which are - Zeus, My-doom, Storm-worm, Slammer etc.

Different approaches have been used to detect & classify malwares. But malwares have emerged in such an extraordinary way that the traditional approaches are now not very effective.

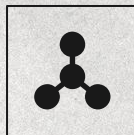




Motivation



Traditional **signature-based** methods are less effective



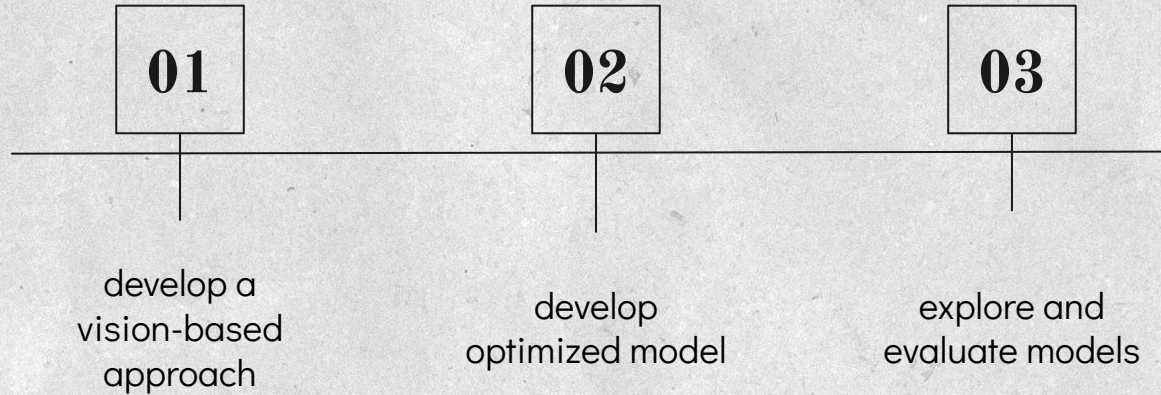
A **vision-based** malware classification framework will work well



Neural Network has emerged as a promising solution



Objectives



Literature review



Paper Title & Author	Dataset & Models Used	Accuracy
Malware Images: Visualization and Automatic Classification. [1] L. Nataraj (2011)	<ul style="list-style-type: none">Malware binaries are visualized as gray-scale imagesClassified using standard image features (PCA) and knn model by 10 fold cross validation using k=3	Accuracy - 98%
Using convolutional neural networks for classification of malware represented as images. [2] Daniel Gibert (2018)	<ul style="list-style-type: none">Malimg dataset & Microsoft Big datasetConvolution neural network(CNN) is used to classify	Accuracy: Malimg dataset - 96% & Microsoft Big dataset - 97.3%
Towards Building an Intelligent Anti-Malware System: A Deep Learning Approach using Support Vector Machine (SVM) for Malware Classification.[3] Abien Fred M. Agarap (2019)	<ul style="list-style-type: none">Malimg datasetCNN-SVM, GRU-SVM, and MLP-SVM models used for classification	Accuracy: 84.92%



Dataset Description

The deep learning (DL) model in this study will be evaluated on the Maling dataset[4],

- The dataset contains 9,339 malware samples
- These malware samples are from 25 different malware families



Dataset Description Cont.

Table 1 shows the frequency distribution of malware families and their variants in the Maling dataset[4].

No.	Family	Family Name	No. of Variants
01	Dialer	Adialer.C	122
02	Backdoor	Agent.FYI	116
03	Worm	Allaple.A	2949
04	Worm	Allaple.L	1591
05	Trojan	Alueron.gen!J	198
06	Worm:AutoIT	Autorun.K	106
07	Trojan	C2Lop.P	146
08	Trojan	C2Lop.gen!G	200
09	Dialer	Dialplatform.B	177
10	Trojan Downloader	Dontovo.A	162
11	Rogue	Fakerean	381
12	Dialer	Instantaccess	431
13	PWS	Lolyda.AA 1	213
14	PWS	Lolyda.AA 2	184
15	PWS	Lolyda.AA 3	123
16	PWS	Lolyda.AT	159
17	Trojan	Malex.gen!J	136
18	Trojan Downloader	Obfuscator.AD	142
19	Backdoor	Rbot!gen	158
20	Trojan	Skintrim.N	80
21	Trojan Downloader	Swizzor.gen!E	128
22	Trojan Downloader	Swizzor.gen!I	132
23	Worm	VB.AT	408
24	Trojan Downloader	Wintrim.BX	97
25	Worm	Yuner.A	800

Table 1: Malware families found in the Maling Dataset[4].



Dataset Description Cont.

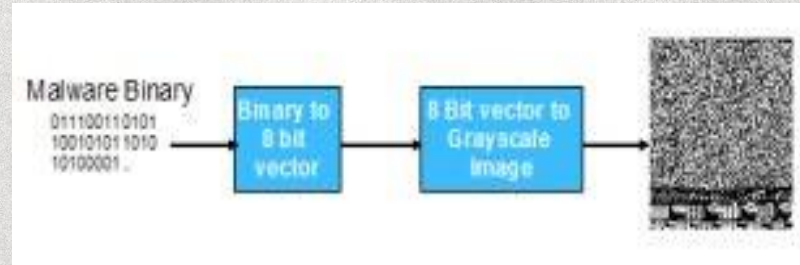


Figure 1: Image from [4]. Visualizing malware as a grayscale image.

Proposed Method

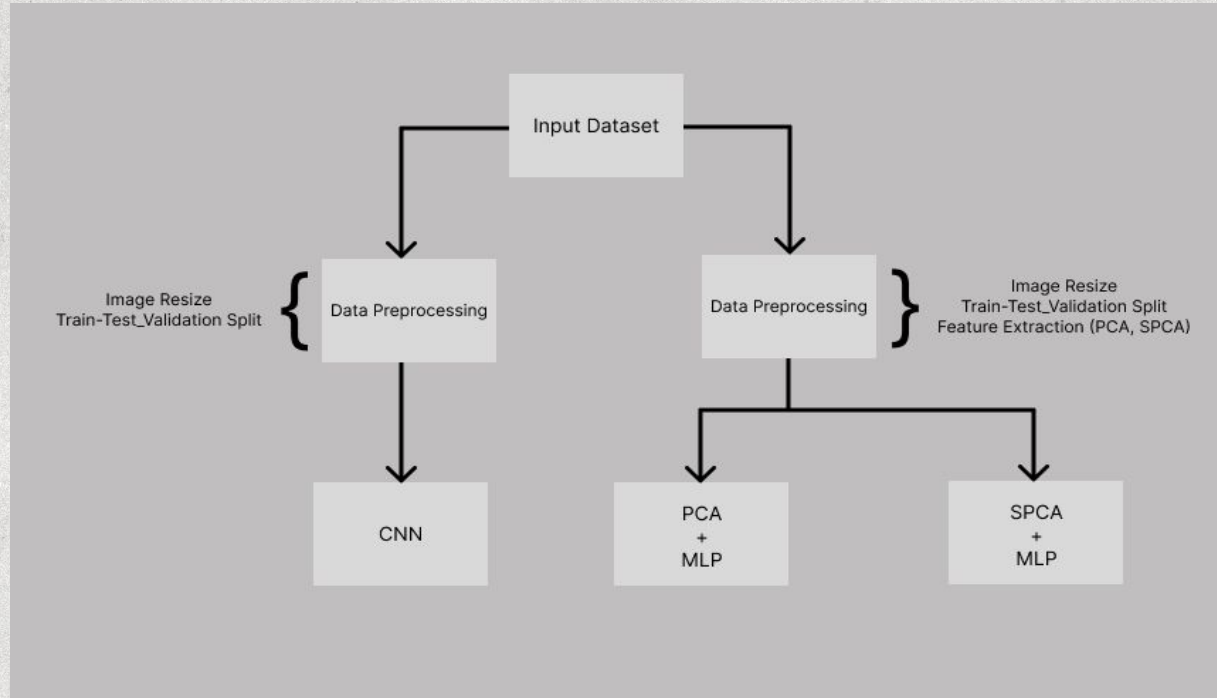


Figure 2: Workflow of the proposed method

Proposed Method Cont.

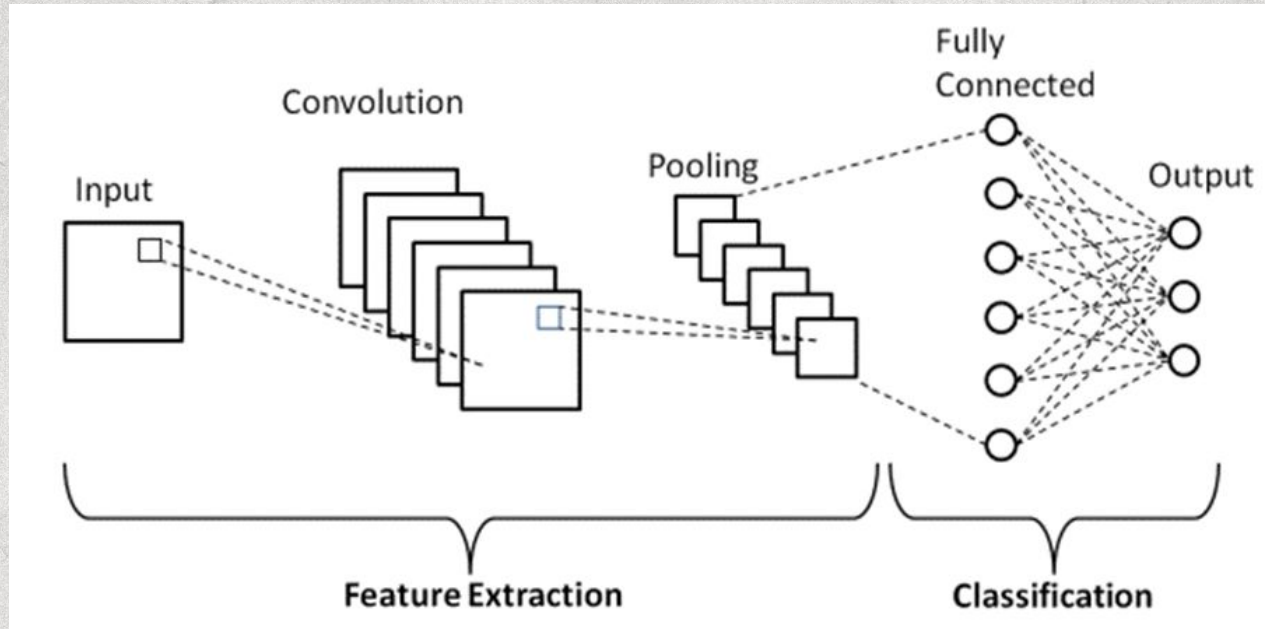


Figure 3 : Basic CNN architecture

Proposed Method Cont.

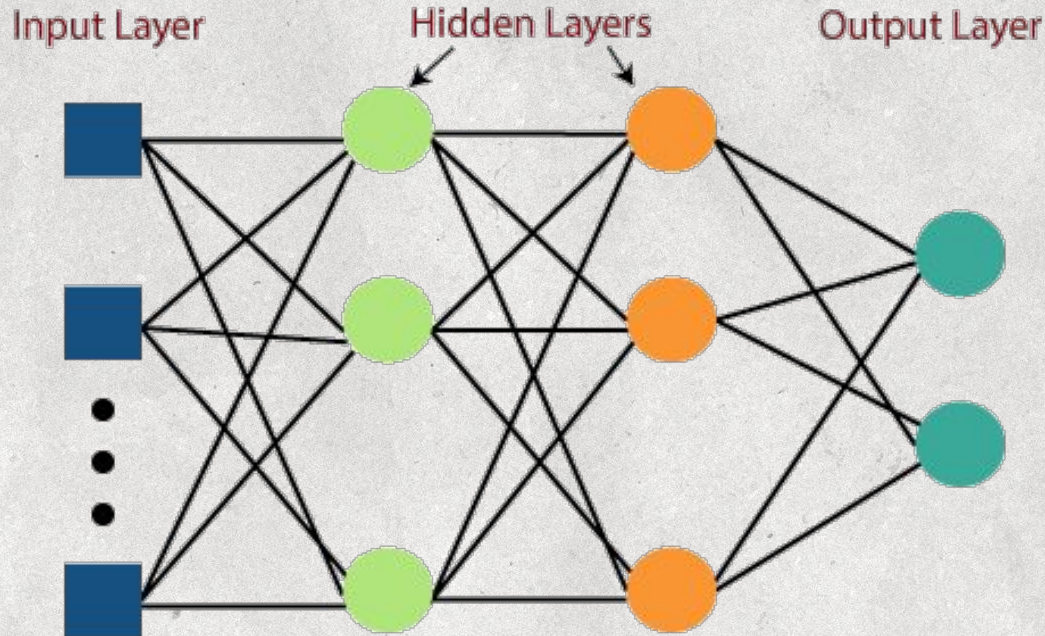
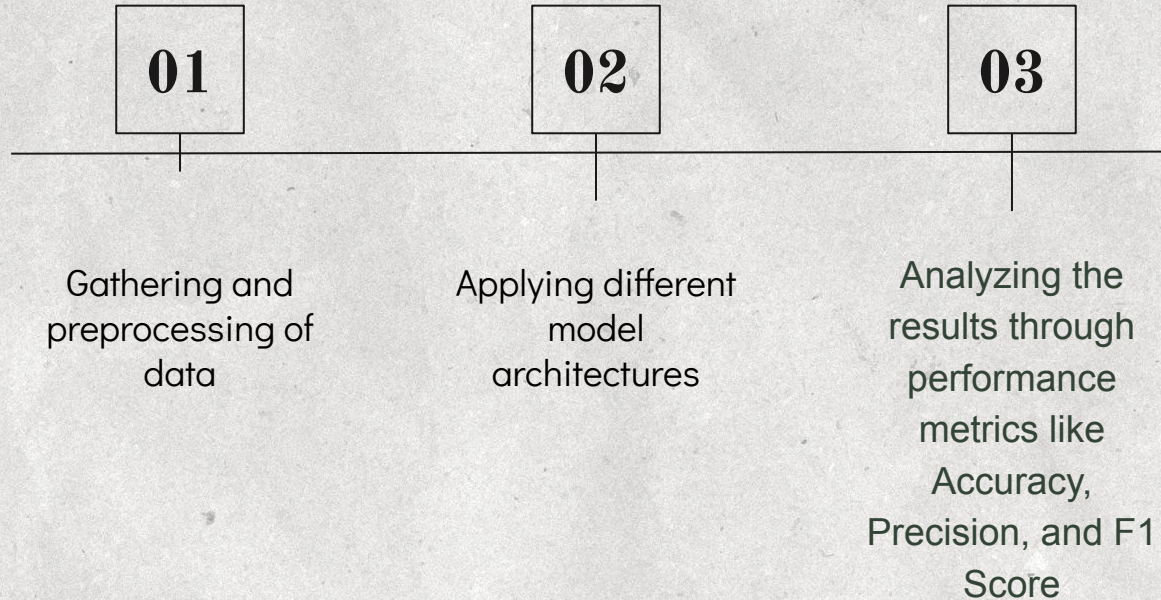


Figure 4 : Basic MLP architecture



Workflow





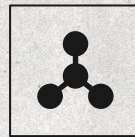
Implementation



Gathered Dataset
and done sampling
and scaling



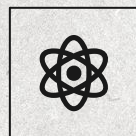
Converted malware
binaries into
grayscale image



Applied CNN for
classification



Applied PCA and
SPCA for
dimensionality
reduction



Applied Multi Layer
Perceptron for
classification

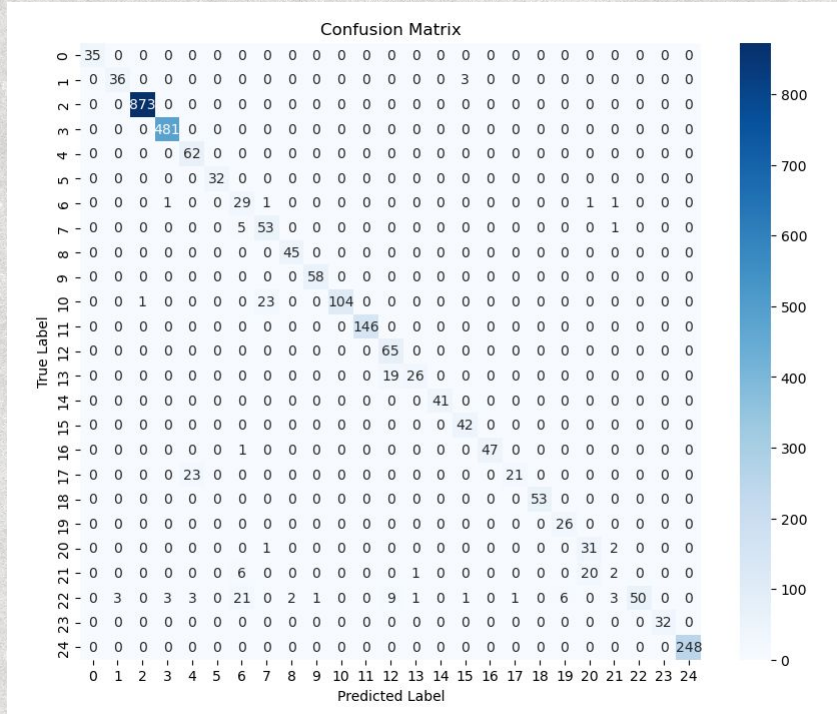


Implementation Cont.



Figure 5 : Grayscale images of
malware binaries

Result Analysis (CNN)



Accuracy :
97.57%



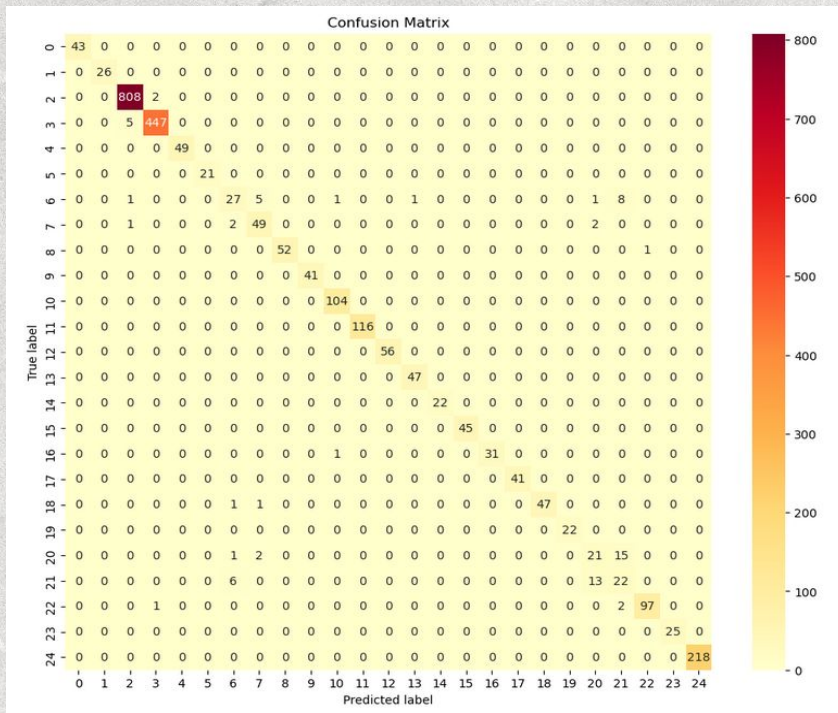
Precision :
95.68%



F1 Score :
95.68%

Figure 6 : Confusion Matrix using CNN model

Result Analysis (PCA-MLP)



Components:
30



Accuracy :
97.13%



Precision :
94.23%



F1 Score :
94.00%

Figure 7 : Confusion Matrix using PCA-MLP model

Result Analysis (PCA-MLP)

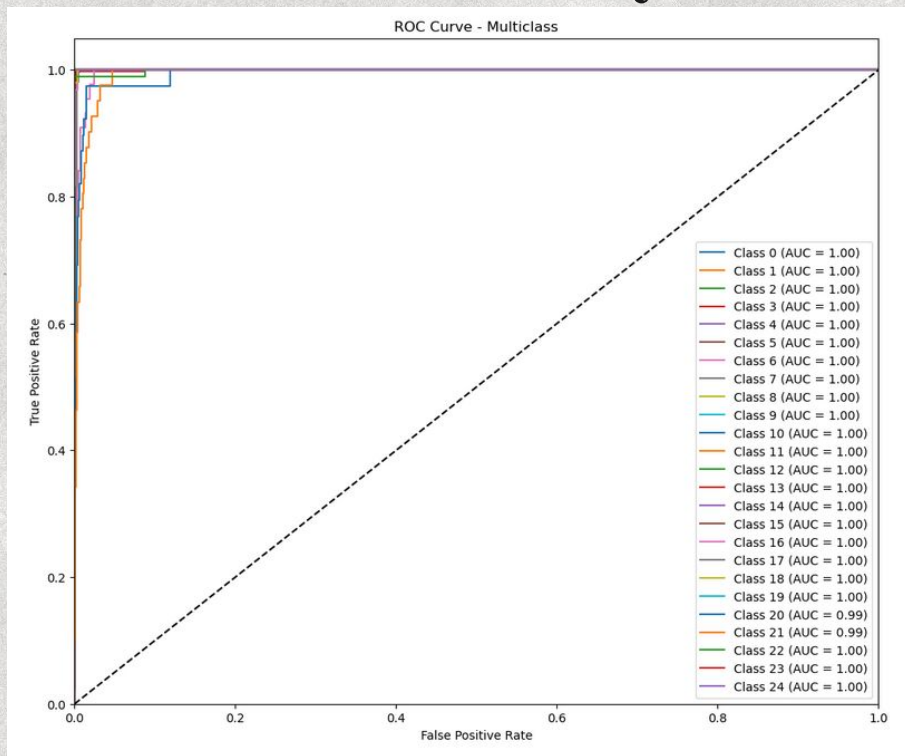


Figure 8 : ROC Curve for PCA-MLP model



Components:
30



Accuracy :
97.13%



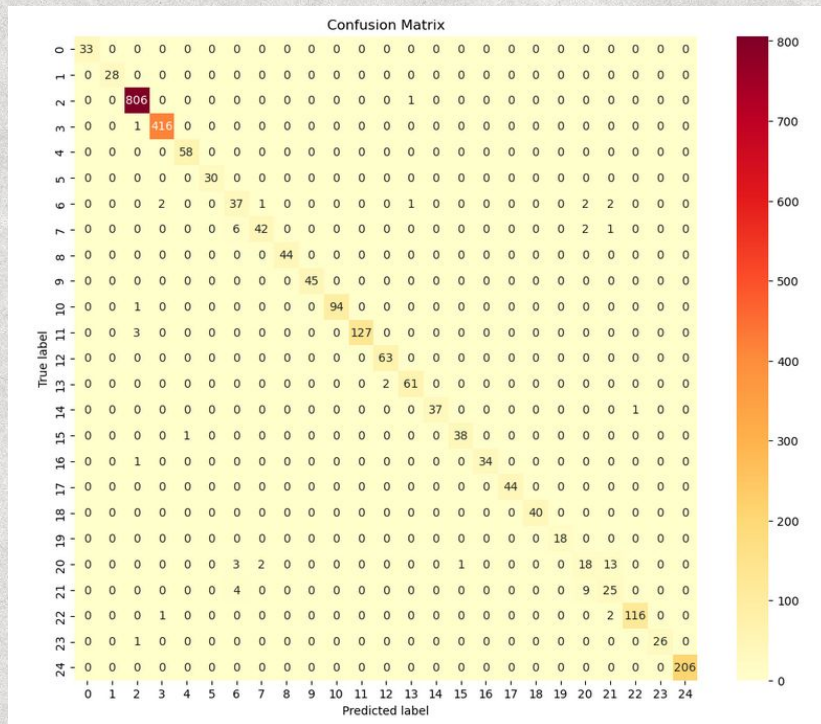
Precision :
94.23%



F1 Score :
94.00%



Result Analysis (SPCA-MLP)



Components:
30



Accuracy :
97.49%



Precision :
94.82%



F1 Score :
94.51%

Figure 9 : Confusion Matrix using SPCA-MLP model

Result Analysis (SPCA-MLP)

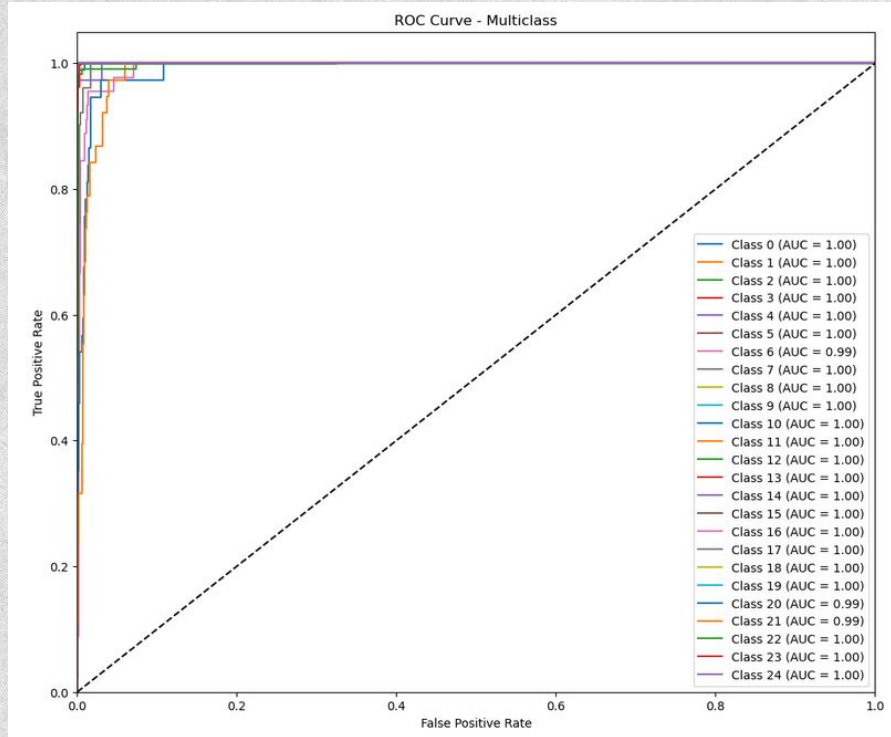


Figure 10 : Confusion Matrix using SPCA-MLP model



Components:
30



Accuracy :
97.49%



Precision :
94.82%



F1 Score :
94.51%



Model Comparison

Model	Accuracy	Recall	Precision	F1 Score
CNN	97.57%	95.68%	95.68%	95.68%
PCA-MLP	97.13%	94.69%	94.23%	94.00%
SPCA-MLP	97.49%	95.44%	94.82%	94.51%

Table 2: Comparison of the models



Future Work



Implement ResNet-152
architecture for
feature extraction



Implement other deep
learning models



Search for better
accuracy and compare
the outcomes



Conclusion

1

Vision-based malware detection can detect malware that has been encrypted to avoid detection by traditional methods

2

Vision-based malware detection can be used to detect new and unknown malware that has not yet been categorized or identified by traditional methods

3

Deep learning approach performs better than traditional methods providing better accuracy and efficiency.



Reference

- [1] Nataraj, L., Karthikeyan, S., Jacob, G., Manjunath, B.S.: Malware images: visualization and automatic classification. In: Proceedings of the 8th International Symposium on Visualization for Cyber Security, VizSec '11, pp. 4:1–4:7. ACM, New York, NY, USA (2011).
- [2] Gibert, D., Mateu, C., Planes, J. *et al.* Using convolutional neural networks for classification of malware represented as images. *J Comput Virol Hack Tech* **15**, 15–28 (2019).
- [3] Agarap, A. F. (2017). Towards building an intelligent anti-malware system: a deep learning approach using support vector machine (SVM) for malware classification. *ArXiv Preprint ArXiv:1801.00318*.
- [4] https://www.dropbox.com/s/ep8qjakfwh1rzk4/malimg_dataset.zip?dl=0

Thanks!

Do you have any questions?

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