

# University of New Haven

Al and CyberSecurity
DSCI6015
Midterm
Project

Jagadeesh Chandra Bose Mende

00869339 Dr. Vahid

Behzadan

# **SUMMARY**

This report outlines the successful development of a cloud-based Portable Executable (PE) static malware detection API using AWS Sagemaker. The API employs a LogisticRegression

classifier trained on a labeled dataset of binary feature vectors to classify PE files as either malicious or benign.

The project utilized AWS Sagemaker for both model construction and training, as well as for deploying the model. Additionally, a user-friendly web application was developed to allow remote users to upload their URL and assess potential threats. Python was the primary language used for the project, with machine learning libraries such as sklearn, pefile, and nltk employed for model creation and implementation.

# **INTRODUCTION**

Executable files, known as PE files, are crucial components of Windows operating systems, serving as a standardized format for storing executable code and associated data. These files contain essential information required for program execution, encompassing machine instructions, resources, imported libraries, and metadata. Widely utilized for applications, drivers, and dynamic link libraries (DLLs), PE files adhere to a structured layout characterized by headers that provide insights into the file's attributes, including its architecture, entry point, and section arrangement. Proficiency in understanding the PE file format is invaluable for various tasks such as software analysis, reverse engineering, and malware detection, facilitating the examination and modification of executable content.

### **APPROACH SUMMARY:**

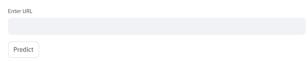
**Model Building and Training**: Utilizing an instance of scikit-learn version 1.2.1 on AWS Sagemaker, we trained a LogisticRegression using a meticulously labeled dataset of binary feature vectors. Subsequently, the trained model was saved into a joblib file.

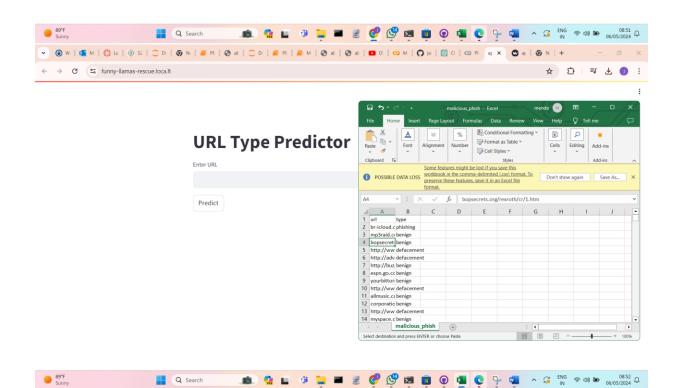
**Model Deployment as a Cloud API**: Leveraging Amazon Sagemaker, we deployed the trained model to create an endpoint, thereby establishing a cloudbased API for real-time predictions. The model was loaded from the saved joblib file and deployed using the sagemaker.SKLearn module. This involved configuring and creating an endpoint to host the model.

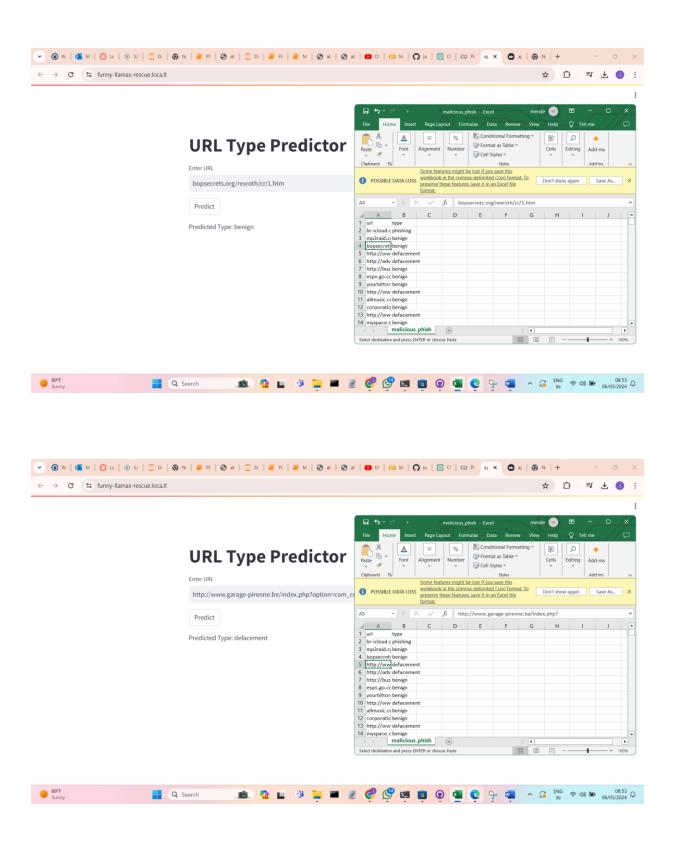
Client Application Development: A Streamlit web application was developed to furnish users with an intuitive interface. Through this application, users can upload executable URL. The application, utilizing pefile library and other pretrained data, extracts requisite features from the uploaded .URL, converts them into JSON format, and dispatches them to the deployed API. The application then showcases the classification results, discerning between Malware (Danger) and Benign (Safe) files. For client deployment, Google Colab was utilized to initiate the Streamlit application.



## **URL Type Predictor**







### **RESULTS**

The project has effectively delivered its desired results:

Developed a proficient Malware detection model that can classify PE files as either malicious or benign after thorough training.

Deployed the trained model on Amazon Sagemaker, establishing a real-time prediction API accessible via the internet.

Designed a web-based interface for end-users to upload files and verify their malicious status.

### **CONCLUSION**

The obtained accuracy of 99.58% (0.99582770439428) suggests that the model accurately predicts the type of exe file most of the time, achieving a high level of correctness in its predictions.