MICHIGAN STATE UNIVERSITY

Project Plan Defeating Malware Payload Obfuscation

The Capstone Experience

Team Proofpoint

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Functional Specifications

- Create a machine learning system to classify files as malicious or benign
 - Accuracy goal: have at least the same accuracy as sandbox detonation
 - Performance goal: be at least 50% faster than detonation in Cuckoo
- Display information in web dashboard
 - High level system information
 - Ability to look at details for individual files

Design Specifications

System Overview

- Files will be placed in the queue by Proofpoint's process
- Extract file metadata and feed that into Machine Learning algorithm
- Machine Learning algorithm will classify the file as benign or malicious

Web Application

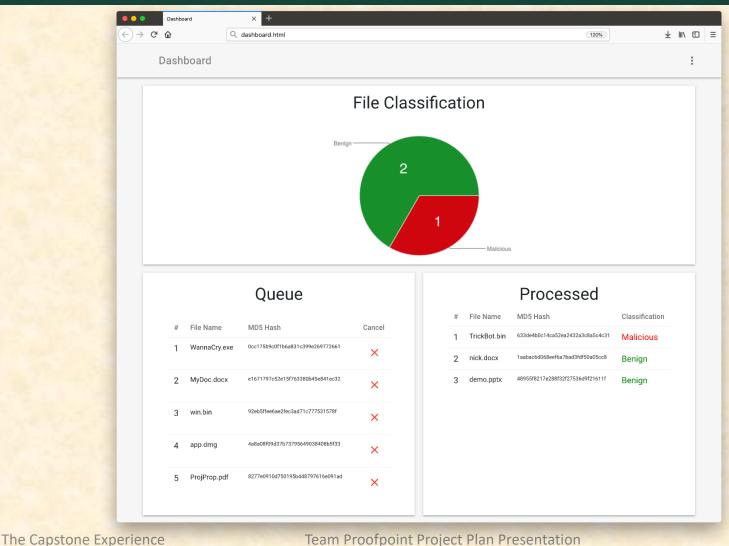
- System Overview
- Display detailed file information
- System Health

Machine Learning Framework

- Train a Machine Learning algorithm to accurately detect malicious files
- Determine characteristics of files that point to malicious behavior
- Malware that can't be classified will be detonated in Cuckoo

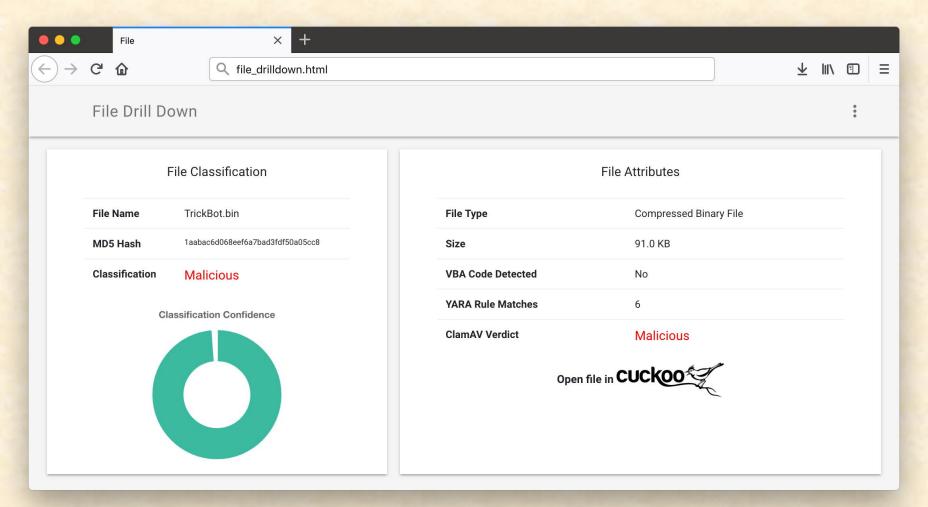


Screen Mockup: Job Pipeline

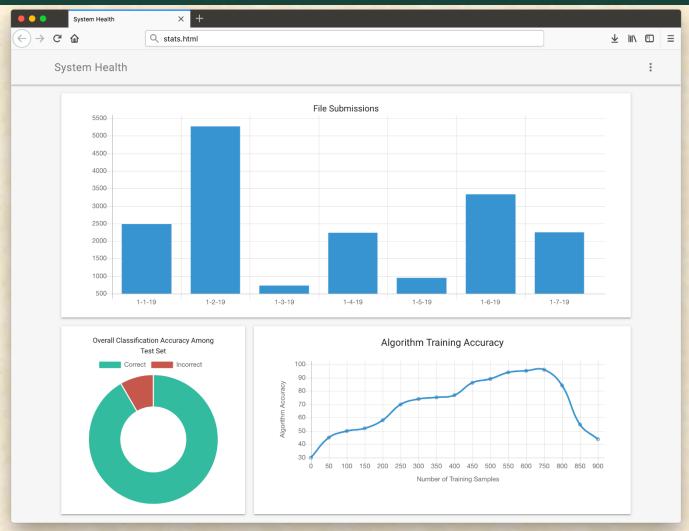




Screen Mockup: File Drill Down

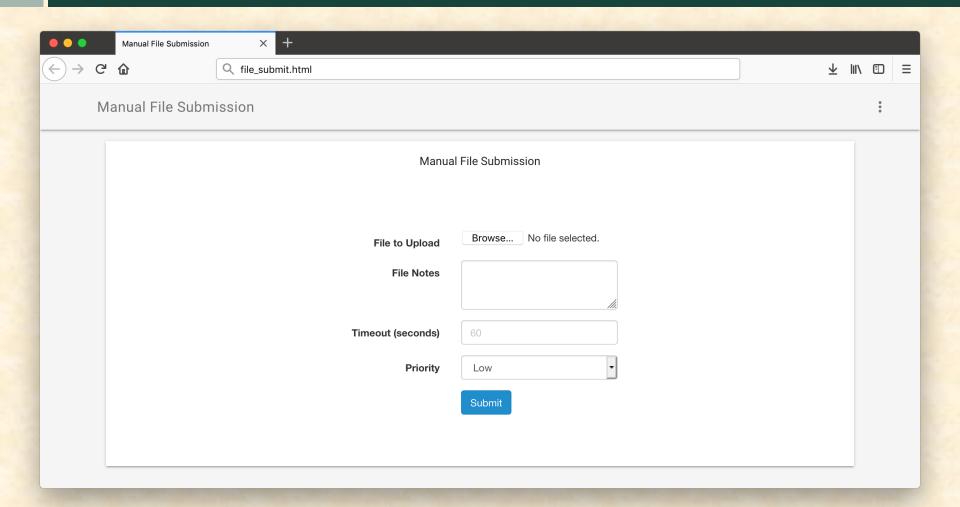


Screen Mockup: System Health





Screen Mockup: Manual File Submission



Technical Specifications

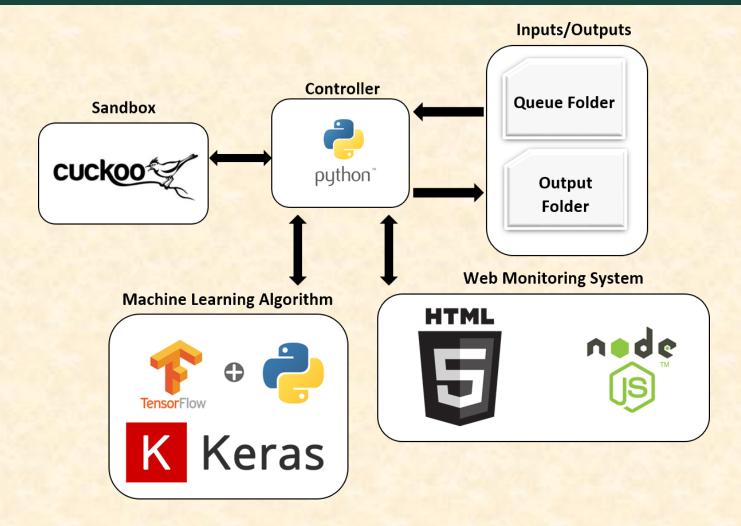
Backend Analysis

- Python controller is used to collect and determine file types and other characters
- Yara and pefile for initial file attribute collection on file samples
- TensorFlow and Keras are used as Machine Learning Algorithms
- Cuckoo is used if Machine Learning Algorithms cannot find definitive classification

Frontend

- HTML5 and Bootstrap CSS
- JavaScript and Node.js

System Architecture



System Components

- Hardware Platforms
 - Proofpoint VMware ESXI server hypervisor
 - Ubuntu virtual machines
- Software Platforms / Technologies
 - VS Code, PyCharm
 - Python 3
 - Cuckoo
 - TensorFlow + Keras

Risks

Feasibility of using Machine Learning to analyze categories of malware

- It is not known if it is possible to encompass all different types of malware using a single machine learning algorithm.
- Mitigation: Try to have our feature extraction be as modular as possible so that when it feeds into the machine learning algorithm, the algorithm does not need to worry about different file types.

Determining what file characteristics can classify malware

- For the ML algorithm to learn, it must be fed many files and analyze the characteristics of those files to learn what is malware. It is difficult to determine what characteristics can be used to detect malware.
- Mitigation: Trial and error research into what kind of characteristics are consistent across different malware files and attempting to detect them with those characteristics before training the algorithm on it.

Identifying steganography

- The malware our project is concerned with is hidden within various payload files. It is not known
 if this can be detected without a full detonation.
- Mitigation: Measure entropy to determine encryption. Also check for hidden values in least significant bits of RGB values in pictures.

Meeting target performance requirements

- For our project to be successful, it must be able to detect and classify malware at a faster rate than sandbox detonation but with just as much accuracy.
- Mitigation: Design our algorithms with the best practices in mind and strive to have high efficiency and accuracy.



Questions?

