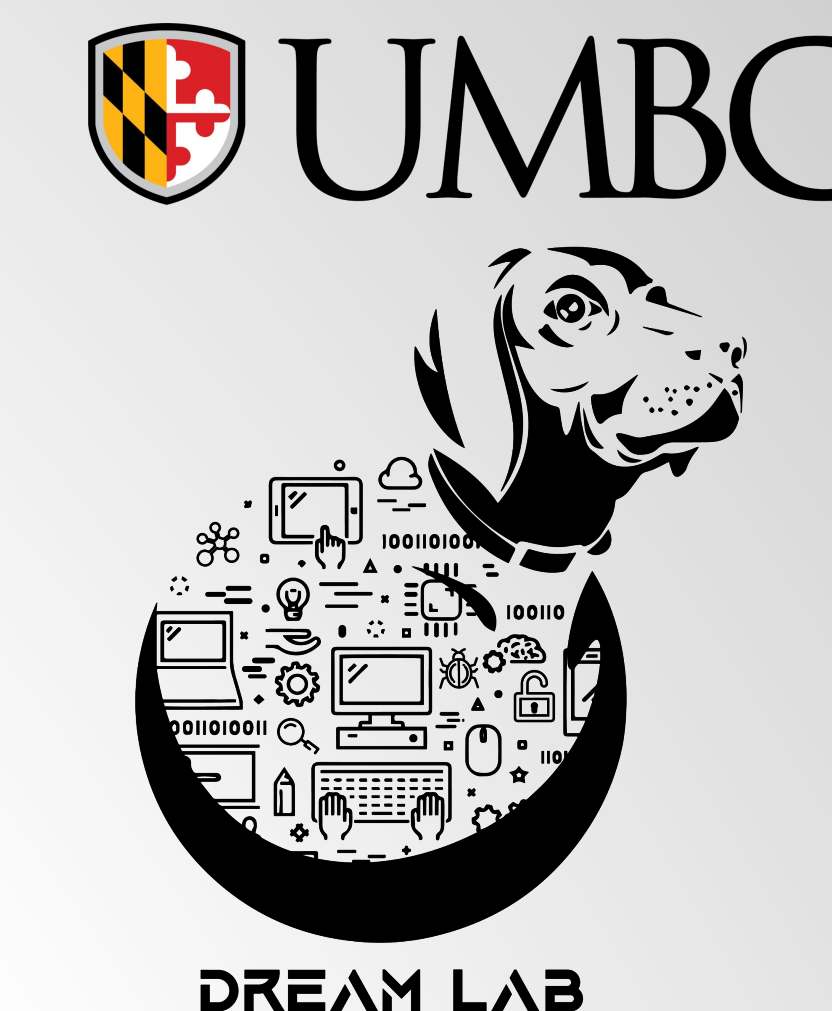


Malware Technical Exchange Meeting

Trends in Malware

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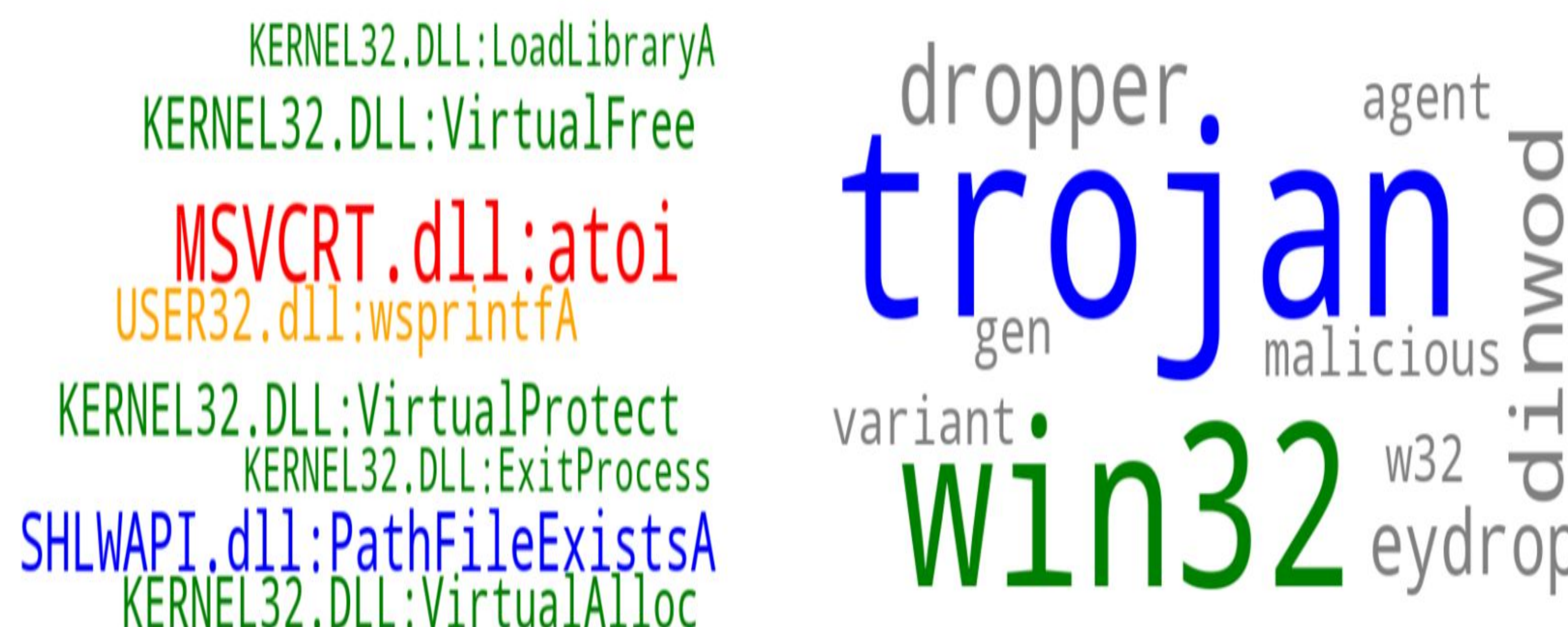
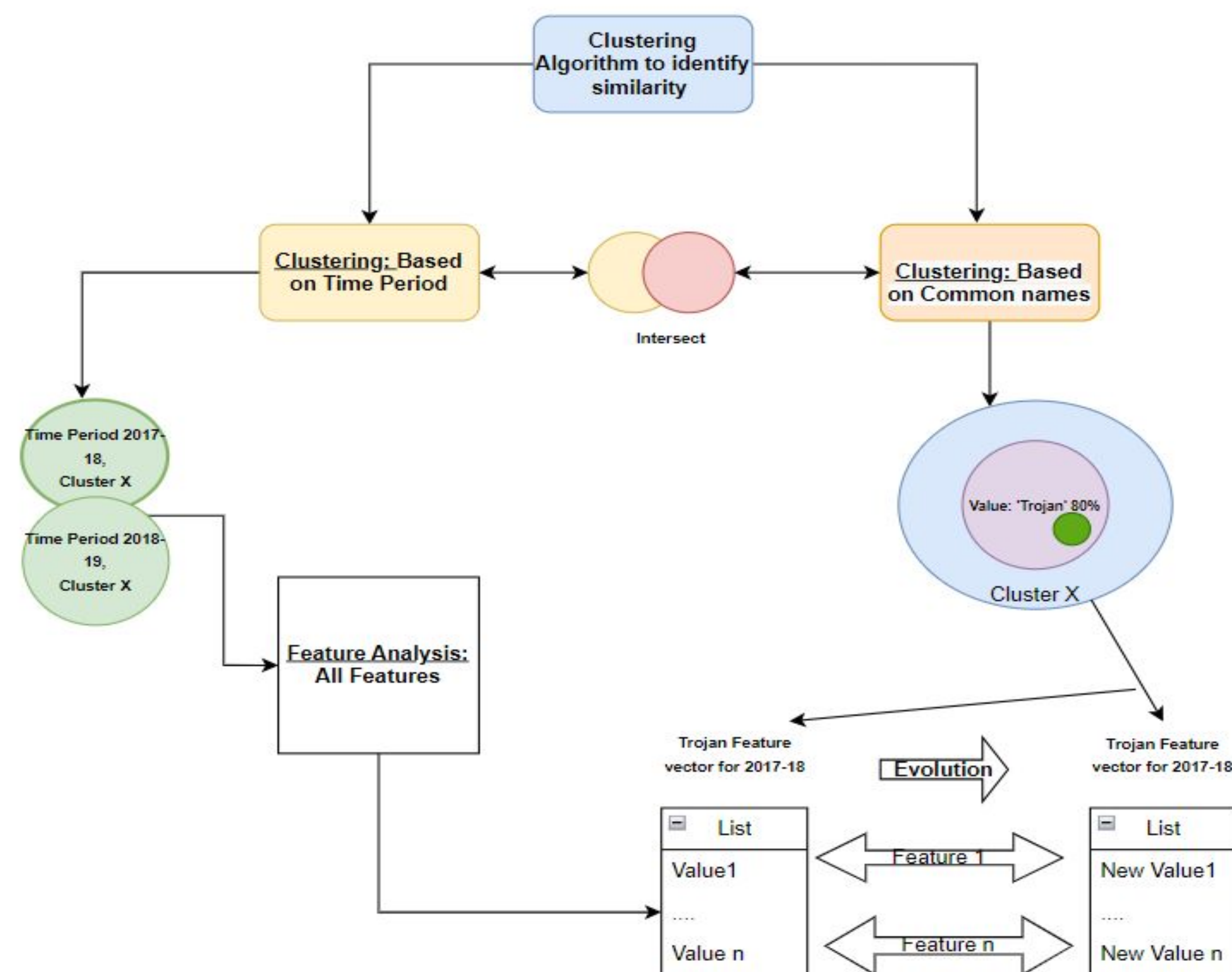
Objectives

- Understanding trends in malware is necessary to identify variants in malware.
- Malware analysts need to be aware of the extent of updates of a malware families.
- Our work classifies samples in the SOREL-20M^[1] dataset using clustering and automatically generate a timeline showing evolution of malware

Methodology

- Parse SOREL-20M dataset
 - AV tokens are not standardized
 - Many variations of the same word e.g. “trojan”/“trjen”/“tr”
- Apply binary vectorization - more comprehensible
- Density-based clustering (DBC) with Euclidean distance in 3 phases:
 - Phase 1: Data Cleaning: Identify the features based on potential information gain
 - Phase 2: Cluster based on similarity
 - Phase 3: Cluster based on Time Period
- Identify common feature values of 80%+ homogeneity within cluster convergences
- Organize the results

Clustering Algorithm



Results

- ~70 malware clusters were found
- Grouped into three major time periods:
 - Late 2016 to mid 2017
 - Mid 2017 to the early 2019
 - Early 2019 to late 2019
- Most clusters have emergent names of ‘trojan’, ‘worm’, ‘win32’
- Clustered at 100% homogeneity-
 - grouped trojan and its variants - ‘Eldorado’, ‘behaveslike’, ‘trojandropper’
 - Function names - “MSVCRT.dll: ‘atoi’”, “KERNEL32.dll: ‘virtualFree’, ‘ExitProcess’, ‘VirtualProjectA’, ‘LoadLibraryA’, ‘VirtualAlloc’, ‘GetProcAddress’”, “SHLWAPI.dll: ‘PathFileExists’”, “USER.dll: ‘wsprintfA’”

Future Work

- Evaluate and compare other clustering algorithms and distance metrics and feature engineering techniques
- Improve tokenization algorithms
- Try using co-clustering algorithm

References

1. Richard Harang and Ethan M. Rudd. Sorel-20m: A large scale benchmark dataset for malicious PE detection, 2020. arXiv.org