

Information Security Centre of Excellence

SMS Mobile Botnet Detection Framework

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Agent in Service Provider 🗲

Detection Module

Defence Module



strategies.

Design of SMS Botnet Detection Module

SMS Botnet Defence Module

Service Provider



 \checkmark Focusing on incoming and outgoing SMS messages.

✓ Real-time content-based signature detection.



> SMS Profiles Collection:

 \checkmark Responsible for collecting, combining, storing and retrieving data to perform anomaly detection.

SMS Clustering:

- \checkmark Provides a rational summary of the collected data in terms of text-clusters.
- \checkmark Takes a set of data and then groups it based on the similarities.
- \checkmark X-means clustering: (Based on K-means -Find the number of clusters dynamically.)
- \checkmark Analyze the result of clusters and group them into four class labels.



Uses the output received from a detection module make logical decisions

Signature Generation

- Signatures that are representative of attack patterns.
- ✓ Utilize Content-based approach:
 - Very fast and robust algorithm.
 - Create automatic signatures of SMS. Ο

Phone number Blacklist (PNBL).

A list of phone numbers that the SMS botnet \checkmark

- ✓ Using two approaches:
 - Pattern-matching approaches.
 - Rule-based techniques.
- ✓This approach was evaluated through the use of 50,000 text messages.

Signature Detection Experimental Results

Types	Features	# of SMS	Total	Percentage
	SMS body	0		
	Phones#	0		
Malicious	URLs	26	165	0.5%
	Commands	139		
	From Phone #	0		
	ToPhone#	0		
	Phones#	869		
Suspicious	URLs	144	3115	5.5%
	Commands	2182		
Normal			51721	94%



2: Precision and Recall Comparison for The First Experimen

>SMS Classification:

- \checkmark The clustering technique can increase the classification accuracy of detection
- ✓Apply machine-learning algorithm to classify the SMS messages SMS to one of the four class labels list:
 - The TF-IDF is a statistical-based approach.
 - Similarity Measurement.

> Profiles Analysis:

- \checkmark Used to look for evidence of compromise rather than any specific attack.
- \checkmark Profile Aggregation: takes into account the similarity between particular profile features.
- \checkmark Profile Prioritization : prioritize each prole based on the following two features:
 - Dangerous permissions
 - User connectivity time.

>SMS Correlation:

- ✓ Identify the relationship between the outputs of the profiles and each detected SMS message.
- ✓ Apply rule-based correlation approach:
- The set of rules helps to produce fewer false positive alarms.
- It also has the ability to label unknown attack

detection app should block.

Malicious Application Analysis

- Analyzing reported apps and extracting their features is therefor a strong method defence against SMS botnets.
- Security administrator can perform static and dynamic analysis using common tools.

Response Action

 \checkmark To take down SMS bots and cutting the C&C channel, it requires the Android user to carry out action by removing the malicious application.

Detection Performance Experimental Results of

two dataset

Detection Metric	A.1 experiment	A.2 experiment
Accuracy	0.968	0.965
Precision	0.986	0.982
Recall(TPR)	0.978	0.978
FNR	0.022	0.022
TNR	0.909	0.888
FPR	0.091	0.112
F-measure	0.982	0.980