

SODA

A System for Cyber Deception Orchestration and Automation

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Demo can be found here: https://youtu.be/Xv8PkUJ72d0

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Agenda

- Motivation
- System Overview
 - Deception playbook creation
 - Real time orchestration
- Evaluations
- Conclusion
- Q&A





Motivation

Malware attacks leverage the asymmetry in cyber warfare

• **Role:** Adversaries can inflict harm on systems, but defenders are limited on defending them.

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• **Discovery**: enabling the malware to progress in order to learn the motive/intention/goal of the attack





Motivation

Understanding malware behavior on the runtime.

- **MSGs:** Finding out Malicious Subgraphs (MSGs) responsible for malicious behaviors.
- **MSG-to-MITRE mapping:** Mapping MSGs to the MITRE ATT&CK framework to determine the malware's behaviors at the kill chain tactical level which helps selecting correct deception actions.

Automatic orchestration

- Manual: Existing deception approaches are manual and lack in agility and automation.
- Customization: Existing deception approaches (example: DodgeTron*) are mostly rule based and do not give the users option to customize them accordingly to their need. SODA provides an automated customizable deception playbooks against arbitrary malware.





Contributions

- We propose a dynamic security orchestration, automation, and deception system, SODA, enabling users to orchestrate deception ploys with appropriate strategies and goals dynamically.
- We propose an automated MSG extraction and MSG-to-MITRE mapping, allowing SODA to understand malware behaviors at the run time to activate relevant deception ploys.
- We propose an embedded deception technique based on API hooking, allowing SODA to execute deception ploys in real-time.
- We evaluated SODA with recent malware to determine the accuracy and the scalability of our approach. We
 observed an accuracy of 95% in deceiving malware with negligible overhead and deployment time. Furthermore,
 our approach successfully extracted MSGs with a 97% recall value and MSG-to-MITRE achieved a top-1
 accuracy of 88.75%.





Screenshots (Scenario 1: Regular attack without SODA)



Malware – Attacker's side

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COLLEGE OF COMPUTING AND INFORMATICS Malware – Victim's side

Attacker is trying to identify the "current working directory"



Screenshots (Scenario 1: With SODA)



Malware – Attacker's side after SODA in action





Screenshots (Scenario 2: Regular attack without SODA)



Malware – Attacker's side

Malware – Victim's side

Attacker is trying to "steal passwords" from well known files



Screenshots (Scenario 2: With SODA)

We basicRAT_server.py - Shortcut	
sajid@bankofamerica.com ihavenomoney! cat completed.	*
[192.168.56.104:54408] basicRAT> Connected by ('192.1 client 192.168.56.104:54443 Client 192.168.56.104:54443 selected.	168.56.104', 54443>
[192.168.56.104:54443] basicRAT> cat password.txt Running cat msajid@uncc.edu 801019270 sajid@facebook.com facebookisfun001 sajid@bankofamerica.com ihavenomoney! cat completed.	Response without SODA
[192.168.56.104:54443] basicRAT> Connected by ('192.1 client 192.168.56.104:54449 Client 192.168.56.104:54449 selected.	168.56.104', 54449>
[192.168.56.104:54449] basicRAT> cat password.txt Running cat msajid@uncc.edu 123456654321 sajid@facebook.com honeyfacebookpass:> sajid@bankofamerica.com iamrich:p cat completed.	Response with SODA
[192.168.56.104:54449] basicRAT>	*

Malware – Attacker's side after SODA in action





SODA: DECEPTION PLAYBOOK CREATION

- Extract Malicious subgraphs (MSGs)
 - A malicious sub-graph (MSG) represents a sequence of WinAPI calls that work together to perform a malicious task.
- Map MSGs to MITRE/Malware behaviors to understand malware goals.
- Create deception ploys based on deception 4D goals and strategies
- Create deception playbook profiles

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- Each profile incorporates ploys for the behaviors that are likely to happen together
- Implement deception ploys inside API hook functions



MSG example: (Steal from the browsers)

CreateFile – GetFileSize – VirtualAlloc – ReadFile – CryptUnprotectData – CreateFile – WriteFile - CloseHandle

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MSG Extraction

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RegOpenKey {'key handle': '0x000000bc', 'regkey': 'HKEY LOCAL MACHINE\\SOFTWARE\\Mozilla\\Mozilla Firefox'} RegQueryValue {'key_handle': 0x000000bc, 'value': '41.0.2 (x86 en-US)', 'regkey': 'HKEY_LOCAL_MACHINE\\SOFTWARE\\Wow6432Node\\Mozilla\\Mozilla Firefox\\CurrentVersion'} 3 RegCloseKey, 0, {'key handle': '0x00000bc'} 4 5 CreateToolhelp32Snapshot{'Out Handle': '0x000000f8'} 6 Process32First {'process name': '[System Process]', 'snapshot handle': 0x000000f8 7 Process32Next{'process name': 'System', 'snapshot handle': 0x000000f8}} 8 Process32Next ... 9 10 FindFirstFile {'filepath': 'C:\\Users\\Administrator\\AppData\\Local\\Google\\Chrome\\User Data\\Default\\Login Data'} 11 CreateFile {'out file handle': 'Ox000000dc', 'create disposition': 1, 'filepath': 'C:\\Users\\Administrator\\AppData\\Local\\Google\\Chrome\\User Data\\Default\\Login Data'} 12 GetFileType {'file handle': '0x00000dc 13 ReadFile {'file_handle': '0x000000dc, 'buffer':'FileContent', 'length': 260, 'offset': 0} 14 SetFilePointer {'file handle': 0x000000dc, 'move method': 0, 'offset': 4096} 15 ReadFile {'file_handle': '0x000000dc', 'buffer':'FileContent', 'length': 4096, 'offset': 0} 16 CloseHandle { 'handle': 0x000000dc' 17

Malware execution trace log



MSGs extracted from the trace above



SODA: DECEPTION PLAYBOOK CREATION

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How MSG-to-MITRE Mapping Works





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Deception ploys creation

Malware	Manned MSC (API Sequence)	Stratogy	De	cepti	on G	oal	Decention Actions
Behavior	Mapped MSG (API Sequence)	Strategy	D_1	D_2	D_3	D_4	Deception Actions
	1) Search file and steal: FindFirstFile, PathFileExist, CreateFile, CatFileSize, VirtualAlloc, ReadFile, CloseHandle, VirtualFree	FakeFaliure	1	-	-	_	Diversion: pretend the File doesn't exist by
Stealing from		ruterunure	· ·				returning false when PathFileExist is called
credentials files	FindNevtFile FindClose	(m)		-	- 1		 Depletion: replace sensitive file reading with
creaentials mes	 2) Read sensitive file (known file). CreateFile, ReadFile 3) Steal from the browsers. CreateFile, GetFileSize, VirtualAlloc, ReadFile, CryptUnprotectData, CreateFile, WriteFile, CloseHandle 	FakeSuccess	-			1	static HoneyFile containing Honey Credentials.
							2) Discovery: watch out for Exfiltration.
							1) Diversion: forward the execution to HoneyFactory
							(false target).
					1		2) Depletion: communicate with HoneyFactory. Ask
			-	-			for HoneyFile containing Honey Credentials. Replace
							sensitive data with the content of HoneyFile.
		NativeExecute	-	-	-	1	Discovery: watch out for Exfiltration

Table 1: Deception ploy creation and verification (D_1 = diversion, D_2 = distortion, D_3 = depletion, D_4 = discovery).





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Deception playbook profile creation

- We perform frequent item-set mining to identify highly associated MSGs.
- For example, if deception ploys *T*1,*T*2,*T*3, ...,*T*6 are created for malicious behaviors *B*1,*B*2,*B*3, ...,*B*6, respectively. If *B*1,*B*2 and *B*3 are in a frequent itemset then we create profile *P*1 containing *T*1,*T*2 and *T*3.
- The target is to provide a generic profile for each malware type such as InfoStealer, Ransomware, Spyware etc.
- However, user can create their own profile based on requirement.

 $Support(B) = \frac{Number \ of \ malware \ in \ which \ B \ appears}{Total \ number \ of \ Malware \ in \ the \ dataset}$ $Confidence(B1 \rightarrow B2) = \frac{Support(B1 \cup B2)}{Support(B1)}$





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Deception factory creation



B. With API Hooking

Figure 12: Call flow with and without API Hooking. In some case, Response is return via the original API (5a, 5b), otherwise, Detour function responds directly (5).

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Figure 7: A code snippet: How the deception technique is implemented inside API hooks



SODA: REAL-TIME ORCHESTRATION

- Playbook profile creation using a user interface
 - Requests are handled using REST APIs
 - Profile information is sent to Orchestration Engine Server (OES)
 - OES responses with End-point DLL and a configuration file
- Detection agent:
 - Detects the malware
 - Injects the End-point DLL into the malware
- Finally, Realtime deception takes place
 - Deception actions are implemented within embedded API-Hookings
 - Which deception actions to take is determined from the configuration file





Screenshots

HoneyFactory IP address: 192.168.56.102								
	<u><u> </u></u>	Goals						
Malware Benavior	Strategy	Diversion	Distortion	Depletion	Discovery			
	FakeFailure	0						
Pomoto ovocution	FakeSuccess		0		0			
Remote execution	Honey	0			0			
	Allow							
	FakeFailure	0						
Screen Capture	FakeSuccess		0		0			
	Honey	0			0			
	Allow				۲			
eylogging ata collection from Clipboard	FakeFailure	0						
	FakeSuccess		0		0			
	Honey	0			0			
	Allow				۲			
	FakeFailure	0						
Data antian from Clinhoord	FakeSuccess		0		0			
Data conection from Chpboard	Honey	0			0			
	Allow				۲			
	FakeFailure	0						
Staaling from anodontials filos	FakeSuccess		0		0			
Stearing from credentials files	Honey	0			0			
	Allow				۲			
	FakeFailure	0						
	FakeSuccess		0		0			
Read remote files	Honey	0			0			
	Allow							

User Interface for Deception Playbook Profile creation

Submit





Screenshots (With SODA)



End-Point DLL is injected





Screenshots (Scenario: With SODA)

- Behavior: List the current working directory
- **4D goal:** Diversion
- Strategy: FakeSuccess



Malware – Attacker's side after SODA in action





Evaluations

Recall of MSG extraction

- > Ground-Truth (GT1):
 - We downloaded 42 malware source code from the GitHub along with the comments and descriptions explaining the malware's capabilities/behaviors.
 - Manually extracted 94 distinct MSGs.
- > Experiment:
 - We obtain the binaries of these malware by building the source code
 - We run them in the API tracer (Our analyzer/Extended Cuckoo Sandbox)
 - Our MSG extractor was able to detected 113 unique MSGs.
 - We manually verified out of these 113, 91 of them are as expected (belonging to the GT1)

$$Recall_{GT1} = \frac{TP}{TP + FN} = \frac{91}{91 + 3} = 0.968$$





Evaluations

Comparison with other State-of-the-art tools and Sandboxes

Tools	Malware Behavior/Capability Identification (Using GT1)							
	Identified	Not Identified	Recall					
Kris et al. [32]	58	36	0.62					
FORECAST [11]	32	62	0.34					
DodgeTron [37]	8	86	0.09					
SODA	91	3	0.97					

Comparison with other State-of-the-art tools in terms of discovering malware behaviors/capabilities using GT1 and their individual recall values

Family	Malware Family	Discovery	Cuckoo	Any.run	SODA
	Foreit	Т	8	7	8
	raren	Р	39	126	149
	LokiBot	Т	7	2	11
InfoStoolor		Р	21	173	243
infostealer	Donu	Т	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	17	
	Fony	Р	231	191	582
	Pacoon	Т	7	8	16
	Kacoon	T 7 P 45 T 6 P 27	45	23	51
	Drash	Т	6	3	6
Domocrystere	Куик	Р	27	32	102
Kansoniware	CandCrah	Т	8	ackooAnyrunSODA87839126149721121173243841723119158278164523516362732102810101921092452264576910121012521314416	
	Ganucrab	Р	192	109	245
	Chlot	Т	2	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	
	Gliost	Р	4	57	69
PAT	VanilaRat	Т	1	0	12
I INAI	vaimanat	Р	1	0	12
	Queeer	Т	5	2	13
	Quasar	Р	14	4	16

Number of techniques (T) and procedures (P) discovered by SODA compared to Cuckoo sandbox and Any.run.



Evaluations

MSG Classifier Evaluation (MSG-to-MITRE)

- > Ground-Truth (GT2):
 - We used Remote Access Trojans (RATs) to create this ground truth.
 - We downloaded 13 different RATs from GitHub, capable of performing 33 distinct malicious behaviors.
 - We ran each malicious behavior at a time and manually collected MSGs
 - We manually mapped each of these MSGs to MITRE
 - Summary: we obtained 80 MSGs (correspond to 33 distinct malicious behaviors) and mapped them manually to 31 MITRE techniques.
- > Experiment:
 - $_{\odot}$ $\,$ We feed these MSG to the MSG classifier.
 - MSG-to-MITRE mapping achieved a top-1 accuracy of 88.75%.

Minimum word frequency	4
API TF-IDF enriching threshold	20%
Word2Vec similarity threshold	70%
Maximum number of words per MITRE technique	40

Table 4: MSG Classifier Parameters



Experiment	Excluding Stackoverflow	Excluding Enriching	After Result Analysis			
Top-1 Accuracy	48.75%	58.75%	88.75%			
Top-2 Accuracy	62.5%	73.75%	96.25%			

Table 6: Top-n accuracies after analysis as well as excluding StackOverflow and enriching component.





Evaluations - Performance Analysis of SODA

Deployment Time:

- > Deployment consists the following three aspects:
 - > generating the configuration file.
 - preparing necessary HF and
 - > Forwarding the End-Point DLL and the configuration file to the OEC.
- > Experimental setup and result:
 - > User creates deception playbook profile with 50 deception ploys.
 - We performed the experiment 5 times by varying the ploys.
 - > The maximum deployment time recorded is 72 sec.

Scalability Measurement:

- Simultaneous request from multiple clients
- > Experimental setup and result:
 - > We used the same profile to keep the consistency.
 - > We performed the experiment using 2-10 clients.
 - > Note: The execution time of the malware is 127 sec without SODA.
 - From this experimental result, we concluded that even though the orchestration time increased, OES still was able to serve its service successfully with negligible overhead (maximum of 7s) compared to the entire execution time of the malware (127s).



SODA deployment time with different ploys



Deployment time Malware response time Total orchestration time

Avg time of a single OES to orchestrate and serve multiple OECs





Evaluations - Performance Analysis of SODA

Overhead/Response delay time:

- > We find out the overhead/response delay time of SODA by running the malware with and without SODA and the let the malware reach to the same execution point.
- We performed the experiment with four different types of malware (RAT, InfoStealer, Ransomware and Spyware)
- > Our data shows that the maximum overhead time was 18 seconds (14% increment compared to the normal malware execution) which is minimal/insignificant compared to the total running/campaign period of an APT/malware.

Malware Type	Focused Malicious Behavior	Strategy	Deception Goal	Deception Action	Expectation (Observance to consider deception ploy worked)	T1 (sec)	T2 (sec)	0 (%)
RAT	Remote command Execution	FakeExecute	Depletion	Execute the remote command in HF and show it to the malware	Command executed in HIF and is shown to the attacker (C&C server is in our control)	13	15	15%
InfoStealer	Steal credentials from the browsers	FakeExecute	Depletion	Show honey credentials from the HF	Honey credentials is seen to be exfiltrated (using packet capture)	38	43	13%
Ransomware	Encrypt files for Impact	FakeSuccess	Diversion	Pretend the encryption took place without performing it	This malware creates a ransom note after successful encryption (observe the note being created)	126	144	14%
Spyware	Capture screen	FakeExecute	Discovery	Capture screen from the HF and send it to the attacker	Captured screen of the HF is uploaded to our FTP server (redirected using ApateDNS)	61	65	7%

Table 7: Malware deception overhead (T_1 = time without deception, T_2 = time with SODA deception, O = Overhead).



Evaluations – End-to-End Accuracy of SODA

- We find out the E2E accuracy of SODA, we used 6 RATs with 37 distinct malicious behavior. \succ
- Based of different deception strategies and goals, we identified 116 valid deception ploys that can be \geq deployed to deceive these RATs.
- > We observed SODA could deceive the RATs in 107 cases out of those 116 valid deception ploys.

					100.00% - 98.00% - 96.00% -	6.30%	4.08%		3.57%		6.06%
Malware Type	Number of malware	Total number of valid deception ploys	Total number deception ploys where SODA successfully deceives malware	curacv	94.00% - 92.00% -						
RATs	6	116	107	Ac	90.00% -	93 70%	95.92%		96.43%		93 94%
InfoStealers	122	49	47		88.00%	55.7070					53.5470
Ransomware	96	28	27		86.00% -						
Spyware	31	33	31		64.00%						•
Total	255	237	224 (95%)			RATS	InfoSteale Malv	rs R ware	ansomwar Type	e	Spyware

Successfully Deceived Not Deceived

Accuracy of SODA across different malware types



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Conclusion

- We propose an **autonomous cyber deception orchestration** system capable of analyzing real-world malware, discovering attack techniques, constructing Deception Playbooks, and orchestrating the environment to deceive malware.
- SODA advances the state-of-the-art by providing dynamic real-time deception and customization options to the users to choose their own deception ploys.
- Our proposed method of MSG extraction, followed by MSG-to-MITRE mapping, showed a promising result in **bridging the gap** between **malware traces** and the **MITRE ATT&CK framework**.
- Our extracted MSGs and MSG-to-MITRE mapping can play a vital role in **improving** the **existing tools**.
- We conducted rigorous evaluations to validate SODA's efficiency and scalability against 225 recent malware and observed:
 - An accuracy of **95%** in deceiving them.
 - A recall of **97%** in extracting MSGs
 - An accuracy of **88.75%** in mapping MSG-to-MITRE





Q&A



