A SURVEY ON MALWARE DETECTION AND ANALYSIS TOOLS

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ABSTRACT

The huge amounts of data and information that need to be analyzed for possible malicious intent are one of the big and significant challenges that the Web faces today. Malicious software, also referred to as malware developed by attackers, is polymorphic and metamorphic in nature which can modify the code as it spreads. In addition, the diversity and volume of their variants severely undermine the effectiveness of traditional defenses that typically use signature-based techniques and are unable to detect malicious executables previously unknown. Malware family variants share typical patterns of behavior that indicate their origin and purpose. The behavioral trends observed either statically or dynamically can be manipulated by using machine learning techniques to identify and classify unknown malware into their established families. This survey paper gives an overview of the malware detection and analysis techniques and tools.

KEYWORDS

Malware, Detection, Analysis, Tools, Machine Learning.

1. Introduction

Malware is any malicious software used to interrupt machine activity, capture sensitive information, or obtain access to private computer systems. Malware is characterized by its malicious intent, which works against the computer user's specifications, and does not include software which causes unintended harm due to a deficiency. The term badware is sometimes used and applied to both true (malicious) malware and unintentionally harmful software. These are intended to

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gain access to computer systems and network resources, disturb computer operations, and gather personal information without taking the consent of the system's owner, thus creating a menace to the availability of the internet, the integrity of its hosts, and the privacy of its users. Spreading of malware has affected everyday life, from e-governance [1] to social networks [2], from digital automation [3] spreading up to mobile networks [4]. Malware comes in a wide range of variations like Virus, Worm, Trojan-horse, Rootkit, Backdoor, Botnet, Spyware, Adware etc. These classes of malware are not mutually exclusive meaning thereby that a particular malware may reveal the characteristics of multiple classes at the same time. In order to evade detection, malware authors introduce polymorphism to the malicious components. This means that malicious files belonging to the same malware "family", with the same forms of malicious behavior, are constantly modified and/or obfuscated using various tactics, such that they look like many different files. Malware is one of the most terrible and major security threats facing the Internet today. According to a survey [5], conducted by Symantec in February 2019, 47% of the organizations experienced malware security incidents/network breaches in the past one year, as depicted in figures 1 and 2. The malware is continuously growing in volume (growing threat landscape), variety (innovative malicious methods) and velocity (fluidity of threats). These are evolving, becoming more sophisticated and using new ways to target computers and mobile devices. McAfee [6] catalogs over 100,000 new malware samples every day means about 69 new threats every minute or about one threat per second. With the increase in readily available and sophisticated tools, the new generation cyber threats/attacks are becoming more targeted, persistent and unknown. The advanced malware is targeted, unknown, stealthy, personalized and zero-day as compared to the traditional malware which was broad, known, open and one time. Once inside, they hide, replicate and disable host protections. After getting installed, they call their command and control servers for further instructions, which could be to steal data, infect other machines, and allow reconnaissance. Attackers exploit vulnerabilities in web services, browsers, and operating systems, or use social engineering techniques to make users run the malicious code in order to spread malware. Malware authors use obfuscation techniques [7] like dead code insertion, register reassignment, subroutine reordering, instruction substitution, code transposition, and code integration to evade detection by traditional defenses like firewalls, antivirus and gateways which typically use signature based techniques and are unable to detect the previously unseen malicious executables. Commercial antivirus vendors are not able to offer immediate protection for zero-day malware as they need to analyze these to create their signatures. To overcome the limitation of signature-based methods, malware analysis techniques are being followed, which can be either static or dynamic. The malware analysis techniques help the analysts to understand the risks and intentions associated with a malicious code sample. The insight so obtained can be used to react to new trends in malware development or take

preventive measures to cope with the threats coming in the future. Features derived from analysis of malware can be used to group unknown malware and classify them into their existing families.

This paper presents a review of techniques/approaches and tools for detecting and analyzing the malware executables. There has been some study performed on comparison of static, dynamic, and hybrid analysis for malware detection [8], whereas some researchers tried to bridge the static/dynamic gap [9]. Mobile technology in healthcare has also been a target of malware [10]. Few recent studies have been done on static and dynamic analysis of Android malware [11], detection using permission [12–14], based on system call sequences and LSTM [15].

Many studies use static analysis for malware detection using exact decompilation [16], similarity testing framework [17], based on register contents [18], using two-dimensional binary program features [19], subroutine based detection [20], statistics of assembly instructions [21], file relation graphs [22], de-anonymizing programmers via code stylometry [23], based upon a wavelet package technique [24], analysis and comparison of disassemblers for opcode [25].

The studies that use dynamic analysis perform synthesis the semantics of obfuscated code [7], multi-hypothesis testing [26], analyzing quantitative data flow graph metrics [27], using simplified data-dependent api call graph [28], downloader graph analytics [29], access behavior [30, 31], APIs in initial behavior [32], log-based crowdsourcing analysis [33].

There have been many studies on the detection and analysis of malware using machine learning that study fine-grained features [34], deep learning [35–37], dynamic features [38], static features [36, 39], concept drift [40], predicting signatures [41], hybrid framework [42], malware metadata [43], reverse engineering of large datasets of binaries [44].

Our Contributions: This paper presents the following contributions:

- Techniques and tools for detecting and analyzing malware. This paper provides the first comprehensive survey on techniques and tools for detecting and analyzing malware. There have been numerous surveys in the area of malware detection specific to machine learning, android and a few surveys on static and dynamic analysis. However, none of the work addresses the techniques and available tools.
- **State-of-the-art Survey**. This paper reveals that the most existing surveys in this area are either outdated [45] or fail to provide a holistic view of the problem, since they usually focus on a specific subset of the standard [46].

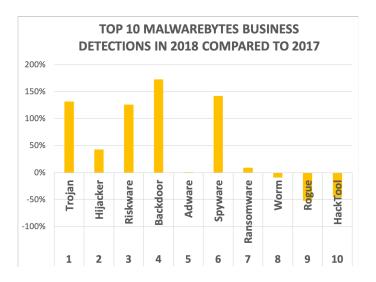


Figure 1. Top 10 malwarebytes business detections in 2018 compared to 2017 [47]

- List of comprehensive tools. This paper presents a novel overview of the list of comprehensive tools available for malware detection, memory forensics, packet analysis, scanners/sandboxes, reverse engineering, debugging, and website analysis. Further, it differentiates the malware analysis tools based on specific domain and approach.
- Guide for malware analysts. Finally, it is realized that the contribution claimed in this paper will help, guide and assist researchers and malware analysts on getting appropriate tools for their domain-specific analysis.

The rest of the paper is organized as follows. Section II describes the different types of malware. Section III describes the ways of malware analysis. Section IV discusses malware analysis tools. Finally, Section V concludes the paper with a highlight on the scope of future work.

2. DIFFERENT TYPES OF MALWARE

With so many different types of malware – and the vast range of malicious software programs within each type – it's important that every malware item can be unambiguously classified and easily distinguished from other malicious programs. The term malware includes viruses, worms, Trojan Horses, rootkits, spyware, keyloggers and more. To get an overview of the difference between all these types of threats and the way they work, it makes sense to divide them into groups:

2.1. Viruses and worms: the contagious threat

Viruses and worms both are designed to spread without the user's knowledge. A computer virus is a small program written to alter the way a computer operates, without the permission or knowledge of the user. A virus must meet two criteria [8]:

- 1. It must execute itself. It will often place its own code in the path of execution of another program.
- 2. It must replicate itself. For example, it may replace other executable files with a copy of the virus-infected file.

Viruses can infect desktop computers and network servers alike. Some viruses are programmed to damage the computer by damaging programs, deleting files, or reformatting the hard disk. Others are not designed to do any damage, but simply to replicate themselves and make their presence known by presenting text, video, and audio messages. Even these benign viruses can create problems for the computer user. They typically take up computer memory used by legitimate programs. As a result, they often cause erratic behavior and can result in system crashes. Also, many viruses are bug-ridden, and these bugs may lead to system crashes and data loss.

Computer worms, on the other hand, spread across the internet by replicating itself on computers via their network. Both viruses and worms can carry a so-called "payload", malicious code designed to do damage. Worms are programs that replicate themselves from system to system without the use of a host file. This is in contrast to viruses, which requires the spreading of an infected host file. Although worms generally exist inside of other files, often Word or Excel documents, there is a difference between how worms and viruses use the host file. Usually the worm will release a document that already has the "worm" macro inside the document. The entire document will travel from computer to computer, so the entire document should be considered the worm. PrettyPark is a particularly prevalent example.

2.2. Trojans and Rootkits: the masked threat

Trojans and rootkits are grouped as they both seek to conceal attacks on computers. Trojan Horses are malignant pieces of software pretending to be benign applications. Users therefore download them thinking they will get a useful piece of software and instead end up with a malware infected computer. A Trojan horse, or Trojan, in computing is a generally non-self-replicating type of

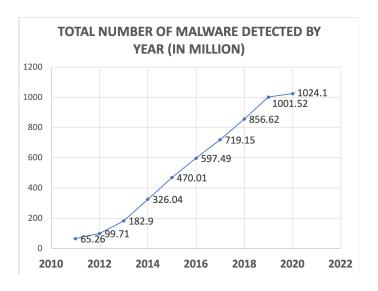


Figure 2. Total number of malware detected by year (in million) [48]

malware program containing malicious code that, when executed, carries out actions determined by the nature of the Trojan, typically causing loss or theft of data, and possible system harm. The term is derived from the story of the wooden horse used to trick defenders of Troy into taking concealed warriors into their city in ancient Anatolia, because computer Trojans often employ a form of social engineering.

Rootkits are different. They are a masking technique for malware, but do not contain damaging software. Rootkit techniques were invented by virus writers to conceal malware, so it could go unnoticed by antivirus detection and removal programs. A rootkit is a stealthy type of software, typically malicious, designed to hide the existence of certain processes or programs from normal methods of detection and enable continued privileged access to a computer. The term rootkit is a concatenation of "root" (the traditional name of the privileged account on Unix operating systems) and the word "kit" (which refers to the software components that implement the tool).

2.3. Spyware and keyloggers: the financial threat

Spyware and keyloggers are malware used in malicious attacks like identity theft, phishing and social engineering, threats designed to steal money from unknowing computer users, businesses and banks. Spyware is a type of malicious software (also called "malware") that scammers try to install on your computer. As the name suggests, spyware programs allow people to spy on what you are doing on your computer: the websites you visit, the files you use and the details you store on your PC.

Key-loggers are a particular type of spyware. Key-loggers secretly record what keys you press on your keyboard and sends this data back to the scammer over the internet. Scammers use these programs to steal passwords such as online banking passwords. They may also use spyware to steal other personal information from you such as documents that you have stored on your computer. Scammers use a wide range of tricks to get their spyware and key-loggers loaded on to your computer. This usually involves tricking you into clicking on a link in a spam email they have sent, or visiting a website that they have set up solely to infect people's computers. Other sources of spyware and key-loggers are free games or music that you can download from the internet. When they are delivered in this way, they are sometimes called "Trojans"—a file that claims to be for some harmless purpose so it can get under your guard, but contains a nasty surprise.

3. MALWARE ANALYSIS

Before creating the signatures for newly arrived malware, these are required to be analyzed to understand the associated risks and intentions. The malicious program and its capabilities can be observed either by examining its code or by executing it in a safe environment.

3.1. Static analysis

Analyzing malicious software without executing it is called static analysis. The detection patterns used in static analysis include string signature, byte-sequence n-grams, syntactic library call, control flow graph and opcode (operational code) frequency distribution etc. The executable has to be unpacked and decrypted before doing static analysis. The disassembler/debugger and memory dumper tools can be used to reverse com pile windows executables. Disassemble/Debugger tools like IDA Pro and OllyDbg displays the malware's code as Intel X86 assembly instructions, which provide a lot of insight into what the malware is doing and provide patterns to identify the attackers. Memory dumper tools like LordPE [9] and OllyDump [10] are used to obtain protected code located in the system's memory and dump it to a file. This is a useful technique to analyze packed executables which are difficult to disassemble. Binary obfuscation techniques, which transform the malware binaries into self-compressed and uniquely structured binary files, are designed to resist reverse engineering and thus make the static analysis very expensive and unreliable. Moreover, when utilizing binary executables (obtained by compiling source code) for static analysis, the information like size of data structures or variables gets lost thereby complicating the malware code analysis [11]. The evolving evasion techniques being used by malware writers to thwart static

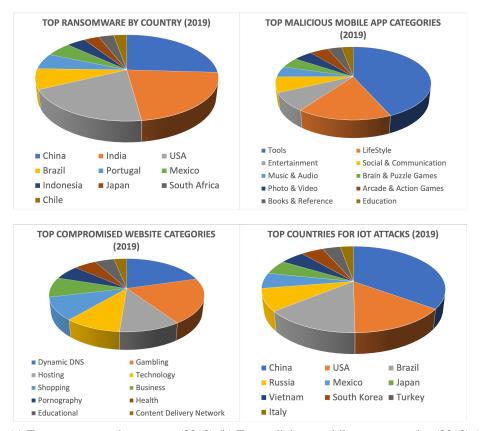


Figure 3. (a) Top ransomware by country (2019) (b) Top malicious mobile app categories (2019) (c) Top compromised website categories (2019) (d) Top countries for IoT attacks (2019) [5]

analysis led to the development of dynamic analysis. Moser et al. [12], explored the drawbacks of static analysis methodology. In their work, they introduced a scheme based on code obfuscation revealing the fact that the static analysis alone is not enough to detect or classify malware. Further, they proposed that dynamic analysis is a necessary complement to static analysis as it is less vulnerable to code obfuscation conversion.

3.2. Dynamic analysis

Analyzing the behavior of a malicious code (interaction with the system) while it is being executed in a controlled environment (virtual machine, simulator, emulator, sandbox etc.) is called dynamic analysis. Before executing the malware sample, the appropriate monitoring tools like Process Monitor [13] and Capture BAT [14] (for file system and registry monitoring), Process Explorer [15] and Process Hackerreplace [16] (for process monitoring), Wireshark [17] (for network monitoring) and Regshot [18] (for system change detection) are installed and activated. Various techniques that can be applied to perform dynamic analysis include function call monitoring,

function parameter analysis, information flow tracking, instruction traces and autostart extensibility points etc. [11]. Dynamic analysis is more effective as compared to static analysis and does not require the executable to be disassembled. It discloses the malware's natural behavior which is more resilient to static analysis. However, it is time intensive and resource consuming, thus elevating the scalability issues. The virtual environment in which malware are executed is different from the real one and the malware may perform in different ways resulting in artificial behavior rather than the exact one. In addition to this, sometimes the malware behavior is triggered only under certain conditions (on specific system date or via a specific command) and can't be detected in virtual environment. Several online automated tools exist for dynamic analysis of malware, e.g. Norman Sandbox [19], CWSandbox [20], Anubis [21] and TTAnalyzer [22], Ether [23] and ThreatExpert [24]. The analysis reports generated by these tools give in-depth understanding of the malware behavior and valuable insight into the actions performed by them. The analysis system is required to have an appropriate representation for malware, which are then used for classification either based on similarity measure or feature vectors. However a large number of new malware samples arriving at anti-virus vendors every day requires an automated approach to limit the number of samples that require close human analysis. Several Artificial Intelligence techniques, particularly machine-learning based techniques have been used in the literature for automated malware analysis and classification.

4. MALWARE ANALYSIS TOOLS

Analysts use tools for analyzing malware to protect and predict future attacks, and share knowledge among them. Open source tools are often the first choice to carry out such actions. It's no secret that distributing malware is a big business and the fast-growing malware epidemic will only grow in ability and efficiency in the years to come. Using open source malware analysis tools, researchers will check, identify and log different variants of malicious triggers when analyzing the life-cycle of attack. As malware trading forums are proliferating on the dark web, the crypters, botnets and zero-days needed to carry out powerful attacks have become easier than ever to get. With the growth of complexity of malware variants, the jobs of understanding and benchmarking the specific type have become harder. It's the job of security researchers and analysts to find out the right tool to analyze each specific type of attack. We now present some open source malware analysis tools that can help the researchers and security engineers.

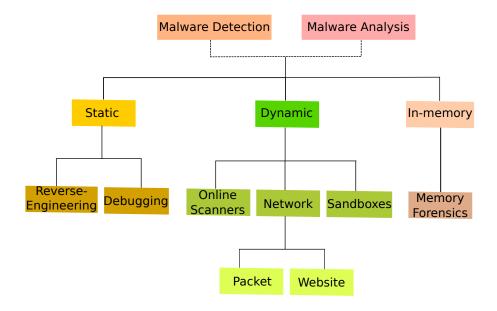


Figure 4. Malware Detection and Analysis Tools

4.1. Open Source Malware Analysis Tools

Google Rapid Response (GRR). The GRR platform is an incident response system developed by security researchers at Google, identifying common malware footprints workstations focused on remote live forensics. This consists of an application that is installed on the target system to communicate with the agent and a server infrastructure. GRR is a python client (agent) that is installed on target systems, and python server infrastructure that can manage and talk to clients. Once both the server side and the agent are deployed they can become GRR clients and start receiving messages from the servers. Then the incident response staff on the host computer will perform various technical operations, such as reviewing the memory, looking for different settings and handling software choices. GRR has been designed to run on a scale so analysts can easily capture and process data from large numbers of computers. GRR's goal is to support forensics and investigations in a simple, flexible way that allows investigators to rapidly triage incidents and conduct remote analysis.

REMnux. REMnux is a free Linux toolkit designed to assist malware analysts with malware reverse engineering. This seeks to make this easy for forensic investigators and accident witnesses to continue using the variety of free-to-use software that can analyze ransomware, although it may be difficult to locate or set up. This Linux toolkit has been developed as a one-stop shop for researchers searching for examples of reverse engineering malware. REMnux is focused on Ubuntu and integrates several resources into one for quickly analyzing malware based on Windows and Linux. The cornerstone of the project is the Ubuntu based REMnux Linux system. This

lightweight distro provides various resources to detect Windows and Linux ransomware, review browser-based vulnerabilities such as obfuscated JavaScript, investigate unusual text files, and uninstall other harmful objects. The distro can be used by investigators to intercept suspicious network traffic in an isolated laboratory. It helps researchers investigate browser-based malware, perform forensics on memory, analyze multiple samples of malware, extract and decode suspicious items, etc.

Cuckoo Sandbox. Created by a team of volunteers during the Google Summer of Code initiative back in 2010, it is an open source framework that automates malicious file analysis for Windows, OS X, Linux and Android and offers comprehensive and practical input on how each presented file operates in isolated environments. And since it is open source software, developers are constantly writing plugins that provide enhanced features. Cuckoo is used by malware detection and security firms to help ease the strain of manually wading through troves of potentially malicious data. The modular design allows the recording and analysis phases simple to configure. In recent years, it has, understandably, become one of the most commonly used open source tools. In 2012, Cuckoo published Malwr, a sandbox-as - a-service that allows users to use their collected data through an easy-to-use GUI. The goal was to act as an option for users who can't handle Cuckoo properly but still want to exploit their intellect.

Zeek. The Zeek Network Security Monitor (formerly Bro) is a versatile network dependent analytics system that transforms network traffic into events to cause scripts. It is comparable to an IDS (intrusion detection system) in that it gives users a bird's-eye view of their network activity, using both signature-based (looks for rules or trends of documented malicious traffic) and anomaly-based monitoring (looks for unusual activity). Nevertheless, its features go far beyond those of conventional IDS that can be used to conduct investigations in forensics, network monitoring and interface research. Although focussing on tracking network security, Zeek also provides a comprehensive forum for more general analysis of network traffic. Well grounded in more than 20 years of research, Zeek has since its inception succeeded in bridging the traditional gap between academia and operations.

Yara Rules. Another open source malware identification tool that can identify samples of malware based on textual or binary trends once they are tested in Cuckoo. Investigators use Yara to compose pattern-based definitions of the malware families. YARA stands for "Yet Another Recursive Acronym" as the descriptions are called rules. This helps researchers to identify and categorize apparently similar malware types and can be adapted for use inside Cuckoo. IBM calls Yara the "pattern matching Swiss army knife" of the malware researcher and can be used on both Windows and Linux computers. Yara's creators released a new service still in alpha called YaraRules Ana-

lyzer, that lets users analyze files in the cloud using full rulesets. This ensures that users are always analyzing samples against the most recent ruleset version and frees them from needing to install Yara locally. Yara rules have been added to many Endpoint Detection and Response framework to help them identify the malware samples they encounter, classify them and share their findings with clients and the community later.

Table 1 and table 2 list some tools available for malware detection, memory forensics, packet analysis, scanners/sandboxes, reverse engineering, debugging, and website analysis.

4.2. Mobile Malware Analysis Tools

APKTool. A tool for reverse engineering 3rd party, closed, binary Android apps. By making several changes, it can decode resources to almost original form and restore them. It also makes it easier to deal with a device owing to the project such as file creation and completion of some repetitive tasks such as creating the apk etc. We can decode APK resources to almost original form with the help of Apktool; we can modify the source code on the fly and rebuild the decoded resources back into APK. Its project-like structure makes working with them easy. Apktool can decode APK resources (resources.arsc, classes.dex and XMLs), rebuild decoded resources back to binary APK, organize and handle APKs that depend on framework resources along with automating the repetitive tasks.

Smali. Smali / baksmali is a dex format assembler / disassembler used by Dalvik, a Java VM implementation for Android. The syntax is loosely based on the syntax of Jasmin / dedexer, and follows the complete dex format features (annotations, debug data, line details, etc.). Also, code created by the baksmali is often considered to be written in the Smali language. Baksmali is a dex Bytecode software disassembler. The terms "Smali" and "Baksmali" are just the corresponding Icelandic versions of "assembler" and "disassembler." It wasn't an easy task to debug smali code before, but recently a wonderful plugin was launched for Intellij IDEA / Android Studio - Smalidea.

Dex2Jar. Dex2Jar is a free tool for dealing with the files Android ".dex" and Java ".class." Android programs are assembled into ".dex" (Dalvik Executable) scripts, which in effect are zipped onto the computer into a single apk file. Android will automatically create the ".dex" folders, by converting the compiled applications written in the Java. Dex2Jar reads the dex instruction to dexir format and can convert to ASM format. It can also be used to perform some basic deobfuscation. Dex2Jar's core feature is converting an APK classes.dex file to classes.jar, or vice versa. So, us-

Table 1. Tools available for malware detection, memory forensics, packet analysis, scanners and sandboxes.

Detection Tools		Online scanners and sandboxes	
AnalyzePE	Wrapper for a variety of tools for reporting on Windows PE files.	AndroTotal	Online analysis of APKs against multiple mobile antivirus apps
chkrootkit	Linux rootkit detector.	APK Analyzer	Dynamic analysis of APKs
Detect-It-Easy	A program for determining types of files.	AVCaesar	Online scanner and malware repository
hashdeep	Compute digest hashes with a variety of algorithms.	Cryptam	Analyze suspicious office documents
Loki	Host based scanner for IOCs.	Cuckoo Sandbox	Open source sandbox and automated analysis system
MASTIFF	Static analysis framework.	Comodo Valkyrie	File verdict system that conducts several analysi using run-time behavior and hundreds of feature from a file
MultiScanner	Modular file scanning/analysis framework	DeepViz	Multi-format file analyzer with machine-learning classification
nsrllookup	A tool for looking up hashes in NIST's National Software Reference Library database.	detux	A sandbox developed to do traffic analysis of Linux malware and capturing IOCs
PEV	A multiplatform toolkit to work with PE files, providing feature-rich tools for proper analysis of suspicious binaries.	Document Analyzer	Analysis of DOC and PDF files
Rootkit Hunter	Detect Linux rootkits.	DRAKVUF	Dynamic malware analysis system.
totalhash.py	Python script for searching in Total-Hash.cymru.com database.	File Analyzer	Free dynamic analysis of PE files
TrID	File identifier.	firmware.re	Unpacks, scans and analyzes firmware packages
YARA	Pattern matching tool for analysts.	Hybrid Analysis	Online malware analysis tool
	Memory Forensics Tools	IRMA	An asynchronous and customizable analysis plat form for suspicious files
DAMM	Differential Analysis of Malware in Memory, built on Volatility	Joe Sandbox	Deep malware analysis.
evolve	Web interface for the Volatility Memory Forensics Framework	Jotti	Online AV scanner
FindAES	Find AES encryption keys in memory	Limon	Sandbox for Analyzing Linux Malware
Muninn	A script to automate portions of analysis using Volatility, and create a readable report	Malheur	Automatic sandboxed analysis of malware behavior
Rekall	Memory analysis framework (from a Volatility fork).	Malwr	Free analysis with an online Cuckoo Sandbox in stance
TotalRecall	Script based on Volatility for automating various malware analysis tasks	MASTIFF Online	Online static malware analysis
Volatility	Advanced memory forensics framework.	Metadefender.com	Scan a file, hash or IP address for malware
WinDbg	Kernel debugger for Windows systems	NoDistribute	Scan files with over 35 anti-viruses.
Packet Analysis Tools		NVISO ApkScan	Dynamic analysis of APKs
Network Miner	A Network Forensic Analysis Tool (NFAT) for Windows	PDF Examiner	Analyse suspicious PDF files
NetworkTotal	Online analysis of .pcap files to detect viruses, worms, trojans and malware.	SEE	"Sandboxed Execution Environment", a frame work for building test automation in secured en vironments
PacketTotal	Online engine for analyzing .pcap files and visualizing the network traffic within, useful for mal-	URL Analyzer	Dynamic analysis of URL files
	ware analysis and incident response. My review		

ing any Java decompiler, it is possible to view the source code of an Android application, and it is fully legible. Here we get files from.class and not the real Java source code that the program author wrote.

Mobile-Sandbox. Mobile-Sandbox provides static and dynamic malware analysis for Android OS smartphones. The system is designed to automatically evaluate Android software in two novel ways: (1) by integrating static and dynamic analysis, i.e., static analysis findings are used to direct dynamic analysis and expand coverage of executed code, and (2) by using different logging methods for native API calls. It can evaluate the application with different modules within the static analysis component to get a summary of the program. To achieve this, it uses the VirusTotal service to perform several anti-virus scans, parse the manifest file and finally decompile the application to better identify suspect code. Within the dynamic analysis, it can run the application in an emulator and log every application operation, i.e. it logs both actions performed in the Java Virtual Machine Dalvik and actions performed in native libraries that may be bundled with the application.

4.3. Other Analysis Tools

Malzilla. Malzilla is a useful malware hunting tool for analyzing websites containing malicious code. Web pages that contain exploits often use a sequence of redirects and obfuscated code to make it difficult for someone to track them. This allows users to access websites and obtain all of their source code, such as wget, without visiting the site and potentially damaging their device. This program has the option of switching the user agents and picking the user's referrer. This shows the full list of webpages for browsers and all the headers for HTTP. It also has proxy features, complex decoders and, most notably, JavaScript code deobfuscation, all in one program.

Wireshark. Wireshark, a network monitoring application once known as Ethereal, records packets and shows them in the human-readable format in real-time. It intercepts traffic and transforms the binary data into a readable format for users. Wireshark includes filters, color coding, and other features that allow individuals to dig deep into network traffic and inspect individual packets. It is the leading network traffic analyzer in the world, and an essential tool for any skilled security or device administrator. This free software allows people to track network traffic in real-time, and is often the best tool on any network for troubleshooting problems. Common issues that Wireshark can deal with troubleshooting include lost messages, latency issues and malicious network operation. It enables network data to be held under a microscope and offers resources for filtering and digging into that information, zooming into the root cause of the issue. It is used by management

Table 2. Tools available for reverse engineering, debugging, and website analysis.

Reverse	Engineering and Debugging Tools	strace	Dynamic analysis tool for Linux executables
angr	Platform-agnostic binary analysis framework	Triton	A dynamic binary analysis (DBA) framework
bamfdetect	Identifies and extracts information from bots and malware	Udis86	Disassembler library and tools
BARF	Open source multiplatform Binary Analysis and Reverse engineering Framework.	Vivisect	Python tool for malware analysis
binnavi	Binary analysis IDE for reverse engineering	X64dbg	Debugger for windows
Capstone	Disassembly framework for binary analysis and reversing		Website Analysis Tools
codebro	Web based code browser with basic code analysis.	Desenmascara.me	Tool to retrieve metadata from websites
dnSpy	.NET assembly editor, decompiler and debugger	Dig	Online dig and other network tools
Evan's Debugger (EDB)	Modular debugger with a Qt GUI	dnstwist	Domain name permutation engine for detecting typo squatting, phishing and corporate espionage
Fibratus	Windows kernel exploration and tracing tool	Firebug	Firefox extension for web development.
GDB	The GNU debugger	IPinfo	Gather information about an IP or domain by searching online resources
GEF	GDB Enhanced Features, for exploiters and reverse engineers	Java Decompiler	Decompile and inspect Java apps
hackers-grep	Uility to search for strings in PE executables	Java IDX Parser	Parses Java IDX cache files
IDA Pro	Windows disassembler and debugger	JSDetox	JavaScript malware analysis tool
Immunity Debugger	Debugger for malware analysis	jsunpack-n	Javascript unpacker that emulates browser functionality
ltrace	Dynamic analysis tool for Linux executables	Krakatau	Java decompiler, assembler, and disassembler
objdump	Static analysis tool for Linux binaries	Machinae	OSINT tool for gathering information about URLs, IPs, or hashes
OllyDbg	Debugger for Windows executables	mailchecker	Cross-language temporary email detection library
PANDA	Platform for Architecture-Neutral Dynamic Analysis	Malzilla	Analyze malicious web pages.
PEDA	Python Exploit Development Assistance for GDB	PunkSpider	Web application vulnerability search engine.ÂăMy review
pestudio	Static analysis tool for Windows executables	RABCDAsm	ActionScript Bytecode Disassembler
plasma	Interactive disassembler for x86/ARM/MIPS	SenderBase	Search for IP, domain or network owner
PPEE (puppy)	PE file inspector.	Spidermonkey	Mozilla's JavaScript engine, for debugging malicious JS
Process Monitor	Advanced monitoring tool for Windows programs	Sucuri SiteCheck	Website Malware and Security Scanner
Pyew	Python tool for malware analysis	swftools	Adobe Flash decompiler.
Rdare2	Reverse engineering framework	TekDefense Automator	OSINT tool for gathering information about URLs, IPs, or hashes
ROPMEMU	Framework to analyze, dissect and decompile complex code-reuse attacks	xxxswf	Analysis tool for Flash files
SMRT	Sublime Malware Research Tool, a plugin for Sublime Text 3 focused on malware analysis.	ZScalar Zulu	Zulu URL Risk Analyzer

to detect defective network equipment that loses packets, latency problems caused by machines transmitting traffic around the world and data exfiltration or even intrusion attempts against any entity.

SysAnalyzer. SysAnalyzer is an open-source tool developed to provide an interactive resource for malware researchers to easily compile, analyze, and monitor the behavior that a binary performed when operating on the network. It is an interactive framework for the malcode run time analysis that tracks different aspects of device and method states. SysAnalyzer was designed to allow analysts to rapidly create a detailed report. SysAnalyzer's main components function off of comparing device snapshots over a given user time interval. Similar to a live monitoring system, the reason a snapshot method was used is to reduce the amount of data analysts need to wade through when performing their research. By using a snapshot method, audiences can easily display only the persistent changes that have been identified since the first run of the application.

VirusTotal. Virustotal is a service that analyzes suspicious files and URLs and helps to detect viruses, worms, Trojans and all kinds of malware detected by antivirus engines quickly. In addition to a variety of methods for removing signals from the studied material, VirusTotal inspects products with over 70 antivirus scanners and URL / domain blacklisting services. Every person can use their browser to pick a file from their device, and submit it to VirusTotal. VirusTotal offers various methods for uploading data, including the default public web portal, desktop uploaders, browser extension, and a programmatic API. The web interface has the greatest scanning priority among the forms of application which are available to the public. The specifications can be made using the HTTP-based public API in any programming language. It also offers a variety of other functions, including the VirusTotal Community: a network that allows users to report on files and URLs and exchange comments with each other. This can be helpful in detecting malicious content and also in finding false positives – regular and harmless objects identified as dangerous by one or more scanners.

5. CONCLUSION

This survey paper presents a summary of malware detection and analysis techniques and tools. In particular, the different tools available for malware detection, memory forensics, packet inspection, scanners/sandboxes, reverse engineering, hacking, and website analysis have been thrown light. Since most of the current surveys typically concentrate on a specific subset of the model, this paper offers an in-depth study of methods to identify and evaluate malware with a clear understanding of domain-specific analytics.

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