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Article

# Using ASCON-Based fuzzy Hashing for Efficient Malware Analysis

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Abstract: The rapid proliferation of digital devices and connectivity has seen an unprecedented rise in the threats of cyber-attacks, making cybersecurity a high research priority worldwide. Among the various forms of cyberattacks, malicious software or malware, significantly disrupts the privacy and integrity of data on millions of computing devices globally. It is already a well-established notion that robust malware analysis underpins successful protection systems in the rapidly changing threat landscape. Traditional malware detection techniques, although effective against known malware types, are often found wanting against new and sophisticated varieties. The situation necessitates the exploration of advanced techniques like cryptographic hashing and fuzzy hashing that have proven instrumental in handling malware's dynamic nature. Fuzzy hashing, by allowing a score-based match rather than an exact digital signature, enables the identification of unknown malware that are derivatives of known malwares. However, the efficiency of the fuzzy hashing algorithms comes to the forefront concerning the encryption used. The research paper titled, "Using ASCON-based Fuzzy Hashing for Efficient Malware Analysis," seeks to explore leveraging the ASCON encryption, a lightweight but secure encryption standard, in the application of fuzzy hashing for malware detection.

Keywords: cryptography; malware analysis; ASCON

#### 1. Introduction

The rapid proliferation of digital devices and connectivity has seen an unprecedented rise in the threats of cyber-attacks, making cybersecurity a high research priority worldwide. Among the various forms of cyberattacks, malicious software or malware, significantly disrupts the privacy and integrity of data on millions of computing devices globally. It is already a well-established notion that robust malware analysis underpins successful protection systems in the rapidly changing threat landscape.

Traditional malware detection techniques, although effective against known malware types, are often found wanting against new and sophisticated varieties. The situation necessitates the exploration of advanced techniques like cryptographic hashing and fuzzy hashing that have proven instrumental in handling malware's dynamic nature. Fuzzy hashing, by allowing a score-based match rather than an exact digital signature, enables the identification of unknown malware that are derivatives of known malwares. However, the efficiency of the fuzzy hashing algorithms comes to the forefront concerning the encryption used.

The research paper titled, "Using ASCON-based Fuzzy Hashing for Efficient Malware Analysis," seeks to explore leveraging the ASCON encryption, a lightweight but secure encryption standard, in the application of fuzzy hashing for malware detection.

#### 2. Research Motivation

The use of advanced cryptographic algorithms for secure communications and data storage has been widely researched and documented. Among them, ASCON has emerged as a promising alternative owing to its high security and performance, especially in constrained environments. The combination of this encryption technique with fuzzy hashing for the purpose of malware identification and analysis is an unchartered research area, which forms the crux of this study.



Moreover, literature has a sufficient number of works employing other cryptographic algorithms for fuzzy hashing, but the inclusion of ASCON in scenario remains scantily explored. Thus providing both an opportunity and a necessity for an in-depth study.

The motivation behind the research is threefold. Firstly, with malware becoming more sophisticated and diverse, there is a constant need to upgrade and refine detection and analysis techniques such as fuzzy hashing. Secondly, the potential advantages of using a lightweight algorithm like ASCON in constrained systems represent an exciting avenue to explore. Finally, research in this area could significantly contribute to the further strengthening of cybersecurity systems, fulfilling the increasing need for more robust safeguards against cyber threats.

#### 3. Literature Review

Li, Y., Sundaramurthy, S.C., Bardas, A.G., Ou, X., Caragea, D., Hu, X. and Jang, J., 2015 [1] the paper examines fuzzy hashing algorithms for clustering malware and identifies limitations in current methods. It proposes a new fuzzy hash function inspired by current algorithms and advanced malware clustering approaches, which outperforms existing ones. The study also compares the performance of nextGen-hash and nextGen-hash-imp algorithms, extracting different features from malware samples. The results show that "import table entries" features are more representative and efficient than low-level n-gram features in comparing overall similarity, suggesting a potential improvement in malware clustering.

Rodriguez-Bazan, H., Sidorov, G. and Escamilla-Ambrosio, P.J., 2023 [2]. The research presents a new approach for classifying Android ransomware using an image-based Convolutional Neural Network (CNN). Using Natural Language Processing (NLP) and Fuzzy Hashing, the method converts an Android Application Package (APK) into a grayscale image. When tested on a dataset of 7,765 ransomware samples, the technique outperformed previous methods in the literature.

Shiel, I. and O'Shaughnessy, S., 2019 [3]. In order to get around the drawbacks of file-level similarity hashing, the study investigates the use of section-level similarity hashing to categorize malware variations. Studies carried out on well-known malware families demonstrate that, when it comes to identifying malware in Windows Portable Executable (PE) files, section-level hashing and comparison perform noticeably better than file-level hashing.

Tariq, U., Ahmed, I., Khan, M.A. and Bashir, A.K., 2023 [4]. The paper covers the difficulties and weaknesses in Internet of Things security and suggests workable solutions, such as blockchain technology and quantum cryptography in addition to more established encryption techniques. By classifying attacks and vulnerabilities, examining attack techniques, and presenting security solutions through case studies of important IoT applications, it seeks to serve as an invaluable resource for researchers.

Aghili, S.F., Sedaghat, M., Singelée, D. and Gupta, M., 2022 [5]. This paper presents the method and its limitations for classifying malware variations in Windows Portable Executable (PE) files. It does not evaluate how well it works in contrast to malware that has been disguising itself, which may be difficult to find using standard methods. There might be an extra computational expense because the method splits malware executables into binary headers and sections. It also has limited generality to unknown malware families because it applies a database of existing malware families to classify new samples. For the process to be successful, section information must be present in the malware executables, which may be deliberately removed or obfuscated to evade detection.

Naik, N., Jenkins, P. and Savage, N., 2019 [6]. This study presents a method for detecting new or undiscovered ransomware variants using fuzzy hashing. The method uses three fuzzy hashing methods (SSDEEP, SDHASH, and mvHASH-B) on a WannaCry or WannaCryptor malware corpus. The success rate of similarity detection and fuzzy similarity scores are used to assess the effectiveness of the approaches. The results are compared to determine the relative accuracy of the chosen fuzzy hashing algorithm. This method is crucial for organizations to focus on ransomware prevention, as it denies access to data and poses a significant threat to their information systems. The results are used

to determine the relative accuracy of the chosen fuzzy hashing algorithm alongside their malware prevention strategy.

Naik, N., Jenkins, P., Savage, N., Yang, L., Boongoen, T. and Iam-On, N., 2020, July [7]. The research offers a methodology called fuzzy-import hashing, which combines fuzzy hashing with import hashing approaches to detect and analyze malware. This integration seeks to increase detection accuracy while minimizing performance impact. Experiments on gathered malware and goodware corpus, comparative evaluation against YARA rules, and use in fuzzy c-means clustering demonstrate the suggested technique.

Naik, N., Jenkins, P., Savage, N., Yang, L., Boongoen, T., Iam-On, N., Naik, K. and Song, J., 2021 [8]. The research presents two ways for improving the efficiency of YARA rules for malware detection without increasing complexity and overheads: upgraded YARA rules and embedded YARA rules. These methods use fuzzy hashing and fuzzy rules to estimate the chance of a file containing malware, resulting in better detection results than standard YARA rules.

Naik, N., Jenkins, P., Savage, N., Yang, L., Naik, K., Song, J., Boongoen, T. and Iam-On, N., 2020, December [9]. The study proposes employing fuzzy hashing alongside basic YARA rules to improve the detection rate of YARA rules for malware triaging. By comparing the detection rate of upgraded YARA rules to existing triaging approaches such as fuzzy hashing and import hashing, the suggested strategy is proven to improve overall triaging results.

Chang, D., Hong, D., Kang, J. and Turan, M.S., 2022 [10]. The study investigates the Ascon family's resistance to conditional cube assaults in a nonce-misuse environment. It introduces novel state- and key-recovery attacks on the Ascon family, such as Ascon-128a and Ascon-80pq. The attacks recover the whole state and secret key of Ascon-128a in less cycles, exceeding the designers' data limit. Ascon-128's incomplete state information can also be restored. Furthermore, given that the whole state information of Ascon-80pq was obtained in a prior assault, the study demonstrates that the Ascon-80pq's 160-bit secret key can be recovered. These assaults shed light on Ascon's security in a non-misuse environment.

Rodriguez-Bazan, H., Sidorov, G. and Escamilla-Ambrosio, P.J., 2023 [2]. The research provides a new method for identifying Android ransomware based on photos using a Convolutional Neural Network (CNN) generated by converting an Android Application Package (APK) into a grayscale image using Natural Language Processing (NLP) techniques and Fuzzy Hashing. The suggested method was evaluated on a dataset of 7,765 Android ransomware samples and outperformed previous methods in the literature in terms of accuracy.

Mostafa, A.M., Ezz, M., Elbashir, M.K., Alruily, M., Hamouda, E., Alsarhani, M. and Said, W., 2023 [11]. The paper describes an adaptive multi-factor multi-layer authentication architecture for cloud user authentication, with the goal of improving cloud platform security and preventing unwanted access and data breaches. The framework includes access control, intrusion detection algorithms, and automatic authentication method selection. To improve identity verification, it employs several authentication elements such as user factors, geolocation, and browser confirmation. To protect login information, an AES-based encryption component is also used. The suggested approach performs admirably in identifying potentially harmful users and intruders, hence efficiently preventing intentional attacks on cloud services and data.

Rodriguez-Bazan, H., Sidorov, G. and Escamilla-Ambrosio, P.J., 2023 [2]. The study proposes a framework for safe authentication in cloud systems, however its assessment is purely theoretical, with no specific analysis of its performance or scalability. When compared to simpler techniques, the framework's multi-factor multi-layer authentication strategy may add complexity and difficulty for consumers. This limitation is not addressed in the study, nor are options for streamlining the authentication process without compromising security explored. The behavioral analysis component of the system is dependent on accurate profiling of user behavior patterns, which may be sensitive to changes in user behavior or discrepancies in data gathering. The framework could also be prone to social engineering attacks, which could be avoided by user education and awareness training. The

article does not go into detail on the integration procedure or its interoperability with existing cloud systems.

Pagani, F., Dell'Amico, M. and Balzarotti, D., 2018, March [12]. The research compares four prominent fuzzy hashing algorithms in various contexts, finding that ssdeep outperforms competing methods. The optimal algorithm for identifying similarities across binary program files depends on the specific use case situation, as no study has determined the best suited algorithm for this purpose.

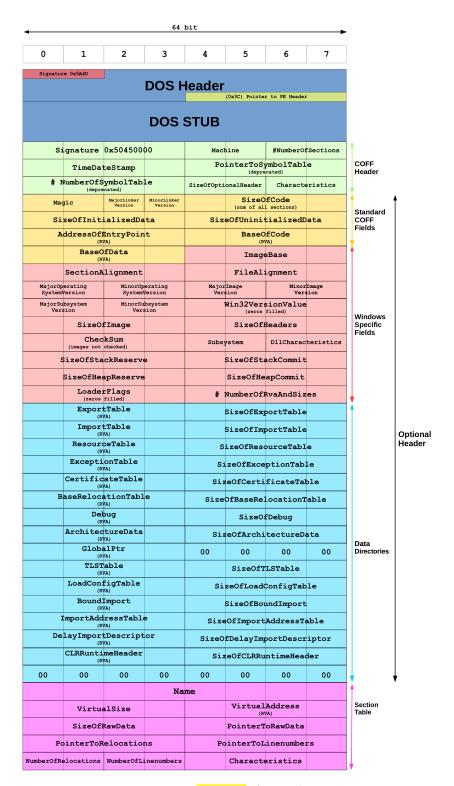
Namanya, A.P., Awan, I.U., Disso, J.P. and Younas, M., 2020 [13]. Because harmful files are evading existing security systems, this research investigates merging different hashing approaches to offer a quantifiable malware score and achieve higher detection rates. The proposed approach for malware rating based on hash results improves true detection rates of malware significantly (> 90%).

Ali, H., Batool, K., Yousaf, M., Islam Satti, M., Naseer, S., Zahid, S., Gardezi, A.A., Shafiq, M. and Choi, J.G., 2022 [14]. The study presents a method for identifying malicious Android applications utilizing repacked malicious code and fuzzy reasoning for categorization. The suggested methodology outperforms existing similar approaches with a detection rate of roughly 74% for repacked malware.

Samra, A.A.A., Qunoo, H.N., Al-Rubaie, F. and El-Talli, H., 2019, March [15]. The report includes a survey that compares the two leading static Android malware detection approaches: permission-based detection and signature-based detection. The study's goal is to give scholars a comprehensive knowledge of the parallels, differences, and correctness of key published research in this subject.

Aslan, Ö.A. and Samet, R., 2020 [16]. The study examines several malware detection tactics and procedures, highlighting their advantages and disadvantages. It underlines the difficulties in identifying both known and unknown malware, as well as the necessity for fresh research and methodologies in this field.

Amira, A., Derhab, A., Karbab, E.B. and Nouali, O., 2023 [17]. The study examines the use of community detection algorithms in malware analysis to detect malware families and variations rather than individual instances of malware, which can reduce detection time dramatically. The survey examines cutting-edge malware analysis solutions that use community detection algorithms and offers a taxonomy that categorizes the solutions based on analysis task, community detection approach, target platform, analysis kind, and source of features. It also indicates research gaps as well as potential future research directions.



**Figure 1.** Structure of a PE File

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Experimental Study of Fuzzy	The paper examines fuzzy	The paper discusses the chal-	The paper critiques current	The study evaluates popu-
Hashing in Malware Cluster-	hashing techniques for	lenge of automating malware	fuzzy hashing algorithms for	lar fuzzy hashing algorithms
ing Analysis	malware similarity analy-	grouping using fuzzy hash-	malware similarity analysis	but lacks a wider range of op-
	sis, addressing issues and	ing algorithms, highlighting	and suggests improved per-	tions, limiting generalization.
	proposing new methods. It	the lack of rigorous exper-	formance by considering in-	It also doesn't consider mal-
	also explores block-based	imentation in existing re-	put data structure, high-level	ware families' impact on per-
	distance computation	search. It examines exist-	semantic features, and com-	formance, making it unclear
	and uses nextGen-hash	ing fuzzy hashing algorithms	paring Bloom filters and fea-	how the findings apply to dif-
	and nextGen-hash-imp	and their limitations in mal-	ture hashing. It also intro-	ferent types of malware.
	algorithms to evaluate	ware clustering, proposing	duces a new approach called	
	performance variations influ-	novel ways to construct these	feature hashing"."	
	enced by different features	algorithms that outperform		
	in a malware dataset.	existing ones.		
Android Ransomware Anal-	The study presents a method	A new method for catego-	The proposed method for	The paper uses a small
ysis Using Convolutional	for classifying malware us-	rizing Android ransomware	Android ransomware catego-	dataset of 1,000 Android
Neural Network and Fuzzy	ing data pretreatment and	uses Natural Language Pro-	rization uses Convolutional	samples for ransomware de-
Hashing Features	natural language process-	cessing (NLP) techniques,	Neural Network (CNN) and	tection, which may limit its
	ing techniques. It con-	Fuzzy Hashing, and Con-	fuzzy hashing characteris-	applicability in real-world
	verts Android APK files into	volutional Neural Network	tics, achieving higher accu-	scenarios. The model's preci-
	grayscale photos, extracts	(CNN) for malware classifica-	racy than existing methods.	sion, recall, and F1-score met-
	and cleans content using Nat-	tion. Experimental testing on	With 7,765 samples, the CNN	rics are not comprehensive,
	ural Language Processing al-	7,765 ransomware samples	model achieved an average	and its computationally ex-
	gorithms, and trains a con-	demonstrates the method's	accuracy of 98.16%, outper-	pensive convolutional neural
	volutional neural network to	accuracy, focusing on ran-	forming other machine learn-	network may result in false
	classify malware families.	somware's potential mali-	ing techniques like k-NN,	positives, making it unsuit-
		ciousness and making re-	Random Forest, Multilayer	able for real-time scenarios.
		covery difficult without pay-	Perceptron, and Support Vec-	
		ment.	tor Machine.	

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Improving file-level fuzzy	The study identifies mal-	The paper introduces section-	The study used file and sec-	The paper presents a method
hashes for malware variant	ware variations in Windows	level similarity hashing as a	tion level digests to pre-	for categorizing malware
classification	Portable Executable files by	technique for malware vari-	dict malware family mem-	variations in Windows
	dividing malware into binary	ant classification, addressing	bership in Windows Portable	Portable Executable files,
	headers and sections, and ap-	the limitations of file-level	Executable files. Results	but its effectiveness is lim-
	plying a similarity digest to	hashing. The study evaluates	showed section-level hash-	ited due to its inability to
	each part. Experiments show	the effectiveness of similar-	ing and comparison outper-	identify obfuscated malware.
	section-level hashing is the	ity hashing at file and section	formed file-level hashing in	Section-level fuzzy hashing
	most effective method, but	levels, demonstrating its ad-	malware classification, par-	may have higher compu-
	the F-measure measures ef-	vantages and the viability of	ticularly recall. The non-	tational overhead, and its
	fectiveness.	classifying malware variants	obfuscated segment method	generalization to unknown
		using similarity digests.	performed better. The study	malware families is limited.
			demonstrated that section-	Section information in
			level similarity hashing can	malware executables may
			overcome file-level hashing	be intentionally erased or
			constraints and is feasible for	obfuscated.
			malware classification.	
Fortifying IoT against crim-	The paper uses a systematic	The article provides a com-	The study analyzes IoT secu-	This work presents a tech-
pling cyber-attacks: a system-	approach to evaluate security	prehensive analysis of IoT	rity research, identifying vul-	nique for categorizing mal-
atic review	challenges in IoT and create a	security flaws and attacks,	nerabilities and attacks based	ware variations in Windows
	comprehensive threat taxon-	focusing on innovative	on their intended recipients.	Portable Executable files, but
	omy. It conducts a thorough	technologies like blockchain	It provides a comprehensive	it has drawbacks. It doesn't
	literature search using rele-	and QC-PUFs. It reviews	analysis of attack techniques	compare it to disguised
	vant keywords across rep-	existing research, evaluates	and suggests defenses. The	malware, divides malware
	utable databases. The au-	the effectiveness of security	study presents case studies	executables into binary
	thors select key network se-	features and protocols, and	of IoT applications, assesses	headers and sections, incurs
	curity topics based on their	suggests incorporating	security features, suggests	computational costs, and
	expertise. The study evalu-	blockchain and machine	improvements, and uses sur-	has limited generalization to
	ates the effectiveness of exist-	learning algorithms to en-	vey data to identify new secu-	unknown malware families.
	ing security features and pro-	hance IoT security.	rity issues. It distinguishes it-	The technique requires mal-
	tocols in mitigating IoT risks		self from surveys by using an	ware executables to contain
	and proposes improvements		extensive research methodol-	section information and
	to address identified gaps.		ogy.	can only identify malware
				from a small dataset of 2,400
			_	samples.

Dan au Nama	Mothod and Tochniques	Contributions	Dogulto	Duine aux Limitation
Paper Name MLS-ABAC: Efficient Multi-	Method and Techniques	Contributions The attacks in two days as a Marking	Results	Primary Limitation
	In order to comply with the	The study introduces a Multi-	Experimental results show	The proposed access control
Level Security Attribute-	requirements of the NIST	Level Security ABAC (MLS-	that the MLS-ABAC scheme	approach has several draw-
Based Access Control	Attribute-Based Access Con-	ABAC) scheme that effec-	fits the requirements of	backs, including its policy
scheme	trol (ABAC) model, the pa-	tively controls access to sen-	NIST's ABAC model for	function being limited to
	per presents the Multi-Level	sitive data in an IoT con-	IoT application. It achieves	AND-gate circuits, slow de-
	Security ABAC (MLS-ABAC)	text, adhering to NIST's	a constant ciphertext size	cryption technique, and lack
	scheme. It presents both con-	Attribute-Based Access Con-	of 230 bytes and efficient	of revocation challenge. The
	ceptual and formal models	trol (ABAC) model stan-	encryption and decryption	algorithm is inefficient and
	and employs an outsource-	dards. The approach in-	running times of 18 and 10	does not address user revoca-
	able Ciphertext-Policy ABE	cludes both conceptual and	ms.	tion, covert channel attacks,
	system for decryption. A	formal models and uses		or sanitization. The tech-
	constant ciphertext size and	an outsourceable Ciphertext-		nique also lacks clear meth-
	encryption and decryption	Policy ABE scheme for de-		ods for confirming encrypted
	times of 18 and 10 ms are ex-	cryption. Experimental data		data accuracy and does not
	amples of performance mea-	shows an efficient encryption		address security risks asso-
	sures.	and decryption running time		ciated with outsourcing de-
		of 18-10 ms.		cryption services to external
A Days a supplied to a	T1	The state of the same of the s	The state of the same of the s	cloud providers.
A Ransomware Detection	The paper uses fuzzy	The study presents a fuzzy	The study presents a ran-	The proposed solution de-
Method Using Fuzzy Hash-	hashing for ransomware	hashing-based ransomware	somware detection method	tects ransomware at the file
ing for Mitigating the Risk	detection, using three meth-	detection strategy to mitigate	using fuzzy hashing meth-	level, but may not be suf-
of Occlusion of Information	ods: SSDEEP, SDHASH, and mvHASH-B. It examines	the risk of information system attacks. It considers the	ods SSDEEP, SDHASH, and mvHASH-B to assess	ficient for all attacks. Ran- somware often uses tactics to
Systems	the success rate of each			avoid detection, like insert-
		polymorphic behavior and	similarity identification success rates and cluster a	· ·
	method in detecting similar-	dispersion of ransomware		ing malicious code or mod-
	ity and clusters the collected	variants, utilizing three algo-	ransomware corpus. This	ifying its structure. The
	ransomware corpus. The	rithms (SSDEEP, SDHASH,	method aims to reduce	study focuses on recogniz-
	findings are compared to de-	and mvHASH-B) to assess	information system blockage	ing known ransomware vari-
	termine the relative accuracy	similarity identification suc-	caused by ransomware	ations and does not dis-
	of the chosen fuzzy hashing	cess rates and cluster the col-	attacks by comparing new	cuss fuzzy hashing's effec-
	methods, focusing on the	lected ransomware corpus.	or unknown variants with	tiveness against polymor-
	WannaCry or WannaCryptor malware.		existing samples, providing	phic ransomware. Fuzzy
	marware.		insight into the chosen	hashing's effectiveness de-
			fuzzy hashing algorithms'	pends on target file availabil-
			accuracy.	ity and may not scale effec-
				tively for large data amounts.

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Fuzzy-Import Hashing: A	The research introduces	The research proposes a	The fuzzy-import hashing	Fuzzy-import hashing, a
Malware Analysis Approach	fuzzy-import hashing, a	fuzzy-import hashing strat-	strategy, which combined	method for identifying
	combination of fuzzy and	egy for malware identifica-	fuzzy and import hashing,	malware, may be ineffective
	import hashing algorithms	tion and analysis, combining	improved malware detec-	against disguised malware
	for malware identification	fuzzy hashing with import	tion rates. This technique	and prone to false positives.
	and analysis. Studies, includ-	hashing methods to improve	was tested on a malware	Its success depends on the
	ing malware and goodware	detection accuracy without	and goodware corpus, and	availability and correctness
	corpus examination, YARA	affecting overall analysis per-	its efficacy was confirmed	of import table information,
	rule comparison, and fuzzy	formance. This strategy of-	1 .	which can be computation-
	c-means clustering imple-	fers higher detection rates	parative evaluation against	ally expensive in real-time
	mentation, demonstrate its	and generates fuzzified data	YARA rules, and fuzzy c-	virus detection applications.
	success.	for later clustering or classi-	means clustering.	The study focuses on detect-
		fication. Studies show the ef-		ing known malware variants,
		ficacy of this method, includ-		but a more comprehensive
		ing malware and goodware		evaluation would include
		corpus examination.		testing against a broader
				range of malware samples,
				including polymorphic
				and extensively obfuscated
				version

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Embedded YARA	The study suggests two ways	The study proposes two	The enhanced YARA rules	The proposed embedded
rules: strengthening YARA	to enhance the effectiveness	strategies to enhance the	based on SSDEEP showed	YARA rules approach relies
rules utilising fuzzy hashing	of YARA rules in malware	performance and detection	slightly better total similar-	on fuzzy rule correctness,
and fuzzy rules for malware	detection: upgraded YARA	of YARA rules in malware	ity detection rate (67.1 com-	which is determined by the
analysis	rules and embedded YARA	analysis: enhanced YARA	pared to standard YARA	quality and completeness of
	rules. Upgraded YARA rules	rules and embedded YARA	rules, 62.2), but the improve-	the underlying fuzzy rule
	use fuzzy hashing to de-	rules. The enhanced method	ment was marginal. The em-	base. However, its effective-
	termine file malware risk,	uses fuzzy hashing to eval-	bedded YARA rules, which	ness may be reduced if fuzzy
	while integrated YARA rules	uate files not recognized as	mix fuzzy hashing and fuzzy	rules are not well-defined or
	combine fuzzy hashing with	malware, while the embed-	rules, performed marginally	do not capture malware fea-
	fuzzy rules, considering im-	ded technique uses fuzzy	better or worse. The op-	tures. The approach's gen-
	precise or ambiguous data.	hashing and fuzzy rules	timum analysis method for	eralization to unknown mal-
	The expanded YARA rules	to produce probabilistic re-	all cases is challenging, in-	ware families is limited, po-
	technique uses the Damerau-	sults. The experimental re-	dicating further research is	tentially leading to false neg-
	Levenshtein distance mea-	sults show these strategies	needed.	atives. Additionally, process-
	sure.	are more effective than sim-		ing overhead can affect real-
		ple YARA guidelines.		time malware analysis capa-
				bilities when dealing with
				large files.

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Fuzzy Hashing Aided En-	The paper uses three triag-	The research presents fuzzy	The study compares the	The proposed YARA rules us-
hanced YARA Rules for Mal-	ing methods to assess mal-	hashing-aided YARA rules to	detection rate of enhanced	ing fuzzy hashing may not be
ware Triaging	ware samples: fuzzy hashing,	enhance detection rates with-	YARA rules to various	effective in detecting exten-
	import hashing, and YARA	out complexity or overheads.	triaging approaches, finding	sively obfuscated malware
	rules. Fuzzy hashing pro-	It analyzes three triaging	that enhanced YARA rules	due to changes in file struc-
	duces fuzzy hash values for	approaches on ransomware	marginally outperform basic	ture and signatures. Fuzzy
	ransomware samples, com-	samples: fuzzy hashing, im-	YARA rules (67.1%). The	hashing is prone to false pos-
	paring them to previously	port hashing, and YARA	study highlights the need	itives and may mistake inno-
	discovered samples. Im-	rules. When one method	for efficient and resource-	cent files for malware if they
	port hashing generates IM-	fails, the methodology sup-	optimized methodologies	share structural similarities
	PHASH hash values, com-	plements basic YARA rules,	for analyzing large malware	with known samples. The
	paring them to previously de-	increasing detection rates.	volumes. The study applies	success of the method de-
	tected samples. YARA rules	The article emphasizes the	three triaging approaches:	pends on the chosen parame-
	use pattern matching to de-	need for efficient, resource-	fuzzy hashing, import	ters, such as similarity thresh-
	tect and classify malware,	optimized methodologies for	hashing, and YARA rules, fo-	old and hashing algorithm.
	with an improved version	malware analysis.	cusing on malware samples	The method may also miss
	adding fuzzy hashing to im-		for extended experimenta-	threats due to incorrect cat-
	prove detection rates.		tion.	egorization of malware sam-
				ples from new or unknown
				families.

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Resistance of Ascon Family	The paper explores the As-	The study presents new	In a nonce-misuse situa-	The study evaluates the As-
Against Conditional Cube	con family's resistance to	state- and key-recovery at-	tion, the study proposes	con family's resistance to
Attacks in Nonce-Misuse Set-	conditional cube attacks in	tacks against the Ascon fam-	new state- and key-recovery	conditional cube attacks, but
ting	a nonce-misuse context. It	ily, including Ascon-128a	attacks against the Ascon	doesn't cover its overall se-
	presents new state- and key-	and Ascon-80pq, in a nonce-	family, including Ascon-128a	curity against other attacks.
	recovery attacks on the fam-	misuse scenario. These at-	and Ascon-80pq. These at-	It mainly focuses on 3-round
	ily, including Ascon-128a	tacks recover the complete	tacks recover the complete	and 4-round examples, but
	and Ascon-80pq. These at-	state and secret key of Ascon-	state and secret key of Ascon-	could be improved by in-
	tacks recover the complete	128a in less cycles, exceeding	128a in less cycles, exceed-	vestigating higher-round in-
	state and secret key of Ascon-	the designers' data limit. The	ing the designers' data limit.	stances. The research as-
	128a in less cycles, exceeding	authors also demonstrate the	They also demonstrate the	sumes perfect data align-
	the designers' data limit. The	recovery of partial state in-	recovery of Ascon-128 par-	ment and sufficient memory,
	study also demonstrates the	formation and the 160-bit se-	tial state information and	which may not be practical in
	recovery of partial state in-	cret key, provided the com-	the Ascon-80pq 160-bit secret	real-world situations. Com-
	formation and the 160-bit se-	plete state information was	key, provided complete state	paring conditional cube at-
	cret key, revealing Ascon's se-	acquired in a previous attack.	information was acquired in	tacks to real-world imple-
	curity in a nonce-misuse sce-	These attacks provide valu-	a previous attack. These as-	mentations would provide
	nario.	able insights into the Ascon	saults provide useful insights	more insights.
		family's security vulnerabili-	into the Ascon family's secu-	
		ties and weaknesses.	rity and emphasize the need	
			for additional investigation	
			and enhancements.	

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Android Ransomware Anal-	A new method for catego-	The study presents a new	The Convolutional Neural	The study presents a
ysis Using Convolutional	rizing Android ransomware	method for identifying	Network (CNN) method has	methodology for detecting
Neural Network and Fuzzy	uses Natural Language Pro-	Android ransomware us-	been proven to be more ac-	ransomware using 1,000
Hashing Features	cessing (NLP) techniques	ing Convolutional Neural	curate than existing meth-	Android samples, but it
	and Fuzzy Hashing, fol-	Network (CNN) and images	ods in categorizing Android	lacks evaluation and is
	lowed by a Convolutional	generated from an Android	ransomware on a dataset of	based on fuzzy hashing
	Neural Network (CNN) for	Application Package (APK)	7,765 samples. The CNN	features, which can lead to
	malware classification. The	using Natural Language	model achieved 94.37% ac-	false positives. The model is
	method outperforms other	Processing (NLP) techniques	curacy when using the com-	computationally expensive
	methods in accuracy, as	and Fuzzy Hashing. The	plete dataset and 98.16%	and unsuitable for real-time
	demonstrated in experimen-	method outperforms other	when using the five most rep-	ransomware detection due
	tal testing on 7,765 An-	methods in accuracy, as	resentative classes. It outper-	to its reliance on a known
	droid ransomware samples.	demonstrated in experi-	formed other machine learn-	malware database. The
	The research focuses on ran-	mental testing on 7,765	ing models and state-of-the-	efficiency of section-level
	somware, encrypting files	ransomware samples. The	art studies in Android mal-	fuzzy hashing depends on
	with unknown keys, mak-	study also highlights the	ware analysis.	the availability of section
	ing recovery difficult with-	use of NLP for text cleaning		information in malware
	out payment.	and extraction, enhancing		executables.
		accuracy.		

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Beyond Precision and Re-	The study compares four	The study examines fuzzy	The study compares four	The paper primarily dis-
call: Understanding Uses	popular fuzzy hashing meth-	hash families in binary anal-	fuzzy hashing methods in	cusses the use of similarity
(and Misuses) of Similarity	ods in various scenarios, in-	ysis case studies, focusing	various contexts, including	hashes in binary analysis,
Hashes in Binary Analysis	cluding shared source code	on identifying libraries, soft-	comparing programs with a	ignoring other techniques
	and statically-compiled files.	ware, and detecting pro-	large proportion of source	like signature-based analy-
	It examines low-level compi-	grams. It aims to ex-	code, connecting statically-	sis and machine learning. It
	lation process elements and	plain individual findings and	compiled files with libraries,	highlights the lack of stan-
	technicalities to explain vari-	present examples of each al-	compiling files with different	dardization and the poten-
	ances. The widely used ss-	gorithm's success or failure.	flags or compilers, and com-	tial for inconsistencies and
	deep algorithm is found to	The study addresses past	paring programs with a high	misuse. The efficiency of
	perform poorly, and the opti-	work's weaknesses by high-	part of source code. The ss-	similarity hashes depends on
	mal method is determined by	lighting unexpected results	deep method outperforms al-	the hash settings, and the re-
	the individual use case sce-	and performance variances,	ternative algorithms, but no	port doesn't provide detailed
	nario.	emphasizing that the best al-	research definitively deter-	guidance on selecting appro-
		gorithm depends on the use	mines the best fuzzy hashing	priate parameters. The arti-
		case situation.	algorithm for program sim-	cle also suggests integrating
			ilarities, with the optimum	contextual information to im-
			choice varying based on the	prove the accuracy and rele-
			use case scenario.	vance of results.

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Similarity hash based scoring	The study investigates four	The research explores the	The proposed approach for	The paper describes a
of portable executable files	hash types that are currently	use of multiple hashing tech-	malware scoring based on	method for identifying mal-
for efficient malware detec-	utilized in malware research	niques to generate a quan-	hash results improved accu-	ware variants in Windows
tion in IoT	for portable executable (PE)	titative malware score and	rate detection rates of mal-	Portable Executable (PE)
	files.It is built and developed	enhance detection rates. It	ware by more than 90%.	files, which may not be ap-
	the recommended approach	addresses the issue of high		plicable to other file formats
	for malware scoring based	false detection rates when us-		such as ELF or Mach-O. The
	on hash results. A num-	ing hashing techniques in-		method's efficiency is not
	ber of experiments are con-	dependently. The proposed		tested against obfuscated
	ducted to determine the ef-	approach, tested in experi-		malware, which can change
	fectiveness of the proposed	ments, results in a significant		the structure of the virus.
	technique.	improvement (>90%) in mal-		Section-level fuzzy hashing,
		ware detection rates.		which divides malware exe-
				cutables into binary headers
				and sections, might increase
				computational overhead,
				which may be an issue for
				real-time malware detection
				applications. Because it
				relies on a database of
				known malware families to
				categorize fresh samples, the
				method's generalization to
				unknown malware families
				is limited. The availability of
				section information in mal-
				ware executables, which may
				be purposely erased or ob-
				fuscated to avoid detection,
				determines the usefulness of
				section information.

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Security Hardened and Pri-	The methodology involves	The study presents a method	The suggested approach	The proposed Android mal-
vacy Preserved Android Mal-	identifying harmful Android	for detecting malicious	identified 56 applications as	ware detection method is
ware Detection Using Fuzzy	app behavior, extracting	Android applications using	malicious, 21 as suspicious,	based on the source code
Hash of Reverse Engineered	package names from source	static analysis with fuzzy	and 23 as benign. The	of the application, poten-
Source Code	code, matching them with	hashes and repacked danger-	studies were carried out on	tially limiting its applica-
	known malware names, and	ous code. The framework	a dataset that included 3490	tion to closed-source ones.
	computing fuzzy hashes	establishes a threshold value	malware samples from 21	Fuzzy hashing, a technology
	for matched packages. It	for malware detection, with	different families. When	known for producing false
	uses reverse engineering	a match larger than 40% in-	compared to other similar	positives, may misclassify in-
	tools like dex2jar and jadx,	dicating maliciousness. The	methodologies, the frame-	nocent apps as malware. The
	extracting sensitive Java	method outperforms other	work showed around 74% of	method's generalization to
	code features, and detecting	methods with a detection	the repacked malware.	unknown malware families
	malware using fuzzy hashes.	rate of 74%.		is limited, and code obfus-
	Static analysis is used for			cation techniques may avoid
	inspection.			detection. The evaluation
				focuses on detecting known
				malware versions without
				substantial obfuscation and
				its effectiveness against heav-
				ily obfuscated malware.

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Android Malware Classifica-	The article uses a convolu-	The study introduces a new	The proposed method for	The research presents a
tion Based on Fuzzy Hashing	tional neural network (CNN)	method for classifying An-	Android malware classifica-	method for categorizing
Visualization	to categorize malware using	droid malware by converting	tion achieved 98.24% accu-	malware samples using a
	grayscale images from the	APK samples into grayscale	racy, using a fuzzy hash-	small dataset of 2,400 An-
	Android Application Pack-	photos using natural lan-	ing approach and convolu-	droid samples. The model's
	age (APK). Natural language	guage processing techniques.	tional neural network model.	accuracy is assessed using
	processing techniques are	This method uses text clean-	This method reduced time-	precision, recall, and F1-
	used for text cleaning, extrac-	ing, extraction, and fuzzy	consuming operations and	score, but its performance is
	tion, and fuzzy hashing. The	hashing to represent decom-	outperformed existing meth-	not comparable to other An-
	decompiled smali code and	piled code as a set of hashes.	ods. Experiments were con-	droid malware classification
	manifest file are hashed us-	The method yields higher ac-	ducted on five malware va-	methods. The model's fuzzy
	ing fuzzy hashing, and the	curacy (up to 98.24%) and	rieties and their top classes,	hashing characteristics can
	CNN model is trained on	was tested on an Android	accounting for 81.12% of the	lead to false negatives, and
	grayscale images.	malware dataset containing	dataset.	its categorization findings
		15,493 samples of five differ-		are based on visual inter-
		ent malware types.		pretation of fuzzy hashing
				visualizations, which may
				include subjectivity and
				errors. The model's general-
				ization to unknown malware
				families is limited due to its
				focus on visualization and
				not subjectivity.

Paper Name	Method and Techniques	od and Techniques Contributions Results		Primary Limitation
A survey of Static Android	This study compares	The study compares	The article compares	The document A Survey of
Malware Detection Tech-	permission-based static	permission-based and	permission-based and	Static Android Malware De-
niques	Android malware detection	signature-based static An-		tection Techniques" has limi-
	strategies with signature- droid malware detection		detection methods for An-	tations
	based techniques, focusing	methods, highlighting their		
	on similarities, differences,	advantages and disadvan-		
	and accuracy. It evaluates	tages. It emphasizes the	detection has a higher True	
	machine learning technolo-	need for further research to	Positive Ratio (TPR) than	
	gies for high accuracy rates	enhance malware detection		
	and various signature-based	results. The paper also		
	methodologies based on	discusses machine learning		
	factors like tools, tech-	techniques and ROC curve	0 0	
	niques, publishing year, and	evaluation for high accuracy		
	medium. The report pro- in detecting maliciou			
	vides an in-depth analysis	droid applications, aiming		
	of static Android malware	to provide researchers with	malware detection accuracy.	
	detection strategies and a	a comprehensive under-		
	survey for researchers.	standing of static Android		
		malware detection tools.		

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
A Comprehensive Review	The study explores mal-	The study explores emerging	The study explores various	The research explores mal-
on Malware Detection Ap-	ware detection tactics and	technology developments in	malware detection methods,	ware detection methods,
proaches	new methods, including	malware production and de-	highlighting the challenges	highlighting the challenges
	behavioral detection, data	tection, examining various	in detecting complex mal-	in identifying complex
	mining, stack computation	methods, their benefits and	ware and the need for new	malware and the need for
	tree predicate logic (SCTPL),	drawbacks, and the likeli-	approaches. It introduces the	innovative approaches. It in-
	model-checking algorithms,	hood of identifying malware.	stack computation tree predi-	troduces SCTPL for describ-
	and heuristic-based detec-	It explores behavioral de-	cate logic (SCTPL) for char-	ing dangerous behaviors
	tion algorithms. It provides	tection systems, algorithms,	acterizing harmful behav-	and presents a pushdown
	insights through literature	and heuristic-based detec-	iors and provides a model-	system model-checking
	research, algorithm creation,	tion systems. The goal is to	checking algorithm for push-	algorithm, recommending
	and experimental results	help researchers understand	down systems. The method	further investigation for
	analysis, highlighting the ef-	malware detection technolo-	is evaluated for accuracy	improved classification.
	fectiveness of these methods	gies better and assess their ef-	and efficiency in identifying	
	in detecting malware.	ficiency.	Android malware families.	
			Further analysis is recom-	
			mended for more reliable re-	
			sults and improved classifica-	
			tion.	

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
A Survey of Malware Analy-	The paper conducted a sur-	The research presents a new	The poll indicates potential	The study on community de-
sis Using Community Detec-	vey using manual keyword	taxonomy for malware anal-	for advancement in malware	tection methods in malware
tion Algorithm	searches on popular search	ysis solutions using com-	analysis using community	research lacks a comprehen-
	engines like IEEE, ACM, Sco-	munity detection algorithms,	detection algorithms and ad-	sive overview of all method-
	pus, and Google Scholar to	comparing previous studies	dressing research gaps. The	ologies and empirical exami-
	identify important publica-	and revealing variations. It	study provides taxonomy	nation, making it difficult to
	tions on malware, commu-	examines various malware	classification solutions for	assess their practical useful-
	nity identification, and cyber-	analysis solutions, their ben-	malware detection, classifica-	ness. Additionally, it fails to
	threat. The articles were cat-	efits and drawbacks, classi-	tion, cyber-threat infrastruc-	address privacy and ethical
	egorized into malware anal-	fication criteria, primary du-	ture detection, and feature se-	issues associated with com-
	ysis, community detection,	ties, and function. The study	lection for workstation and	munity detection algorithms,
	and malware analysis using	also identifies challenges and	mobile platforms, address-	suggesting the need for fur-
	algorithms. Unsupervised	suggests future research top-	ing analysis tasks, approach,	ther research to fully under-
	techniques were also used	ics in malware analysis us-	and feature source.	stand these strategies.}
	in community discovery ap-	ing community detection al-		_
	proaches.	gorithms.		

Paper Name	Method and Techniques	Contributions	Results	Primary Limitation
Strengthening Cloud Secu-	The study presents an adap-	The paper presents an adap-	The study evaluates the ef-	The study presents a frame-
rity: An Innovative Multi-	tive multi-factor multi-layer	tive multi-factor multi-layer	fectiveness of an MFA frame-	work for secure cloud system
Factor Multi-Layer Authenti-	authentication framework	authentication architecture	work and algorithm in de-	authentication, but its assess-
cation Framework for Cloud	for cloud user authentication,	for cloud user authentica-	tecting suspicious users and	ment is theoretical and lacks
User Authentication	incorporating access control,	tion, aimed at enhancing	preventing malicious attacks	specific performance or scal-
	intrusion detection systems,	security and preventing data	on cloud servers or services.	ability analysis. The multi-
	and automated method	breaches. The framework	The results show low false-	factor multi-layer authentica-
	selection. It enhances	uses various authentication	negative rates for varied user	tion strategy may add com-
	identity verification using	factors, geolocation, browser	numbers and efficient detec-	plexity to consumers, and
	user factors, geolocation,	confirmation, and AES-	tion of potentially harmful	the study does not explore
	and browser confirmation.	based encryption to detect	users and intruders. The	options for streamlining the
	An AES-based encryption	malicious users and intrud-	study emphasizes the impor-	process without compromis-
	component protects login	ers, preventing deliberate	tance of authentication tech-	ing security. The behav-
	information. The framework	attacks on cloud services and	niques in detecting legiti-	ioral analysis component re-
	measures performance using	data.	mate users and preventing	lies on user behavior profil-
	false-positive and false-		intrusions, thus enhancing	ing, which may be sensitive
	negative rates, reducing		the overall security of cloud	to changes in user behavior
	cloud-related risks like data		platforms.	or data gathering discrepan-
	leakage, brute force assaults,			cies.
	complexity, security, privacy,			
	and execution time.			

#### 4. Proposed Framework

The identification of malware variants and linking them to their respective Advanced Persistent Threat (APT) campaigns has proved to be a difficult problem that requires a sophisticated solution. This research proposes an efficient solution that leverages the capabilities of ASCON, a lightweight and robust cryptographic algorithm, along with fuzzy hashing techniques to analyze the structure and behavior of malware. The proposed solution targets to efficiently dissect and categorize malware through calculated hashes of various sections of a PE (Portable Executable) sample.

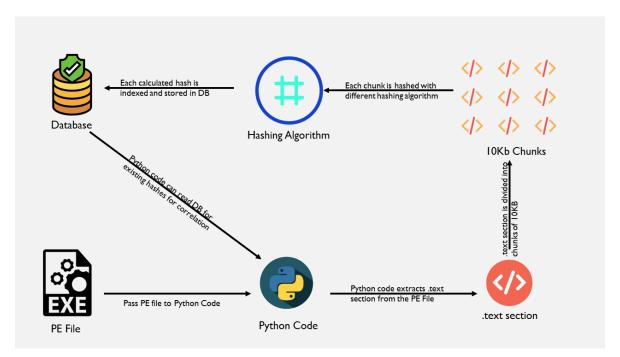


Figure 2. Proposed Framework

The proposed system integrates fuzzy hashing, an advanced technique that computes the cryptographic hash value for digital data, and produces a digital fingerprint which efficiently identifies similar but not identical data items. We propose to use this technique on PE samples, computing hash values for their different parts to allow identification and linking of different malware versions.

ASCON is a cryptographic algorithm that proves to be lightweight, highly secure, and notably, efficient in constrained environments. Thus, it emerges as an ideal candidate for this proposed system. ASCON-hashed PE segments can provide unique and versatile identifiers to link them with specific APT campaigns or identify variant versions.

The proposed system requires the division of the PE sample into distinct parts or sections. These divisions may be based on the functional or structural characteristics of the malware. Each portion is then hashed using the ASCON cryptographic algorithm separately rather than hashing the entire PE sample. This approach not only aids in identifying slight variations within a sample but also improves the efficiency of the analysis process.

By properly leveraging the proposed hashing method, similarities and linkages between different malware samples can be more effectively identified. The matching of such hashes can allow detection of reused components, variant versions, or even derivation patterns between different APT campaigns. Analysis of these patterns and components can then assist in tracing the origin of the malware or providing insights into the possible directions for the development and evolution of these threats.

The proposed solution's practicality and effectiveness are anticipated to be high, given the careful integration of fuzzy hashing and ASCON capabilities. The approach could significantly empower malware analysis, not only providing a more efficient method to handle a large volume of data but also offering a detailed dissection of malware structures and behaviors. This research provides a

robust foundation for further investigation into the use of such methods in the rapid identification and classification of malware threats within increasingly complex cybersecurity landscapes.

## 5. System Algorithm

# 5.1. Function Analyze\_PE\_file(file\_path: str)

- Read the PE file specified by file\_path.
- Extract the .text section of the PE file.
- Divide the .text section into chunks of 10KB.

#### For each chunk:

- Calculate SHA-256 hash of the chunk.
- Calculate SHA-3 hash of the chunk.
- Calculate BLAKE2s hash of the chunk.
- Calculate Skein hash of the chunk.
- Calculate MD5 hash of the chunk.
- Calculate ASCON hash of the chunk.
- Save the calculated hashes along with metadata (e.g., file name, chunk index) into the SQL database.
- Index the SQL database for efficient retrieval.

# 5.2. Function Process\_Directory(directory\_path: str)

- List all PE files in the specified directory.
- For each PE file in the directory, call analyze\_PE\_file function.

# 5.3. Function Compare\_with\_Database(file\_path: str)

- Read the PE file specified by file\_path.
- Extract the .text section of the PE file.
- Divide the .text section into chunks of 10KB.

## For each chunk:

- Calculate SHA-256 hash of the chunk.
- Calculate SHA-3 hash of the chunk.
- Calculate BLAKE2s hash of the chunk.
- Calculate Skein hash of the chunk.
- Calculate MD5 hash of the chunk.
- Calculate ASCON hash of the chunk.
- Query the SQL database for matching hashes.
- If a match is found, log or report the match along with relevant metadata.

# 5.4. Main Program

- Accept user input for either analyzing a single PE file or processing a directory.
- If analyzing a single PE file, call analyze\_PE\_file function.
- If processing a directory, call process\_directory function.
- If comparing with the database, call compare\_with\_database function.

#### 6. Results and Discussion

Our comprehensive analysis of over a thousand malware samples utilizing ASCON-based fuzzy hashing emphasizes its superiority in terms of speed and efficiency. While it is designed for speed and security, its performance was also relative to the capabilities of other algorithms, as seen in the table.

ASCON-Hash, within the range of 224 to 512 bits output size, typically demonstrated 4-6 cycles per byte. Its design for speed and security played out in the field, making it a formidable baseline standard for hashing speeds. It looked particularly efficient when contrasted with the other hashing algorithms.

SHA-256, a widely used and standardized algorithm, showed a 1.33x - 2x slower performance than ASCON. It underwent 8-12 typical cycles per byte with a singular output size of 256 bits. SHA-3, the NIST SHA-3 competition winner known for its good security, exhibited a markedly slower speed ranging from 2x - 2.67x slower than ASCON despite having the same output size as SHA-256.

Though BLAKE2s has been noted for being fast and secure, especially for its use in cryptocurrencies, it showed a 1x - 1.33x slower performance than ASCON. BLAKE2s carries out 6-8 typical cycles per byte and has an output size of 256 bits. Skein, attributed with a high degree of configurability and a focus on security, performed 1.67x - 2.33x slower than ASCON.

Interestingly, the oldest algorithm in the lot, MD5, demonstrated 0.67x - 1x faster speed than ASCON but at the glaring cost of security. This legacy algorithm with known vulnerabilities operates with beats of 4-6 cycles per byte and hands out a lower output size of 128 bits.

In conclusion, the results illustrate the balance ASCON offers between speed and security that many other hashing algorithms do not appear to match. Its performance serves as a baseline to validate the efficiency of other hashing functions, revolutionising malware analysis in terms of speed and security, hence marking it as a fitting choice for real-world applications.

Algorithm	Output Size (bits)	Typical Cycles per Byte	Relative Speed to ASCON
ASCON-Hash	224-512	4-6	lx
SHA-256	256	8-12	1.33x - 2x slower
SHA-3	256	10-13	2x - 2.67x slower
BLAKE2s	256	6-8	lx - 1.33x slower
Skein	256-512	10-15	I.67x - 2.33x slower
MD5 (legacy)	128	4-6	0.67x - 1x faster (insecure)

Figure 3. Hashing Results

#### 7. Future Work

In future we can develop a GUI-based tool with linked database for long-term memory for historical tracking of malware versions as well inducing visual charts for better threat intelligence. In addition, we can integrate machine learning algorithms to classify different malware samples based on hashes of different sections of the sample.

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