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Block chain based Malware Detection using Machine Learning Algorithms for IoT enabled E-Health Applications.

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Abstract:

Due to increasing digitalization and the development of new technologies such as the IoT, the application of machine learning (ML) algorithms is rapidly expanding (IoT). ML algorithms are being used in healthcare, IoT, engineering, finance, and other fields in today's digital age. However, in order to predict/solve a specific issue, all of these algorithms must be taught. There's a good chance that the training datasets have been tampered with, resulting in skewed findings. As a result, we have suggested a blockchain-based approach to protect datasets produced by IoT devices for E-Health applications in this paper. To address the aforementioned problem, the suggested blockchain-based system makes use of a private cloud. For assessment, we created a mechanism that dataset owners may use to protect their data.

Keywords: Blockchain, IoT, Machine Learning.

Introduction

Most of the existing healthcare systems are built on centralised servers that require authorization from many organisations in the network in order to access the medical data. A delay in providing medical treatments adding to the potential for data leaking might result from all this. It's very uncommon for people to have no idea who is keeping and utilising their medical data without permission in these types of healthcare systems.

Machine learning and possible technologies have been widely used [1] in various research and development sectors. As a result of this massive utilisation, machine learning-based forecasts have become increasingly important in decision-making. The healthcare industry confronts numerous problems as the population continues to expand. It's indeed obvious that these questions need to be answered, that's what this article tries to accomplish [2]. IoT development teams emphasize resources allocation, safety, and internet connectivity. Information obtained by these apps is vast and constantly changing, making it more difficult to store in databases, and its conveyance poses safety problems. Well with help of BT's distributed system, we can address such obstacles [3]. Additional concern is that the machine learning model might be modified to provide beneficial outcomes in the health

industry. Machine learning methods' capacity to analyse data and identify data patterns makes them more vulnerable to different sorts of assaults [4].

Due to its excellent properties, including such confidentiality, responsibility, data integrity, and secrecy, to mention a few, blockchain technology has attracted much interest. Since most IoT assets have essential features that render them prone to security attacks [21]. Healthcare is amongst the most rapidly growing sectors [22, 23]. Smarter fitness centres use IoT sensor-based fitness trackers that are placed on the learner to monitor their strenuous activities. Like the way instructors support student training, the Internet of Things is changing the fitness sector. In addition, IoT fitness sensors tackle a number of important concerns, including muscle strain, risk events, and training duration [24]. A broad range of applications has embraced machine learning in recent times [25]. To construct and forecast the future, these methods focus on data analysis to uncover underlying knowledge and useful information from previous data [6]. AI and IOT combine to optimize the performance of IoT devices [29]. There is an IoT paradigm by Lee et. al. that is user-oriented and utilises 2 distinct sorts of methods [28].

Intelligent Things (IoT) has become a sector of tremendous influence, opportunity, and development with the expectation that there will be more than 50 billion linked items by 2020 [34]. For

IoT applications, Blockchain (BC), which was initially successfully utilised in cryptocurrency, has the lot of quality and potential high-security technology that protects your privacy [35, 36]. In the IoT, the use of blockchain can eliminate there is just room for one mistake and provide a safe and efficient way of storing and processing IoT data [37]. There is a risk that an IoT system will be slowed down due to a huge quantity of data which is produced by a big amount of IoT devices [38].

Furthermore, the supplier does not provide real-time upgrades, leaving the system exposed. Infrastructure (such as sensing devices and IoT gateway) and cybersecurity solutions can benefit from machine learning approaches. These systems may evaluate network traffic based on the current cyber-threat information, upgrade risk information database, and defend the underlying systems from new assaults.

The Internet connects and controls all smart gadgets, though. In an IoT-enabled industrial multimedia context, it also poses a serious danger to the communication. Heavily prone to replay, man-in-the-middle and impersonating threats; data leak; sensitive data alteration; and virus insertion. To protect such an ecosystem against different sorts of assaults, it is necessary to restrict interaction inside it. It's common to see machine learning techniques being used in health, IoT, technology and banking in today's world. In ability to forecast a certain issue, all of these methods must be trained. Machine learning techniques are vulnerable to malicious assaults in which non traceable modifications are inserted into the raw data, resulting in incorrect projections of the outputs, misleading machine learning systems [39].

Block chain:

A blockchain can indeed be viewed as a decentralised architecture with built-in security to boost transaction trust and integrity [5]. Following the Internet, blockchain has transformed the flow of information and media. As a result, blockchain technology is viewed as a paradigm-shifting invention and the precursor to a new economic era [7]. One way to think of blockchain is as a relational network that consists of ordered blocks, where the committed blocks are immutable [11]. Several researchers have utilised blockchain in the construction of an eHealth system, although it is not well known. Researchers from Malamas and colleagues employed Blockchain technology to create a framework for medical devices that may have the ability to serve as forensics tool [40]. Using smart contracts on Blockchain, the

system allows for fine-grained permission. Blockchain, according to Mytis et al. [41], ensures the integrity and non-repudiation of information retrieved from a conventional biological database. Using a lightweight Blockchain, Ismail et al. [42] have suggested a healthcare architectural architecture. There are several functions assigned to BC nodes that are based on geography. However, IoT devices are generally underpowered. On the whole, some research was discarded since it was largely concerned with technical elements of blockchain technology and/or its architectural design

By sharing database records among many users active in the blockchain network, a blockchain improves on the traditional centralised architecture [5]. Smart contracts will decide the blockchain economic model rules wherever stimulatory operations are enforced autonomously [7]. There are essentially three types of blockchain technology [6].

- a) Public blockchain
- b) Consortium blockchain
- c) Private blockchain

Apps are used in nearly every business for mapping, tracing, and speeding up commodities flow. Data is encoded on a blockchain and is moving forward with fresh information as it moves past by the supply chain [8, 9, and 10]. Through hardware and software solutions, blockchain technology may also be utilised to improve security and dependability in distributed networks [18, 19, 20]. There have been three generations of blockchains depending on the desired audience (Zhao et al., 2016): Blockchain version 2 (which incorporates SCs) and a programming environment that go well beyond bitcoin transactions; and Blockchain version 3 (which comprises in areas beyond the previous two versions, such as e-commerce) [11]. It's true that there are a few studies focusing on the function of blockchain in IoT development [12, 13] and large data management in a decentralised way [14]. A method to medical data accessibility on the Blockchain was developed by Ramani et al. [17]. As long as a patient consents, it's possible to get medical treatment professionals to add and retrieve patient health data. There was idea of a separate blockchain which was examined for security evaluations. They did not, however, use a simulator or create a prototype to test the performance of the suggested approach.

Access Control:

Unwanted cyberattacks monitoring and prevention are significant challenges in IoE. An access control system that requires two factors to verify each other before generating a private key for safe interactions

can help overcome these difficulties, as per the authors. The sensing data of different smart devices in an IoE system is analyzed safely at surrounding fog servers, and authorized customers can also obtain the real-time data straight from authorized smart devices via access control mechanisms [69].

As a critical component of cybersecurity, access control limits what actions users may do on services. When working with e-Health services and information, access control is essential for determining which data is provided to which client. As a result, third-party access to these data resources must be controlled. A recipient's right to access resources in a system is controlled by access control, which determines which actions are permitted for particular users [72]. Controlling who may see your health information, whether it's your private or medical information, is not a simple process, especially since other parties (such as insurance companies or physicians) may need accessibility to it for various reasons (such as billing). We are still working towards unified and integrated electronic healthcare systems (e-health), which makes this a particular issue [73].

Access Control System (ACL) are among the most prevalent techniques. As the name suggests, an ACL is a list of topics that may access an item, together with their level of access (or permissions). Access control problems have been solved in e-Health situations utilising DLT and, more especially, blockchain thanks to the extensive use of these technologies. According to Maesa et al. [74], ABAC may be implemented with the help of distributed ledger technology using the XACML reference model.

Moreover, the whole access control method is constructed on a blockchain-based identity management system in a trustless IoT environment, which provides safety and confidentiality for users. Blockchain is used in Concentrate to capture ownership identities and preserve them on trustless peer-to-peer systems in order to maintain a collect of ownership names. These identified owners may then apply access control policies to objects within the same control domain, and the connections between objects could create a feedback mechanism to impact the formation of access policies.

Machine Learning:

As the name suggests, machine learning (ML) is the subject of research that focuses on creating applications that learn via experience. Without directly programming it [76], it is the capacity to

educate a computer [77]. In addition to philosophy, information theory, probability and statistics, control theory, psychology and neuroscience, computational complexity, and artificial intelligence [77], ML draws on a wide range of fields. It's important to note that ML algorithms are application-specific. In machine learning, there are a variety of algorithms, including supervised, semi-supervised and unsupervised. Numerous advantages may be gained by utilising machine learning-based systems. Large amounts of training data may be used to train them, and then they can aid clinical practise in identifying risk and devising therapy using inductive inference. Because they do not require human intervention, these technologies can help decrease the risk of human mistake [79].

Accordingly, machine-learning techniques are utilised to provide reliable results from huge complex datasets, and the resulting outcomes may be used to anticipate and identify flaws in IoT-based devices. To tackle data security problems, current IoT systems are increasingly using Blockchain (BC) technologies. As a result of this, ML algorithms and BC methods have been studied in depth. Hence the need for a comprehensive assessment of initiatives done in previous decades to tackle both safety and privacy challenges utilising combined Machine learning and Blockchain methods.

In order to train a machine-learning system, information is necessary. Use past patient information to determine predictions about new patients, for instance, patients are hesitant to provide their data because they are worried about their security. This has been addressed in the research [81, 81, 82]. Using non-linear kernel SVM, the scientists developed a novel application framework e-Diag to accurately categorise patient info, while maintaining user data and services provider's models confidentiality [82]. According to the report, previous investigations had employed HE methodologies that were unsuitable for online medical prediagnosis. Medical insurance expenses can also be predicted using machine learning [84, 85]. A study has also been conducted on the current use of machine learning and blockchain in healthcare, as well as its applications, difficulties, and privacy concerns [86, 87]. It's been shown that certain machine learning techniques may be highly useful in minimising security and privacy threats. In the sections that follow, we'll go into more depth about each of these techniques.

Using a method called re-linearization, Sun et al. [83] presented a better version of completely

HE that reduces the size and distortion of the additive cypher text. Other features included a private hyperplane decision-based classification, a private Naive Bayes categorization, and a private decision-trees comparability. As part of an earlier research project, the same authors succeeded in cutting user-server loops in half, without sacrificing privacy or security

Malware Detection:

The categorization features comprise the permissions required by the programme (given by the uses-permission tag) and the components in the uses-features group. They utilised classification technique to categorize Android systems into benign and malicious programs [89]. Bot-net assaults (malware attacks such as mirai and reaper) have recently drawn the interest of researchers. When IoT-enabled industrial multimodal environments are attacked, the connection is interrupted. Aside from that, the hackers may be able to remotely manipulate the smart gadgets and alter their functionality. The existence of malware assaults in such an environment requires a powerful detection method. To identify malicious in an IoT based multimedia for industrial applications environment using a machine learning techniques reserached is being done [90]. In order to find virus, it uses algorithms to observe behaviour of the system and evaluate it as either regular or unusual. Instead than relying on patterns or signatures, categorization is typically based on machine learning algorithms that employ heuristics and rules rather than patterns [91]. Some of the key reports found in the research are:

- Basic reports providing a summary or trend of dangerous software detection, as well as the system and the consequence (cleaned or left alone), are an excellent place to start when analysing malware detection trends [92].
- As a result of this, several organisations have been able to avert catastrophic harm by logging "leave-alone" occurrences from anti-virus programmes [92].
- It's important to keep track of all failures in anti-virus protection since today's malicious software is highly prepared to defeat anti-virus programmes [92].
- By having internal connections to known malware IP addresses, a company's vital data will be protected from being stolen by malicious software operators. This report may be created using logs (such as firewall logs) and a public blacklist of IP addresses

[92].

- The "Bottom 10" (as opposed to "Top 10") data might help you determine which types of malware are the least common in your company [92].

Malware Detection Using IoT:

Based on the kind of technique, IoT malware detection approaches may be divided into two primary categories: dynamic and static analysis. In the dynamic method [64], executables are monitored and aberrant activity is detected. Most contemporary malware, according to Alex et al. [65], is formed by copying the source code from the internet, or a version of the dangerous code written by the malware author. This study provides a brief overview of the current development and growth of IoT malware by analysing, evaluating, and synthesising numerous research such as [66], [67], and [68] as well as manually analysing several IoT malware samples. Furthermore, the research investigated probable security flaws in Iot systems for four levels, including sensor, middleware, application, network.

There are several possible benefits and incentives for building an IoT framework that is built on a BC-based architecture [78]. The centralisedIoT approach is vulnerable to DDoS assaults, according to the study. There's also a point of failure in its design, which puts the reliability of IoT services in danger. Current IoT security mechanisms are centralised since they rely on third-party security services, which creates data integrity concerns for IoT devices and systems. Ali et al. showed how BC-based IoT security solutions may overcome all of these concerns. IoT cyber risks have been addressed with Different classifiers and BC methods, but integrating these two is something new that has to be investigated.

According to Alsunbul and coworkers [70], they developed a network security system that detects and prevents unwanted access attempts by replacing the regular protocol with a new one that is generated on the fly. Cloud computing security solution by Chang and Ramachandran [71] to guarantee that only authorised and authenticate the firewall and controlling the access are the first security layers that protect the network and information.I. As a second layer,maintenance of user identities and intrusion detection are employed in order to re-identify users and eliminate any malware. A top-down security strategy is provided via convergent encryption, the third layer.

Blockchain and AI in IoT:

In recent years, ninety percent of the world's data has been generated (IBM, 2017). a) The IoT and b) Population Growth (Stats, 2017). While the blockchain and IoT technologies consists a huge amount of potential on their own, their symbiotic connection opens up many more possibilities. Among many other things, Atlam et al. [29] provided an outline of the Internet OT and AI , as well as possibilities and rewards in diverse AI-based IoT applications. He also explained blockchain technology, which is classified into 3 sub-categories: blockchain feature and its classification; blockchain usage; consensus mechanism; and current problems. On the Ethereum platform, Wright et al. [30] introduced an intelligent agreement that relies on intelligent border. For the verification of edge devices in the transaction trade, the proposed platform provides accurate offload computation on nodes.

With the assistance of Salah et al. [31], researchers analyzed the resilience and efficacy of blockchain-enabled AI as well as open problems that make application of artificial intelligence in the Blockchain. The authors also discuss blockchain uses, as well as outstanding concerns, while focusing on AI. On the basis of the following levels of IoT study, Qian et al. found that blockchain technology offers stability for various open study and IoT gadget challenges. To identify abnormal network behaviour, this process utilizes identity verification and ML algorithms [32]. Almost all of the research cited above focuses on blockchains that are connected with AI. In terms of speed, delay, and safety, these investigations have a long way to go.

Increasing network capacity raises the likelihood of an attacker. A conventional network such as a company's office has less problems than an IoT network. In contrast, IoT devices that communicate with one other are generally multi-vendor gadgets that use multiple standards from various manufacturers.

In concerns to documenting financial transactions, blockchain was initially designed to do just that (e.g., Bitcoins and other cryptocurrencies). Since all transactions are visible, any alterations may be tracked and identified with relative ease." Cryptocurrencies such as blockchain can be used to improve IoT security. We'll now look at two instances of how blockchain may be used to secure IoT devices. When a transaction is performed, a block is produced. The block is sent to the nodes in the network, which is broadcast to all nodes. In

bitcoin, one of the nodes validates the block and present it to the network. It is only added to a node's chain of blocks if the block is validated and refers to the prior block in the right manner [75].

AI enabled Blockchain malware detection using IoT in E-Health Applications:

A method to medical data accessibility on the Blockchain was developed by Ramani et al. [17]. As long as a patient consents, it's possible to get medical treatment professionals to add and retrieve patient health data. In order to assess security, the notion of a private blockchain was considered. They did not, however, use a simulator or create a prototype to test the performance of the suggested approach. Due to its relevance in overcoming the compatibility and safety problems of the EHR and EMR technologies in eHealth, blockchain technology has experienced a surge in the health industry. We will be presenting a paper assessment of current research articles on block chain connected to e-health and discussing prospective directions for future research and patterns that may be pursued using this tech as a focal point in the future [33]. Numerous advancements, such as IOT, Internet of Healthcare Devices, Collaborative Training, etc. have contributed to a dramatic increase in electronic information created by Internet - of - things based gadgets in the health industry. In order to diagnose patients in a timely way, machine-learning techniques are essential. To protect medical datasets collected by Iot devices in health apps, this study proposes a methods based on blockchain [39].

With the help of blockchain, these techniques may be made more secure and reliable. An eHealth (mobile health) mobile phone software for cognitive behavioural treatment for sleeplessness was developed and evaluated by a group of researchers [62]. As part of the application, patient health records is sent to a blockchain network. Because of the characteristics of the blockchain, the EMRs in the system were safe and impervious to manipulation after validation, and the patient had access and control over the info [63].

A few of the services covered by distant medication care include constant vital signs tracking with smart or implanted devices sensors, arrhythmia identification and fall detection, implantable cardioverter regulation and tracking of expectant mothers as well as cancer treatment response tracking and glycogen measurement [43]. Due to blockchain's ability to solve important problems such as automated claim validation [53] and public health

management [54], the healthcare industry is a good contender for blockchain technology [50, 51, 52]. For collecting and storing data, the vast majority of IoT applications are centrally controlled. Consequently, these devices are prone to a data loss, especially when dealing with a massive amount of end-to-end interactions [44]. Traditional electronic medical systems can be crippled by malware such as Ransomware and Demand of Service (DoS) intrusions [45].

For example, experts have used blockchain technologies to overcome privacy issues about the administration of medical information and healthcare facilities, as well as safety and third-party faith problems. In addition to network access, safe storage, and exchange of medical information without the need to rely on private entities or mediators, blockchain can also protect the personal data [46, 47]. It has been intended to handle health data independently using blockchain-based ehealth frameworks [43, 48, 49].

Three primary constituents make up healthcare: (a) core providers of medical care services, such as physicians, nurses and hospital administrators; (b) critical services associated with medical care services, such as medical research and health insurance [59]; and (c) beneficiaries of medical and health-related services, such as patients or the public. For the purpose of promoting, maintaining, or recovering beneficiaries' health, the healthcare system is described as contact-based and technology-based remote monitoring services provided by constituency service providers [60, 61].

We require a decentralized network to provide resilience and sustainability, and to minimize several more traffic [55]. Information that has not been maintained on the user's computer or in cloud services must be moved to block chain networks. As a result, we utilize a minimal digital signature [56] technique to verify that information is not changed. In order to process IoT massive data, we require cloud servers. There is a risk, though, that it will create an issue with trustworthy third parties. So we record all activities in separate areas and then construct a composite hash of the each block to achieve this goal. With electronic signature and a minimal Ring system [56], we're able to maintain privacy. It is possible to sign secretly using a ring signature because the sign is mingled with the other individuals (the ring), and that no one (excluding the signatory) recognizes who inked it. We use a double encryption algorithm to secure sensitive information from attackers. Securing the same content with two

keys is not what is meant by the double cryptography here, but rather encryption techniques followed by encryption of key that was used for encryption. Use ARX techniques to protect messages, then cipher the key using recipient public key. Defensively, we use Diffie-Hellman key exchange to send the shared key, which makes it very difficult for an adversary to obtain those [57, 58].

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