

COUNTERFEITING THE PHARMACEUTICAL INDUSTRY USING THE PERSPECTIVES OF BLOCKCHAIN TECHNOLOGY

Rinku Raheja, Assistant professor, National Post Graduate College,
rr_141085@yahoo.co.in,

Mahesh Kumar Tiwari, Assistant Professor, National Post Graduate College,
maheshyogi26@gmail.com,

Saurabh Saxena, Assistant Professor, Institute of Technology & Science,
Ghaziabad, mr.ssaxena@gmail.com,

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ABSTRACT

The issue of counterfeit medications is a pressing worldwide concern that is becoming worse, especially for underdeveloped countries. Today, the yearly economic costs associated with pharmaceutical counterfeiting reach billions of dollars. A significant contributing factor to drug counterfeiting is the inadequacy of the current pharmaceutical supply chain system. The journey of drugs from manufacturers to wholesalers, distributors, pharmacists, and ultimately to customers lacks sufficient information sharing. This lack of transparency leads to manufacturers having limited visibility into their products' locations, regulatory authorities facing challenges in oversight, and cumbersome, expensive product recalls. Additionally, tracking patients effectively within the existing system is a persistent challenge. In this article, we've looked at how Block chain technology may be used in the pharmaceutical supply chain to improve visibility, provide traceability, and strengthen security.

The proposed system is tailored for implementation in the pharmaceutical industry, aiming to track pharmaceuticals from production through delivery to the patient. After drug administration, the system will meticulously document its impact on the patient, facilitating future statistical analysis. A restricted-access block chain will serve as the repository for transactions, exclusively allowing trusted entities to join the network and contribute data securely.

1. INTRODUCTION

The realm of Pharmaceutical Research & Development is an intricate journey that spans numerous years, encompassing drug discovery, development, and the rigorous process of regulatory approval. Once this intricate journey concludes and a standardized product takes shape, manufacturers encounter the subsequent challenge: delivering this product to the intended customer in its original, unaltered state. Their main goal is to guarantee that the consumer gets a genuine product made by the authorised producer instead of one obtained from a fake source. Nevertheless, the current Supply Chain Management (SCM) system within the pharmaceutical industry is outmoded. It falls short in delivering the essential visibility and control required by both manufacturers and regulatory authorities concerning drug distribution. Furthermore, it lacks the resilience necessary to fend off cyber-security threats in the 21st century. The prevailing state of Supply Chain Management (SCM) engenders the manufacturing, distribution, and consumption of counterfeit drugs. This alarming proliferation of counterfeit pharmaceuticals poses a severe and escalating public health hazard, particularly in developing nations, amplifying the gravity of this global issue. Counterfeit drugs have a twofold, highly detrimental impact on health, both through indirect and direct mechanisms. Indirectly, these deceptive medications often lack the prescribed dosage or essential active agents necessary to combat diseases. This shortfall paves the way for the development of drug-resistant strains, rendering even authentic drugs ineffectual in the long run. On a more direct note, counterfeit medications may contain active ingredients, but these components can exist in quantities that are either too low or excessively high. Additionally, the production of these ingredients might be marred by impurities, introducing toxic elements into the equation. In such cases, the use of these counterfeit drugs can lead to exceptionally severe health complications. Counterfeit drug manufacturers sometimes go as far as replicating the branding of legitimate

pharmaceutical companies to produce counterfeit products that mimic everyday items, which, while problematic, generally pose lower health risks. However, a far more troubling scenario unfolds when these counterfeit operations extend their reach to essential medications used in the treatment of conditions like cancer, pain management, cardiovascular disorders, antibiotics, contraceptives, and other prescription drugs. The repercussions of this deceptive practice can lead to exceptionally grave consequences for patients. As per the most recent World Malaria Report, the year 2021 recorded approximately 247 million malaria cases, a slight increase from the 245 million cases reported in 2020. In terms of estimated malaria-related fatalities, there were 619,000 deaths in 2021, in contrast to the 625,000 deaths documented in 2020.

Disruptions linked to the COVID-19 pandemic resulted in an extra 13 million instances of malaria and around 63,000 more malaria-related fatalities throughout the course of the two pandemic-affected years, 2020 and 2021. It is interesting that the WHO African Region still has a disproportionately high number of cases of malaria worldwide. Approximately 95% of all cases of malaria and 96% of fatalities from malaria occurred in this region in 2021. Additionally, a considerable percentage of these deaths—roughly 80%—occurred among children in the Region who were younger than five years old. In less developed countries, every tenth medicine taken by consumers is discovered to be counterfeit and of lower quality, according to a World Health Organisation (WHO) assessment.

The Global Count of Counterfeit Occurrences in Pharmaceuticals:

Below image is showing the rise of counterfeit incidents since 2011.

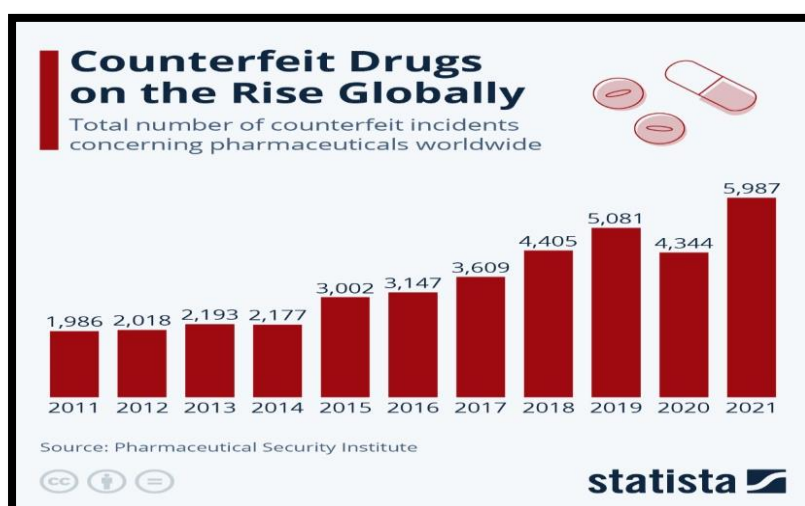


FIGURE.1 THIS CHART SHOWS THE TOTAL NUMBER OF COUNTERFEIT INCIDENTS CONCERNING PHARMACEUTICALS WORLDWIDE.

To combat the proliferation of counterfeit drugs, the pharmaceutical industry requires an effective supply chain management system. The optimal solution for creating an ideal SCM system is the implementation of Blockchain technology.

2. WHY BLOCKCHAIN?

Blockchain operates as a decentralized network of computers, where connected participants collectively uphold a secure ledger of transactions, all without relying on a central server. Each transaction is meticulously recorded in the network, accompanied by a timestamp, effectively eliminating the need for third-party intermediaries. Within this network, each node independently maintains its copy of the ledger, ensuring that even minor modifications to the local ledger are synchronized across the entire network. Consequently, all nodes on the network promptly update their individual ledgers to reflect the most recent transactions. Even in the case of a node failure or disconnection, the network remains unaffected. The primary justification for the incorporation of Blockchain technology into systems such as pharmaceutical supply chain management is its unparalleled security. Blockchain serves as the foremost solution for providing 21st-century cyber-security, and notably, it has maintained an impeccable record of no breaches. Its design is fundamentally geared towards preventing any single entity from manipulating data and transactions, thereby fostering trust and eliminating the biases often encountered in traditional supply chain systems. Utilizing blockchain technology, stakeholders can securely exchange digital assets anonymously. There's no need for them to be familiar with or have faith in one another, nor must they involve a third party in their transactions. Hence, blockchain emerges as the optimal choice for instilling trust within an inherently distrustful environment. Another compelling reason for advocating blockchain in pharmaceutical supply chain management systems is its unmatched capacity to meticulously document the journey of a product throughout the entire supply chain. Whenever the product changes hands, a new transaction is added to the blockchain. By retaining the complete history of a product's journey, it becomes effortless to uncover its actual source and key milestones. This method introduces greater openness into transactions within the pharmaceutical supply network. In the realm of the pharmaceutical supply chain, when a manufacturer generates a product and enrolls it within the network, it assumes the role of a digital asset, subject to transfer to other stakeholders in a manner akin to cryptocurrencies. Concerning privacy safeguards, blockchain takes the forefront yet again. Within a blockchain network, each participant is allocated a key-pair as their identifier. Private participant data is meticulously safeguarded, with each transaction conducted from one participant's

public key to another's within the pharmaceutical supply chain management system. Moreover, in this system, patients' private data remains confidential, while their medical records are accessed publicly but anonymously. Incorporating blockchain technology into quality control and the identification of counterfeit drugs is a significant step toward enhancing safety and ultimately preserving lives.

The ultimate and most compelling rationale for employing blockchain within pharmaceutical supply chain management is the integration of Smart Contracts. A Smart Contract is a segment of code that encapsulates the actual rights and obligations, including the terms and conditions for payment and delivery of goods and services, as agreed upon by all parties involved, and it can be executed automatically. Smart Contracts bring an added layer of intelligence and potency to blockchain technology, enabling the creation of state-of-the-art, customized blockchain-based systems that are on the cutting edge of innovation.

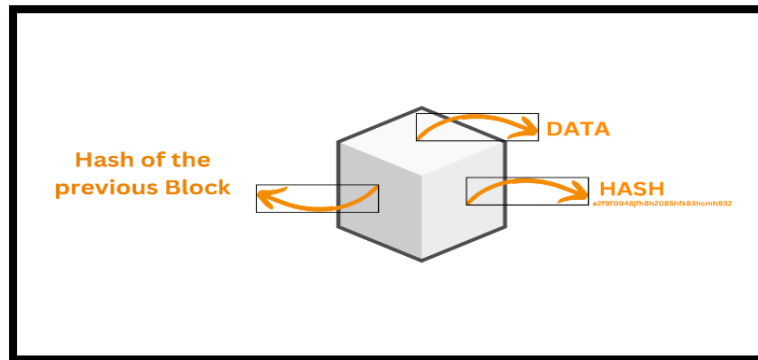


FIGURE 2 SHOWING A BLOCK CONTAINING DATA, HASH, AND HASH OF PREVIOUS BLOCK.

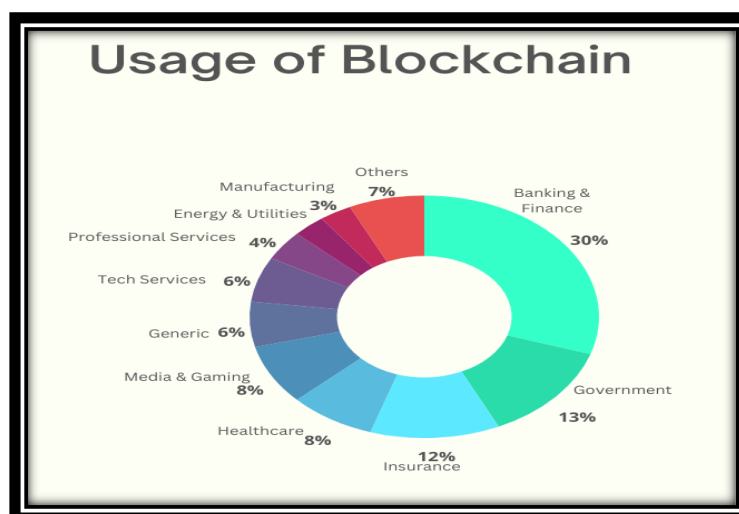


FIGURE.3 SHOWING USAGE OF BLOCKCHAIN IN DIFFERENT AREAS.

3. FOUNDATIONAL COMPONENTS OF BLOCKCHAIN

- **Asymmetric key cryptography:** Within the blockchain network, the utilization of asymmetric key cryptography plays a pivotal role in safeguarding blockchain operations. To execute any transaction, users necessitate a digital wallet secured by their private key, granting access through the generation of suitable signatures using this private key. The wallet's public key serves as the bitcoin address, which remains publicly known and is updated with each transaction to uphold user privacy and anonymity. It is the private keys that take responsibility for digitally signing transactions, ensuring their authenticity and security.
- **Transactions:** The exchange of data directly between nodes occurs through the use of files that contain transaction information. These files are initiated by a source node and disseminated across the entire network for verification. Within this network, every node possesses knowledge regarding the real-time status at each address and retains a replica of the current blockchain. This blockchain serves as an extensive record of all past transactions.
- **Consensus Mechanism:** Given the absence of a central authority to oversee transactions, mediate disputes, or shield against security breaches, a consensus must be reached among all nodes regarding a unified ledger updating protocol that guards against double-counting. In the case of Bitcoin, block miners engage in competitive efforts to discover the subsequent valid block by computing a cryptographic block hash. Nodes that successfully uncover the solution are incentivized with bitcoins, thereby facilitating the creation of new currency. This hash value is commonly referred to as 'proof of work,' and if all transactions and proof-of-work checks out as valid, nodes reach a consensus to accept it by updating their respective copies of the ledger.

IMPLEMENTATION

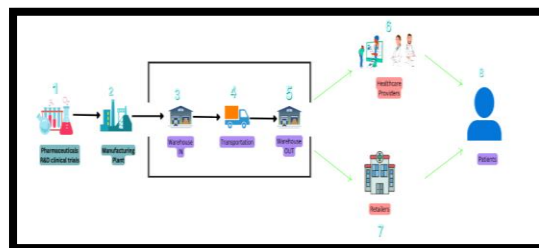


FIGURE 4 SHOWS THE INTERACTION OF NETWORK PARTICIPANTS

4. DESIGN AND STRUCTURE OF DSCMR: A BLOCKCHAIN-POWERED PHARMACEUTICAL SUPPLY CHAIN MANAGEMENT AND RECOMMENDATION SYSTEM

4.1 COMPREHENSIVE FRAMEWORK OF THE PROPOSED DSCMR SYSTEM

The main goal of the blockchain network is to disseminate data in a distributed fashion, with many transactions allowed for each block. For security reasons, these transactions are hashed and encrypted for safekeeping. The proposed system is a service-oriented architecture that provides users with smart contract and distributed ledger functions as services. Figure 5 provides a graphic representation of the central system design. The suggested system consists of two separate modules, one devoted to the administration of the medication supply chain and the other to the recommendation system. End users with the capacity to perform transactions inside this system include manufacturers, suppliers, distributors, pharmacies, physicians, and patients. A front-end web application makes this possible. It lets users order medications, provide raw materials, update medication information, manage orders, keep track of records, supervise drug delivery, exchange data, track the path of drugs through the supply chain, manage drugs, and supervise customer relations, among other things.

The foremost objective of this system is twofold. Firstly, it aims to prevent the circulation of counterfeit drugs, ensuring the authenticity and safety of pharmaceuticals. Secondly, it provides a secure supply chain management system that caters to the needs of end-users, thereby enhancing the overall efficiency and reliability of the pharmaceutical supply chain. Our proposed system prioritizes security, leveraging the robust features of blockchain technology to ensure data integrity and protection. Users within the system have the capability to trace the delivery status of pharmaceuticals through the DSCMR system. Moreover, the proposed system offers comprehensive CRUD (Create, Read, Update, Delete) operations among interconnected peer nodes.

We have implemented the notion of channels to improve security between nodes, therefore separating the whole network into discrete private networks. By shielding sensitive information from unrelated nodes, nodes in the same channel can directly communicate data with certain nodes while preserving privacy.

This channel-based approach empowers each participant to establish their individual private networks, thereby bolstering security.

In our system, patients are limited in their direct transaction capabilities, primarily interacting with the system to verify the authenticity of the drugs they purchase from a pharmacy. This verification is accomplished by simply scanning the barcode affixed to the drug packaging, offering an added layer of security and trust for end-users. Patients have access to essential information concerning the drug, such as the manufacturer company, manufacturing date, expiry date, price, and more. This access is made possible through the utilization of blockchain channels. Additionally, we can impose limitations on suppliers, confining their transactions solely to raw material-related activities with the manufacturer. These stakeholders operate within separate channels, effectively using the system as a private network.

The second module of our proposed system introduces a machine learning-based recommendation system. This component is designed to provide pharmaceutical company customers with recommendations for the most suitable medicines. To achieve this, we have integrated machine learning algorithms, including Ngram, LightGBM, and sentiment analysis. These algorithms have been trained on customer-generated comments from websites, encompassing feedback on the quality of medicines, whether positive or negative. Leveraging this training data, our model can make informed recommendations to a diverse array of customers, including pharmacies, doctors, clinics, hospitals, and patients.

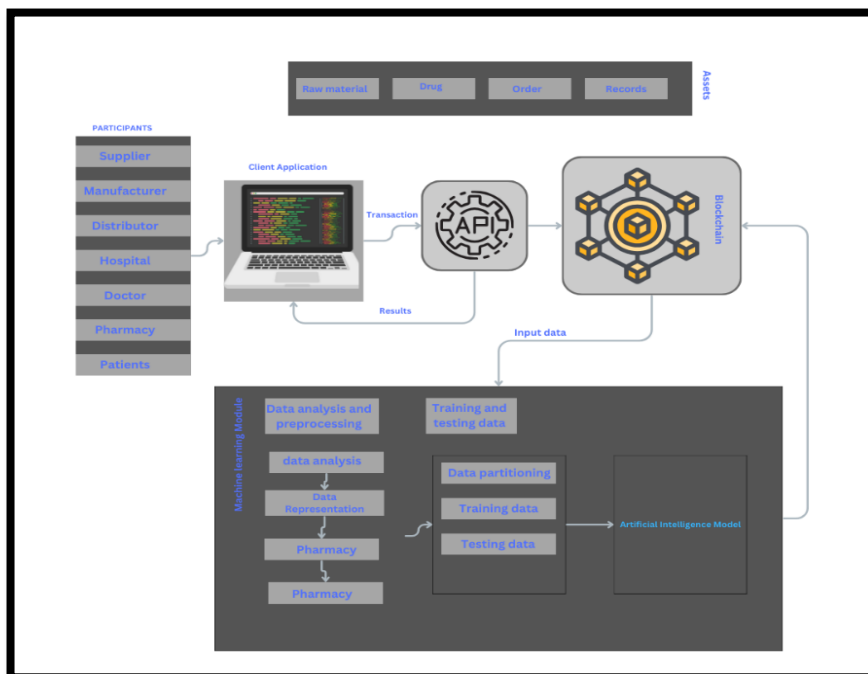


FIGURE 5 DETAILED ARCHITECTURE OF DSCMR SYSTEM.

4.2 TRANSACTIONS EXECUTION PROCEDURE IN DSCMR

This section explains the procedures needed to complete a transaction on the Blockchain network and offers insight into the transactional process inside the pharma supply chain management system. The procedure starts when users are given access to a client application's front end. This allows them to connect to the Blockchain system using their registered credentials and complete transaction requests.

The onus of registering all participants within the blockchain network lies squarely with the administrator. Only once participants have been duly registered by the administrator can they proceed with transaction execution. To initiate a transaction within the blockchain network, users log in to the client application and submit their transaction requests using their registered credentials. Subsequently, the transaction proposal is disseminated to all peer nodes. These peer nodes are categorized into two distinct groups: committers and endorsers. Endorsers bear the responsibility of executing and endorsing transaction proposals. They approve a proposal if it meets the criteria outlined in the smart contract, or they may deny it if it fails to fulfil those criteria. In contrast, committer peers are tasked with validating transaction results before these results are written into the transaction block of the ledger.

To comprehend the distinction between these two types of peers, it can be stated that endorser peers are essentially a specialized category of committer peers with the added role of holding predefined smart contracts. These endorser peers execute requested transactions' smart contracts within their own simulated environments prior to updating the ledger upon receiving the transaction proposals. During this process, endorser peers extract read and written data, collectively known as the RW set. The written data in the RW set relates to changes made to the world state following the execution of the transaction in the simulated environment, whereas the read data in the RW set contains information about the world state before to the transaction. The signed transaction and the RW sets are then returned to the client application by the endorser peers. The signed transaction containing all RW sets is then resubmitted by the client to the consensus management. The committer nodes get the transaction from the consensus manager, who then groups the data into a block.

Following this, the committer nodes engage in transaction validation by comparing the current world state with the transaction data. If the validation is successful, the transaction data is inscribed into the ledger. Consequently, the ledger undergoes updates in accordance with the written data. In the final phase, committer peers

transmit a notification to the client, indicating the transaction's status, which could be either "submitted" or "not submitted."

The communication between the client application and the blockchain network is established through the utilization of REST APIs and a software development kit (SDK).

4.2.1 MACHINE LEARNING-BASED DRUG PREDICTION AND RECOMMENDATION MODULE

In today's scenario, we see a surge in software applications dedicated to healthcare, generating vast amounts of patient health records. Various apps emerge from these datasets, aiming to monitor and manage patient health. Health analysts and data scientists are keen on utilizing this data to build automated systems for the healthcare sector. Selecting and recommending medicines for such a system poses a challenge. To accomplish this, understanding the impact of medications based on patient conditions and symptoms becomes crucial. Despite the numerous reviews, both positive and negative, surrounding each medicine, researchers are grappling with the task of creating a system that can recommend the best medicine considering these diverse opinions.

4.2.2 DEEP LEARNING MODELS

In this section, we are going to explain the models used for the recommendation of drugs in our proposed system; natural language processing model (NLP) N-gram for predicting the probability of a word in a sequence, LightGBM for reducing the low gradient features and sentiment analysis dictionary adding for the purpose of emotional analysis.

4.2.3 N-GRAM

Language models, such as the N-gram model, are employed to calculate the probabilities of words within a word sequence. The N-gram model, a specific type of language model, focuses on determining the probability distribution of a word within a sequence. This probability-centric model undergoes training on a text collection referred to as a corpus. During training, the N-gram model estimates probabilities by tallying the occurrences of word sequences within the provided corpus.

In practical terms, if an N-1-word sequence is presented to an N-gram model, it predicts words with a high likelihood of appearing in that sequence. The term "N-gram" denotes a sequence of N words. For instance, a bigram constitutes a two-word sequence like "don't disturb," "my car," or "your notebook." Similarly, a trigram represents a three-word sequence, such as "please don't disturb" and "close the door."

For example, how the N-gram model predicts the probability of a word in a corpus? Consider we have two sequences "heavy flood" and "heavy rain"; for these kinds of sentences N-gram model will predict the probability of word heavy with rain is more than with flood in the training collection of text. However, the probability of second-word rain will be more and selected by the model.

4.2.4 LIGHT GRADIENT BOOSTING MACHINE (LIGHTGBM)

In the realm of machine learning, the Gradient Boosting Tree algorithm is widely recognized and appreciated. Some notably effective variations of this algorithm include Extreme Gradient Boosting (XGBOOST) and Parallel Gradient Boosting Regression Tree (pGBRT). While these algorithms leverage various engineering optimization techniques, their efficiency and scalability may encounter challenges, particularly in scenarios with high-dimensional features and extensive datasets.

One prominent issue contributing to these challenges is the exhaustive nature of the information gain calculation. Specifically, for each individual feature, the algorithm examines all data instances, resulting in a time-consuming process. This becomes particularly evident and problematic when dealing with datasets characterized by high feature dimensions and extensive data volume. To address this challenge, Microsoft has introduced an innovative solution featuring two novel techniques: Gradient-Based One-Side Sampling (GOSS) and Exclusive Feature Bundle (EFB).

In the GOSS model, the exclusion of small gradient data instances occurs because only instances with high gradients are deemed crucial for estimating information gain. This results in a significant reduction of the data portion used for model training. Consequently, instances with low gradients do not impact the accuracy of model estimation. In our context, only data instances with high gradients are employed for information gain estimation. Despite the reduction in data size, GOSS can still provide accurate estimations.

On the flip side, EFB is employed for feature reduction by grouping mutually exclusive features together. Through the use of a greedy algorithm, EFB effectively reduces features without compromising the accuracy of information gain. Microsoft has coined these implementations based on GOSS and EFB as "lightGBM." Notably,

lightGBM stands out as a faster and more accurate model compared to other Gradient Boosting Decision Tree (GBDT) models. Microsoft claims that it is 20 times faster than alternative GBDT models.

4.2.5 DATASET

We use one of the best publicly accessible drug datasets from the UCI Machine Learning Datasets Repository in our proposed system's drug recommendation module.

Patients' evaluations and ratings of different drugs that they have taken are included in this collection. In particular, we use the datasets from Druglib.com and Drug.com, two well-known pharmaceutical websites. Notably, both consumers and medical experts see Drug.com as one of the biggest and most popular sources of information about pharmaceuticals.

The dataset comprises reviews categorized by specific conditions such as pain, acne, anxiety, blood pressure, etc. Additionally, users have the option to assign star ratings, ranging from 0 to 10, indicating their satisfaction levels with these drugs.

5 IMPLEMENTATION OF BLOCKCHAIN-BASED DSCMR SYSTEM

For providing the users an attractive graphical user interface, we used front-end languages like HTML, CSS, JAVASCRIPT, JQUERY and a third-party framework twitter bootstrap for developing client web-based application. The open-source Hyperledger Fabric's blockchain technology is used for the deployment of this system which requires docker engine, python, node JS and VS code as prerequisites. The Linux operating system Ubuntu 18.04 LTS is used for the deployment of the blockchain network. The Hyperledger composer is utilized for building the business logics of the proposed network where we have defined our participants, assets, access control rules and smart contracts. The client application is connected with the blockchain business network through the composer REST server in which participants can visualize and performs all the activities at their own web-based portals. The user can generate HTTP requests using GET, POST, PUT, DELETE methods through and response will be provided through the Blockchain system accordingly.

5.1 SMART CONTRACT DEVELOPMENT FOR DSCMR SYSTEM:

When building a blockchain network with Hyperledger Fabric, four essential components need definition:

- **Model File:** Used to define participants.

- **Script File:** Defines smart contract functions.
- **Access Control Rules File:** Specifies rules.
- **Query File:** Performs actions such as select and update.

One of the script file's functions that is especially meant for a transaction is shown by the excerpt that has been supplied. The updating of medication data, such as the production date, price, and expiration date, is handled by this transaction function. Creating, removing, and modifying assets and users in our blockchain-powered Drug Supply Chain Management and Recommendation (DSCMR) system are all handled by the script file.

JavaScript is employed for scripting these transaction functions during the modelling of the smart contract. The diagram illustrates two transaction functions: `sharerecordwithpharmacy` and `sharerecordwithdistributors`.

In the access control rules file, rules are crafted to govern user operations, delineating the actions permitted and restricted for all system users. This is achieved through the utilization of the rules access control language, where permissions for reading, writing, and deleting data are explicitly assigned to users. This approach serves as an effective means to grant users limited access within a blockchain network-based system.

```

1 namespace composers.participant
2 abstract participant Person {
3   o String firstname
4   o String lastname
5   o String username
6   o String email
7   o String phoneno
8 }
9 participant manufacturer identified by manufacturerID extends Person{
10  o String manufacturerID
11  o String company
12  o String accountno
13  o String Address
14 }
15
16 participant supplier identified by supplierID extends Person{
17  o String supplierID
18  o String company
19  o String accountno
20  o String Address
21 }
22
23 participant hospital identified by hospitalID extends Person{
24  o String hospitalID
25  o String Address
26 }

```

FIGURE 5. PARTICIPANTS DEFINITION IN HYPERLEDGER COMPOSER.

In order to get or update data within the blockchain network's current global state, queries are essential. These inquiries in the Hyperledger Composer-based business network are expressed in a separate file with a specific query language.

These searches are made up of two primary parts: the statement and the description. They resemble SQL database statements and conditions. The statement section uses SQL-like terms to create rules in the query, such as SELECT, ORDER BY, WHERE, OR, AND, etc.

The second part entails the description of that query, encompassing a function string for deploying specific rules.

```

1  function updateDrugDetail(uint256 drugId, string memory newDetail) public {
2      require(drugId > 0, "Drug ID must be greater than 0");
3      require(newDetail.length() > 0, "New detail cannot be empty");
4      Drug memory drug = _drugs[drugId];
5      require(drug != Drug(0), "Drug not found");
6      drug.detail = newDetail;
7      _drugs[drugId] = drug;
8  }
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10 // Update drug detail
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```

FIGURE 6. SMART CONTRACT DEFINITION IN SCRIPTS FILE FOR A FUNCTION UPDATE DRUG DETAIL.

6. FUTURE SCOPE

The proposed Drug Supply Chain Management and Recommendation (DSCMR) system, integrating Blockchain technology and machine learning, lays the groundwork for a transformative approach to pharmaceutical supply chain management. As the research advances, several avenues for future exploration and enhancement emerge, providing a rich landscape for ongoing research and development:

- BLOCKCHAIN INTEGRATION AND INTEROPERABILITY:**
 Future research could focus on further enhancing the integration of Blockchain technology into pharmaceutical supply chains. Exploring interoperability with other blockchain networks or technologies could facilitate seamless collaboration and data exchange between different stakeholders in the broader healthcare ecosystem.
- SCALABILITY AND PERFORMANCE OPTIMIZATION:**

As the DSCMR system evolves, researchers can explore strategies to optimize its scalability and performance. Investigating techniques such as sharding or advanced consensus mechanisms could enhance the system's ability to handle a growing volume of transactions without compromising efficiency.

- **PRIVACY AND REGULATORY COMPLIANCE:**

Addressing privacy concerns and ensuring compliance with regulatory frameworks, such as GDPR in Europe or HIPAA in the United States, is crucial. Future research can delve into developing privacy-preserving techniques within the blockchain network and ensuring that the DSCMR system adheres to global healthcare data protection standards.

- **MACHINE LEARNING REFINEMENT:**

Continuous refinement of the machine learning algorithms used for drug recommendations is essential. Incorporating more sophisticated natural language processing (NLP) techniques, deep learning models, and expanding the dataset to include diverse patient demographics could improve the accuracy and personalization of drug recommendations.

- **REAL-TIME MONITORING AND IOT INTEGRATION:**

Integrating Internet of Things (IoT) devices for real-time monitoring of drug shipments and storage conditions could further enhance the DSCMR system. This could include sensors to monitor temperature, humidity, and other factors, ensuring the quality and safety of pharmaceuticals throughout the supply chain.

- **GLOBAL ADOPTION AND COLLABORATION:**

Research efforts can be directed towards promoting the global adoption of blockchain-based pharmaceutical supply chain systems. Collaboration with international organizations, regulatory bodies, and pharmaceutical companies can help establish standards and frameworks for the widespread implementation of secure and transparent supply chains.

- **USER EDUCATION AND TRAINING:**

The success of the DSCMR system relies on the active participation of various stakeholders. Future research can explore strategies for educating and training users, including healthcare professionals, manufacturers, and patients, to ensure effective utilization of the system and its features.

- **CYBERSECURITY AND THREAT ANALYSIS:**

Continuous assessment of cybersecurity threats and proactive measures to mitigate potential risks is essential. Future research can focus on threat modelling, penetration testing, and the development of advanced security protocols to safeguard the DSCMR system against evolving cyber threats.

- **BLOCKCHAIN GOVERNANCE MODELS:**

Exploring different governance models for the blockchain network can contribute to the long-term sustainability and decentralization of the DSCMR system. Researching approaches for community-driven governance and consensus on system updates ensures a robust and democratic framework.

- **REGULATORY ADVOCACY:**

Actively engaging with policymakers and regulatory authorities to advocate for the adoption of blockchain-based systems in pharmaceutical supply chains is crucial. Future research can contribute to the development of regulatory guidelines and frameworks that promote innovation while addressing legal and ethical considerations.

In summary, the future scope of the research involves a holistic approach towards refining and expanding the proposed DSCMR system. Continuous innovation, collaboration, and adaptation to emerging technologies and regulatory landscapes will be pivotal in ensuring the effectiveness and sustainability of blockchain-based pharmaceutical supply chain management.

7. CONCLUSION

In conclusion, the widespread issue of counterfeit drugs poses a significant threat to global public health, necessitating a robust solution for pharmaceutical supply chain management. So it advocates for the adoption of Blockchain technology to enhance transparency, traceability, and security within the supply chain. The proposed Drug Supply Chain Management and Recommendation (DSCMR) system leverages Blockchain's decentralized nature, ensuring real-time visibility and preventing the circulation of counterfeit drugs. The integration of machine learning augments the system by providing personalized drug recommendations based on patient reviews. Using Hyperledger Fabrics and Composer, the user-friendly system offers a secure interface for participants. Overall, this chapter presents a concise framework combining Blockchain and machine learning to address pharmaceutical supply chain challenges, promoting a secure and efficient drug distribution network for the evolving pharmaceutical industry.

In an era where counterfeit drugs jeopardize public health, a robust pharmaceutical supply chain is imperative. This study advocates for Blockchain as a transformative solution, introducing transparency, traceability, and security. The Drug Supply Chain Management and Recommendation (DSCMR) system, built on Hyperledger Fabrics and Composer, utilizes Blockchain's decentralized architecture to prevent counterfeit drug circulation and provide real-time visibility. Incorporating machine learning

enhances the system, offering personalized drug recommendations based on patient reviews. The user-friendly interface accommodates various stakeholders, ensuring accessibility and participation. This comprehensive framework aligns with the pharmaceutical industry's need for modern, secure, and intelligent supply chain management, laying the foundation for a safer and more efficient drug distribution network.

8. REFERENCES

- S. F. Roy and M. Jerremy, "African Counterfeit Pharmaceutical Epidemic: The Road Ahead," ACAPPP, 2009
- 2. "WHO | Growing Threat from Counterfeit Medicines," Bulletin of the World Health Organization, vol. 88, no.4, pp, 2010.
- Newton, P.N.; Green, M.D.; Fernández, F.M.; Day, N.P.; White, N.J. Counterfeit anti-infective drugs.
- Lancet Infect. Dis. 2006, 6, 602–613.
- Khatoon, A. A Blockchain-Based Smart Contract System for Healthcare Management. Electronics 2020, 9, 94
- Ijazul Haq , Olivier Muselemu Esuka , “ Blockchain Technology in Pharmaceutical Industry to Prevent Counterfeit Drugs”
- Puthal, D., Malik, N., Mohanty, S., Kougianos, E., & Das, G. (2018). Everything you wanted to know about the blockchain: its promise, components, processes, and problems. ResearchGate. doi: 10.1109/MCE.2018.2816299.
- Chen, M.; Suresh, A.T.; Mathews, R.; Wong, A.; Allauzen, C.; Beaufays, F.; Riley, M. Federated Learning of N-gram Language Models. arXiv 2019, arXiv:1910.03432
- Ke, G.; Meng, Q.; Finley, T.; Wang, T.; Chen, W.; Ma, W.; Ye, Q.; Liu, T.Y. Lightgbm: A highly efficient gradient boosting decision tree. In Advances in Neural Information Processing Systems 30; Curran Associates, Inc.: Dutchess County, NY, USA, 2017; pp. 3146–3154.
- Goldwater, S. ANLP Lecture 6 N-gram Models and Smoothing. 2019. Available at: ANLP Lecture 6 N-gram models and smoothing · Original version used backo, later \modi ed Kneser-Ney" introduced using interpolation (Chen and Goodman, 1998). Fairly complex equations, - [PDF Document] (vdocuments.mx)
- Parsing, C. Speech and Language Processing. 2009. Available at: Microsoft PowerPoint - slp12_f19.ppt [Compatibility Mode] (pitt.edu)

- Ke, G.; Meng, Q.; Finley, T.; Wang, T.; Chen, W.; Ma, W.; Ye, Q.; Liu, T.Y. Lightgbm: A highly efficient gradient boosting decision tree. In Advances in Neural Information Processing Systems 30; Curran Associates, Inc.: Dutchess County, NY, USA, 2017; pp. 3146–3154
- UCI. Drug Review Dataset. Available at: Drug Review Dataset (Drugs.com) - UCI Machine Learning Repository
- Dr.M.Karthika , Dr.T.Sujithra: An Overview of Hyperledger Fabric in Blockchain Framework over Virtual Networks.
- Hyperledger. Hyperledger Blockchain. Available online: Hyperledger - The Open Global Ecosystem for Enterprise Blockchain
- Cachin, C. Architecture of the hyperledger blockchain fabric. In Proceedings of the Workshop on Distributed Cryptocurrencies and Consensus Ledgers, Hangzhou, China, 11–13 May 2016; Volume 310, p. 4
- Khizar Abbas, Muhammad Afaq, Talha Ahmed Khan and Wang-CheolSong : A Blockchain and Machine Learning-Based Drug Supply Chain Management and Recommendation System for Smart Pharmaceutical Industry.