INCORPORATING DATA MINING METHODS FORENHANCING FRAUD DETECTION WITHIN FINANCIAL SECTORS

Dr.E. Balakrishna, Associate Professor CSE, Vaagdevi College of Engineering (Autonomous), India

V. Tharun, UG student, CSE, Vaagdevi College of Engineering (Autonomous), India

V. Sahithi, UG student, CSE, Vaagdevi College of Engineering (Autonomous), India

M. Abhilash, UG student, CSE, Vaagdevi College of Engineering (Autonomous), India

J. Uma maheshwara chary, UG student, CSE, Vaagdevi College of Engineering, India.

ABSTRACT

Fraud detection is a scenario applicable to many industries such as banking and financial sectors, insurance, healthcare, government agencies and law enforcement and more. There has been a drastic increase in recent years ,pushing fraud detection more important than ever. Hundreds of millions of dollars are lost to fraud every year.

Upcoding fraud is one such fraud in which a service provider acquires additional financial gain by coding a service by upgrading it even though the lesser service has been performed. Incorporating artificial intelligence with data mining and statistics help to anticipate and detect these frauds and minimize costs.

Using sophisticated data mining tools ,millions of transcations can be searched to spot patterns and detect fraudulent transactions. This paper gives an insight into the various datamining tools which are efficient in detecting upcoding frauds especially in the healthcare insurance sector in India.

INTRODUCTION

The term "fraud" implies an intention on the part of some party or an individual presumably planning to commit fraud. Different forms of frauds creates threats for the mankind be it social or financial sector. Traditional methods of data analysis was used to detect fraud and it was quite complex and timeconsuming. Fraud comprises of many instances involving repeated contraventions using the same method. Fraud cases can be similar in content and appearance but not identical

In bankingfraud may be using stolen credit cards, falsifying cheques, misleading accounts and more. In insurance, 20% to 25% of claims contain some form of fraud, leading to approximately 10% of insurance payout dollars. Fraud can vary from inflated losses to deliberately causing an accident for the payout. With all the different methods of fraud, detecting it becomes still harder.

Data analysis techniques to prevent fraud were first used by the telephone companies, the insurance companies and the banks. Based on a neural network shell ,Falcon fraud assessment system, FICOwas successfully implemented in the banking industry. Retail industries also suffer setbacks from frauds. Some supermarkets have started to make use of digitized closed-circuit television (CCTV) together with POS data of most susceptible transactions to fraud. Internettranscations is a big concern nowadays with some research pointing that these trans cation fraud is 12 times higher than in-store fraud. Fraud is an adaptive crime, so it requires special methods of intelligent data analysis to detect and prevent it. These methods exist in the areas of Knowledge Discovery in Databases (KDD), Data Mining, Machine Learning and Statistics. They offer applicable and successful solutions in different areas of fraud crimesData Mining is associated with (a) supervised learning based on training data of known fraud and legitimate cases and (b) unsupervised learning with data that are not labeled to be fraud or legitimate. Bedford's law can be interpreted as an example of unsupervised learning (Bolton et al. 2002). The direct application of these methods to forensic accounting is limited due to almost complete nonexistence of large sets of fraud training data (Bolton et al. 2002; Jensen, 1997). Insurance fraud, credit card fraud, telecommunications fraud, and check forgery are some of the main types of fraud. Insurance fraud is common in automobile, travel. The Uniform Suspected Insurance Fraud Reporting Form, adopted by the NAIC Antifraud Task Force 2003, replaced the prior Task Force form.

This form standardizes insurance fraud data for the insurance industry and makes it easier to report and track. Fraud detection involves three types of offenders (Baldock, 1997): i) Criminal offenders, ii) organized criminal offenders who are responsible for major fraud, and iii) offenders who commit fraud.

LITERATURE SURVEY

The paper presents application of data mining techniques to fraud analysis. We present some classification and prediction data mining techniques which we consider important to handle fraud detection. There exist a number of data mining algorithms and we present statistics- based algorithm, decision treebased algorithm and rule-based algorithm. We present Bayesian classification model to detect fraud in automobile insurance. Naïve Bayesian visualization is selected to analyze and interpret the classifier predictions. We illustrate how ROC curves can be deployed for model assessment in order to provide a more intuitive analysis of the models.

Financial fraud is defined as unlawful or criminal duplicity attempted to result organizational or personal gain. This is a terrible threat to the economics of a firm, corporate sector, Government or ordinary customers. Several processes exist to detect different types of financial fraud. But, due to inefficiency of those processes, researchers leverage data mining techniques to detect financial fraud.

This paper aims to build a systematic academic review of the data mining techniques applied to detect financial fraud in the recent years. Although fraud is not a new menace to the economic society, but it is making a great deal of impact in the world of economy in the formof Bank fraud, Credit card fraud, Insurance fraud, Financial Statement fraud, Corporate fraud, Money laundering etc.

In the practice of data mining, six data mining classes (Classification, Clustering, Visualization, Prediction, Regression, Outlier detection) have been used as a core or base while different techniques (K-nearest neighbour, Decision tree, Fuzzy logic, logistic model, Bayesian belief network, Naïve Bayes, Beneish M-Score model, Benford's law, Altman Z-score) have been applied to improve accuracy of fraud detection. In this paper, existing financial fraud detection techniques are compared with the advantage and limitations of the techniques.

This survey paper categorises, compares, and summarises from almost all published technical and review articles in automated fraud detection within the last 10 years. It defines the professional fraudster, formalises the main types and subtypes of known fraud, and presents the nature of data evidence collected within affected industries.

Within the business context of mining the data to achieve higher cost savings, this research presents methods and techniques together with their problems. Compared to all related reviews on fraud detection, this survey covers much more technical articles and is the only one, to the best of our knowledge, which proposes alternative data and solutions from related domains.

Fraud entails deception in order to obtain illegal gains; thus, it is mainly evidenced within financial institutions and is a matter of general interest. The problem is particularly complex, since perpetrators of fraud could belong to any position, from top managers to payroll employees. Fraud detection has traditionally been performed by auditors, who mainly employ manual techniques. These could take too long to process fraud-related evidence. Data mining, machine learning, and, as of recently, deep learning strategies are being used to automate this type of processing. Many related techniques have been developed to analyze, detect, and prevent fraud- related behavior, with the fraud triangle associated with the classic auditing model being one of the most important of these work aims to review current work related to fraud detection that uses the fraud triangle in addition to machine learning and deep learning techniques. We used the Kitchenham methodology toanalyze the research works related to fraud detection from the last decade. This review provides evidence that fraud is an area of active investigation. Several works related to fraud detection using machine learning techniques were identified without the evidence that they incorporated the fraud triangle as a method for more efficient analysis

PROBLEM STATEMENT

Frauddetectionisasituationthatcanoccurinavarietyofbusinesses, including the banking and financial sectors, insurance, healthcare, government organizations, andlaw enforcement. Recent years have seenasharpincrease,makingfrauddetectionmorecrucialthanever. Everyyear, fraudcauseshundredsofmill ionsofdollarstobelost. Upcodingfraudisonesuchfraudinwhichaserviceprovideracquiresadditionalfinanc ialgainbycodingaservicebyupgradingit even though the lesser service has been performed. Existing System uses Data Analysis methods todetectfrauddetectioninfinancialsectorsbut, the problemarises in the analysing of largedata of big corporate companies.

Frauds are becomingverymusteasytodoinanyofthesectorsmainlyinfinancialsectorsinsurance frauds or increasing everyday. Internet transcations is a big concern nowadayswith some research pointing that these trans cation fraud is 12 times higher than in-store fraud.Fraudulent insurance claims are one of the biggest preventable losses that hurt insurers worldwide.The P&C segment accounts for the most fraudulent insurance claims, with auto insurance andworkers' compensation making up the biggest percentage of fraudulent claims that annually impact their surance business.

3.1 LIMITATION OF SYSTEM

Upcoding is illegal and a fraudulent activity followed by medical providers by cheating the insurance providers and gain more income than they are eligble. This practice is seen to be costly for individual patients as well as for the insurance pay3rs. Fraud can distort markets by giving fraudsters a competitive advantage and driving out legitimate businesses. Fraud can cause immediate and long-term environmental damage by polluting the environment and destroying ecosystems and biodiversity. It can also lead to significant clean-up expenses fundamental coding error, such as incorrect CPT codes or upcoding and unbundling, might result in a regulatory probe and severe penalties, including imprisonment.

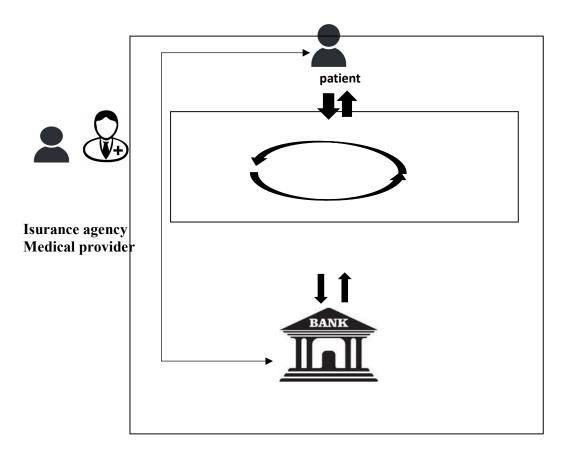
PROPOSED SYSTEM

Frauddetectionand datamining gatwaysgoeshandinhand. With the introduction of this technique and with the support of artificial intelligence and machine learning, frauddetection has become much simpler and easier and Medical claim fraud detection with the help of datamining plays an important role for an insurance company to detect fraudulent claims. Nearly 10% to 20% of the insurance amount is wasted on fraudulent claims.

ADVANTAGES

With the use of mining data techniques, the wastage can be reduced to a great extent, even though it is difficult to completely eradicate fraud claims. Supervised learning, Unsupervised and Hybrid learning methods in data mining pave way to efficient methods in detecting faults or anomalies and thus helps in mitigating frauds. Data mining techniques were used to detect fraud across different financial applications such as health insurance and credit card. Data mining with its wide variety of techniques is able to juice out a lot of useful information from a large set of data. With its ability to find useful knowledge from a given data, it is a potent technique to identify abnormal patterns in data and any underlying unwanted activity. Industries like Insurance, Banking, Credit Card, and Telecom are most vulnerable to financial fraud with large sets of data.

SYSTEM ARCHITECTURE



IMPLEMENTATION

PATIENT

Here patient is a module, patient should register to the application then only he can ableto login into the application. After successful login he can perform some operation such as managehis bank account like can send request to bank to take account and can viewhis status and candepositmoney into his account, Then the patient can view all policies, can purchase the policy also while purchasing thepolicy if patient enter more than 3 time wrong cvv then bank have the chance to block his account. After purchasing the policy patient can send the medical request to the medical provider andgenerate the treatmentbill andcan viewthemedical billsent by the medical provider and logout.

Patientcanperformthesefollowingfunctionsinpatientmodule:

- Register
- Login
- Manageaccount
- Allpolicies
- Runningpolicies
- Medicalrequest
- Viewmedicalbill
- Logout

5.2 MEDICALPROVIDER

Heremedicalproviderfirstastotakeoneaccountwithspecifiedspecializationthenloginintohis account, after successful login he can view medical request, can generate the medical billand send the insurance details request to patient and get the patient insurance details and applyinsurancebehalf ofpatientand get theclaim amount if everything isclear then at lastlogout.

Medicalprovidercanperformthesefollowingfunctionsinmedicalprovider module:

- Register
- Login
- Viewrequest
- **❖** Patientinsurancedetails
- Viewclaims
- **❖** logout

6.3 INSURANCEPROVIDER

Hereinsuranceprovideralsotakeoneaccountintothisapplicationaftersuccessfull loginhe can perform some operations like he can add policies and can view policies and can view allpolicyholder. Andtheinsuranceprovidercanverifytheinsuranceclaimrequest, if theinsuranceamountis less than medical bill using data mining filtering technique our system will identify and show totheinsuranceprovider, hereinsuranceprovider canblock those claim requestand logout Insuranceprovidercan perform these following functions in insurance provider module:

- Register
- **❖** Login
- Addpolicies
- Viewpolicies
- Policyholder
- Claimrequest
- **❖** logout

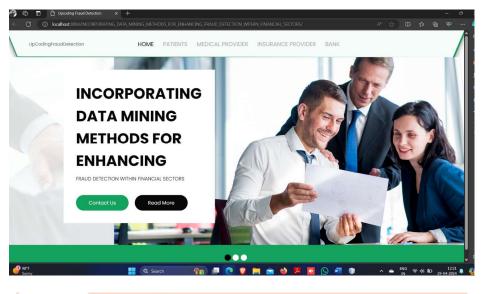
BANK

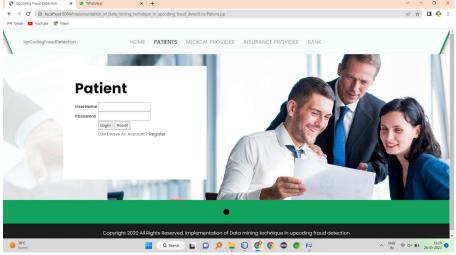
Here bank is a module, bank can directly login with specified details and after successfull loginhecanperformsomeoperationsuchasviewaccountrequest, viewfraudaccounts, viewgraphand logoutBank make the main role in detecting fraud ,any fraudulent activities are only detected bybankmodule.only bank can block oractivate anyaccount

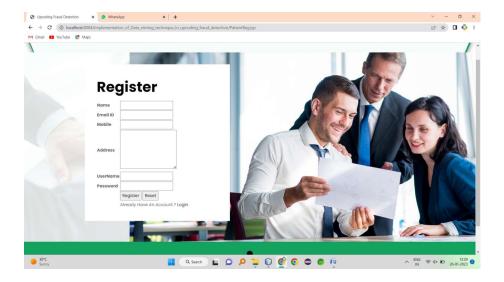
Bankcanperformthese followingfunctions in bank module:

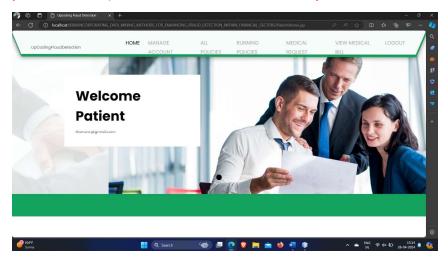
- **❖** Login
- **❖** Accountrequest
- Viewfraudaccount
- Viewgraph
- **❖** logout

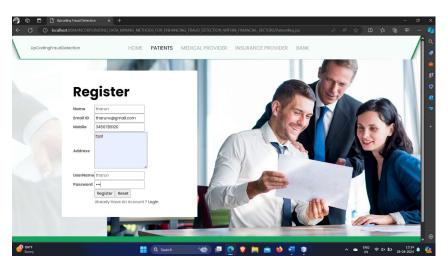
5. EXPECTED OUTCOMES

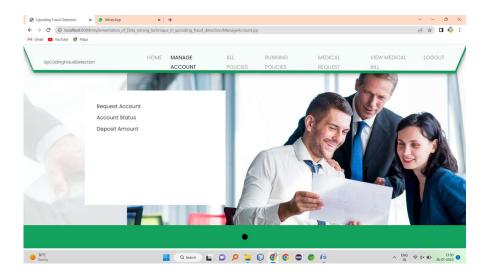




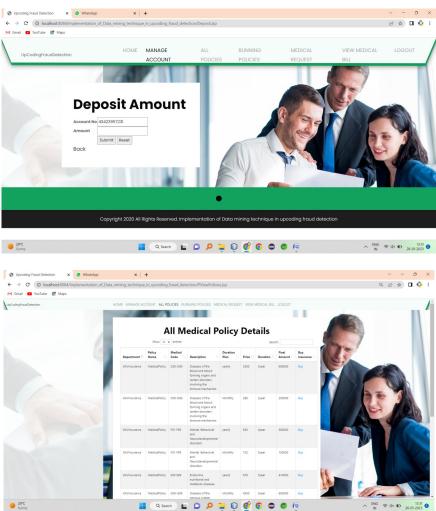


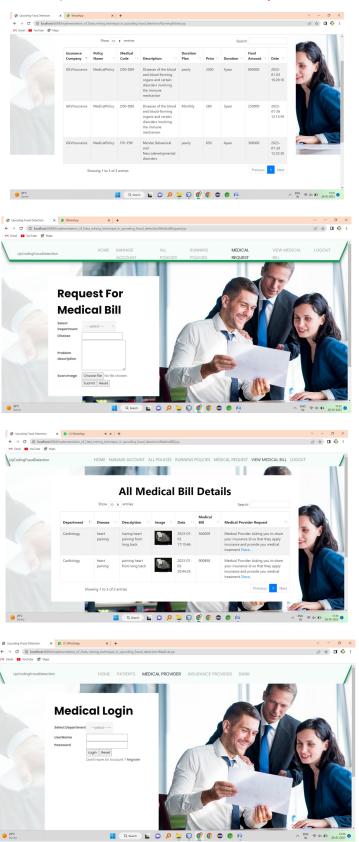


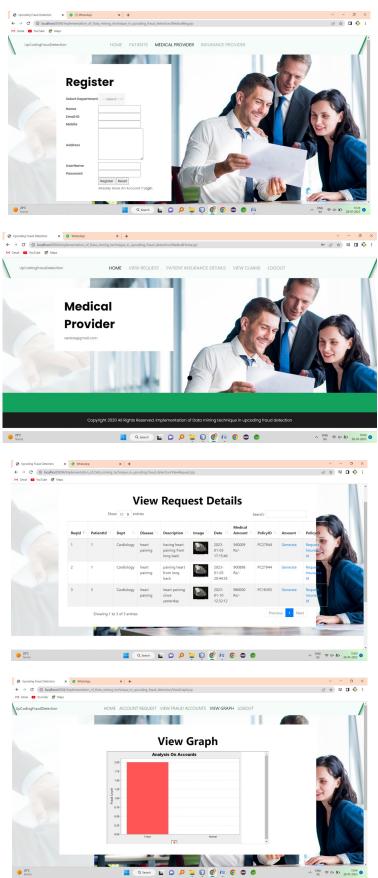












CONCLUSION

Fraud including upcoding puts a heavy financial burden not only to the insurer but also to the customer as this would increase the premium rates payable to the insurer. So efficient mechanisms has to be formulated and implemented to detect these financial frauds as well as to mitigate it.

Healthcare fraud detection studies are limited using supervised and unsupervised learning methodologies and is very few in the case of upcoding healthcare fraud detection. Linear regression ,mixed logit, Bayesian models are the supervised techniques used for upcoding fraud detection. A combination of subgroup creation via decision tree and Fisher's Exact Test are done using the unsupervised learning techniques. The application of additional learning and classification technique in this field will surely pave way for more innovative research options to reduce the risk of upcoding frauds...

FUTURE SCOPE

The future scope for using data mining methods to detect fraud in the financial sector is promising, as technological advancements, evolving fraud techniques, and increased data availability continue to drive innovation in this field. Here are some key areas of future development and opportunities for data mining in fraud detection within the financial sector:

Machine Learning and AI Integration: The integration of machine learning and artificial intelligence (AI) into data mining techniques will enhance fraud detection accuracy. Deep learning models, natural language processing (NLP), and anomaly detection algorithms will play a more significant role in identifying complex fraud patterns.

Big Data and Real-time Analysis: As financial data volumes continue to grow, data mining will need to adapt to handle big data. Real-time analysis of financial transactions and customer behavior will become increasingly important for immediate fraud detection and prevention.

REFERENCES

- [1]Sasongko, Nanang. "Fraud Detection In Non-Cash Transactions Using Information System Audit (A Case Study Of Government Bank in Bandung Indonesia)." ASEAN/Asian Academic Society International Conference Proceeding Series. 2015.
- [2]Palshikar, Girish Keshav. "The hidden truth-frauds and their control: A critical application for business intelligence." Intelligent Enterprise 5.9 (2002): 46-51. [3]https://en.wikipedia.org/wiki/Data analysis techniques for fraud detection
- [3]Bolton, Richard J., and David J. Hand. "Statistical fraud detection: A review." Statistical science (2002): 235-249.
- [4]Bauder, Richard, Taghi M. Khoshgoftaar, and Naeem Seliya. "A survey on the state of healthcare upcoding fraud analysis and detection." Health Services and Outcomes Research Methodology: 1-25.
- [5]https://en.wikipedia.org/wiki/ICD-10
- [6]http://www.who.int/classifications/icd/en/
- [7]Najafabadi, Maryam M., et al. "Deep learning applications and challenges in big data analytics." Journal of Big Data 2.1 (2015): 1.
- [8]Phua, Clifton, et al. "A comprehensive survey of data miningbased fraud detection research." arXiv preprint arXiv:1009.6119 (2010).
- [9]Kirkos, Efstathios, Charalambos Spathis, and Yannis Manolopoulos. "Data mining techniques for the detection of fraudulent financial statements." Expert systems with applications 32.4 (2007): 995-1003.