



# Financial Markets Fraud Detection and Prevention using Kernel Adatron Algorithm with Machine Learning

Dr. Om Prakash Yadav<sup>1</sup>, Mr. Ravi Kant Sahu<sup>2</sup>, Mr. Anand Soni<sup>3</sup>

om.26121@lpu.co.in<sup>1</sup>, ravi.16920@lpu.co.in<sup>2</sup>, anand.soni@bti.moe.bh<sup>3</sup>

Assistant Professor, School of Computer Science and Engineering<sup>1,2</sup>

Lovely professional University, Phagwara, Punjab, India

Sr. Lecturer, Commercial Studies Division Bahrain Training Institute<sup>3</sup>

Ministry of Education, Kingdom of Bahrain

**Abstract:** *The investors may have great interest to invest in stock market. Moreover, financial markets like stock markets are driven by explosive factors such as social media, micro blogs and news that make it hard to predict stock market index based on purely the historical data. The financial services industries that involve financial transactions are suffering from fraud-related losses and damages. Machine Learning (ML) is being rapidly adopted for a range of applications. It is important to begin considering the financial stability implications for every financial asset's organization. Using the machine learning tools and techniques in the finance sector will become necessary because it will closely monitor nascent and rapidly evolving landscape, wherein data on usage are largely unavailable, and bereft of any analysis. Financial assets fraud has seriously affected investors' confidence in the stock market and economic stability. The huge economic losses incurred because of several serious financial fraud events and because of this the intelligent financial fraud detection has thus been the topic of recent advances. In recent years, several studies have used stock market and machine learning techniques to provide solutions to this problem. In this paper, we propose various a state of art fraud detection techniques such as classification, clustering, and regression. This study aims to identify the techniques and methods that give the best results that have been perfected so far. Stock markets can benefit if fraud identification and prevention can be incorporated by using machine learning algorithms.*

**Keywords:** Financial Market, Fraud Detection & Prevention, Machine Learning, Kernel Adatron.

## REFERENCES

- [1]. Abbasi, Ahmed, Conan Albrecht, Anthony Vance, and James Hansen. 2012. "Metafraud: A Meta-Learning Framework for Detecting Financial Fraud." MIS Quarterly 36, no.4: 1293- 1327.
- [2]. Abbasi, Albrecht, Vance and Hansen (2012). Metafraud: A meta-learning framework for detecting financial fraud. MIS Quarterly, 36(4), 1293-1327.
- [3]. <https://www.fraud-magazine.com/article.aspx?id=%204295016799>.
- [4]. Analysis, F., Wireshark, U., In, T., & Internet, I. (2019). International Journal of Arts and Science Research frame analysis using wireshark and TOPAS in industrial internet of thinks. 6(1), 4-11.
- [5]. Anlauf JK, Biehl M. The Adatron: An adaptive perceptron algorithm. Europhysics Letters. 1989;10:687.
- [6]. Beasley, Mark S., Joseph V. Carcello, Dana R. Hermanson, and Terry L. Neal. 2010. "Fraudulent Financial Reporting: 2998-3007; An Analysis of U. S. Public Companies Research." New York: Committee of Sponsoring Organizations of the Treadway Commission (COSO). <https://www.coso.org/Documents/COSO-Fraud-Study-2010-001.pdf>
- [7]. Boston, W. (2015). Volkswagen Emissions Investigation Zeroes on Two Engineers. the Wall Street Journal 5.10.2015. Available: <https://www.wsj.com/articles/vw-emissions-probe-zeroes-in-on-two>

- engineers-1444011602.Retrieved:3.3.2019.
- [8]. Campbell, C. (n.d.). The Kernel-Adatron Algorithm : a Fast and Simple Learning Procedure for Support Vector Machines. Systems Engineering.
  - [9]. Coalition against Insurance Fraud, "Learn about fraud", [http://www.insurancefraud.org/learn\\_about\\_fraud.htm](http://www.insurancefraud.org/learn_about_fraud.htm).
  - [10]. Cortes C, Vapnik V. Support vector networks. Machine Learning. 1995;20:273–97.
  - [11]. CULS, Cornell University Law School, White-Collar Crime: an overview, [http://topics.law.cornell.edu/wex/White-collar\\_crime\(2009\)](http://topics.law.cornell.edu/wex/White-collar_crime(2009)).
  - [12]. Data Science (2017). Supervised vs. Unsupervised Machine Learning. Available: <https://www.datascience.com/blog/supervised-and-unsupervised-machine-learning-algorithms>. Retrieved: 08.03.2019.
  - [13]. De Castro, P. A. L., & Teodoro, A. R. B. (2019). A Method to identify anomalies in stock market trading based on Probabilistic Machine Learning. Journal of Autonomous Intelligence, 2(2), 42. <https://doi.org/10.32629/jai.v2i2.44>.
  - [14]. Dong, Wei, Stephen Shaoyi Liao, Bing Fang, Xian Cheng, Zhu Chen, and Wenjie Fan. 2014. "The Detection of Fraudulent Financial Statements: An Integrated Language Model." Proceedings of 19th Pacific Asia Conference on Information Systems, June 24-28, 2014, Chengdu, China. Atlanta, GA: Association for Information Systems AIS eLibrary, 383. <https://aisel.aisnet.org/pacis2014/383>.
  - [15]. Donning, H., Eriksson, M., Martikainen, M., & Lehner, O. M. (2019). Prevention and Detection for Risk and Fraud in the Digital Age-the Current Situation. ACRN Oxford Journal of Finance and Risk Perspectives, 8, 86–97.
  - [16]. FBI, Federal Bureau of Investigation, Financial Crimes Report to the Public Fiscal Year, Department of Justice, United States, [http://www.fbi.gov/publications/financial/fcs\\_report2007/financial\\_crime\\_2007.htm\(2007\)](http://www.fbi.gov/publications/financial/fcs_report2007/financial_crime_2007.htm(2007)).
  - [17]. Frieß TT, Cristianini N, Campbell C. The Kernel-Adatron algorithm: A fast and simple learning procedure for support vector machines; Proceedings of the Fifteenth International Conference on Machine Learning; Morgan Kaufmann; 1998.[33]
  - [18]. Frieß TT, Harrison R. The Kernel-Adatron with bias unit: Analysis of the algorithm. 1998.
  - [19]. Golmohammadi, K., & Zaiane, O. R. (2012). Data mining applications for fraud detection in securities market. Proceedings - 2012 European Intelligence and Security Informatics Conference, EISIC 2012, 107–114. <https://doi.org/10.1109/EISIC.2012.5>
  - [20]. Hajek, Petr and Roberto Henriques. 2017. "Mining Corporate Annual Reports for Intelligent Detection of Financial Statement Fraud: A Comparative Study of Machine Learning Methods." Knowledge-Based Systems 128: 139-52.
  - [21]. Haykin S. Neural networks: A comprehensive foundation. 3ed. Prentice-Hall; 2007.
  - [22]. <https://m.rbi.org.in/scripts/AnnualReportPublications.aspx?Id=1348>
  - [23]. International, A. (2017). A Design and Development as A system for Machine-to- Machine ( M2M ) working under vision motivated by the Internet of Things ( IOT ). XIII.
  - [24]. J.L. Kaminski, Insurance Fraud, OLR Research Report, <http://www.cga.ct.gov/2005/rpt/2005-R-0025.htm>. 2004.
  - [25]. Menon AK. Large-scale support vector machines: Algorithms and theory. Research Exam, University of California; San Diego: 2009. pp. 1–17.
  - [26]. Murphy, P, R. and Free, C. (2016). Broadening the Fraud Triangle: Instrumental Climate and Fraud. Behavioural research in accounting. 28(1), 41-56.
  - [27]. Ngai E, Hu Y, Wong Y, Chen Y, and Sun X The application of data mining techniques in financial fraud detection: A classification framework and an academic review of literature Decision Support Systems Volume 50, Issue 3, p559-569 (2011).
  - [28]. Online Available: <http://www.acfe.com/uploadedfiles/acfewebsite/content/documents/rtnn-2010.pdf>.
  - [29]. Opper M. Learning times of neural networks: exact solution for a perceptron algorithm. Physical Review A. 1988;38:3824.



- [30]. PWC (2018). Pulling fraud out of the Shadows. Available: <https://www.pwc.com/gx/en/forensics/global-economic-crime-and-fraud-survey-2018.pdf>. Retrieved: 20.02.2019.
- [31]. Rezaee, Zabihollah. 2005. "Causes, Consequences, and Deterrence of Financial Statement Fraud." *Critical Perspectives on Accounting* 16: 277-98.
- [32]. S. Yin, J. Yin, Tuning kernel parameters for SVM based on expected square distance ratio, *Inf. Sci. (Ny)*. 370-371 (2016) 92-102.
- [33]. S.Bhattacharyya, S.Jha, K.Tharakunnel, J.C. Westland, Data mining for credit card fraud: A comparative study, Elsevier, *Decision Support Systems*, Volume 50, Issue 3, p602-613(2011).
- [34]. Sadgali, I., Sael, N., & Benabbou, F. (2019). Performance of machine learning techniques in the detection of financial frauds. *Procedia Computer Science*, 148(Icids 2018), 45-54. <https://doi.org/10.1016/j.procs.2019.01.007>.
- [35]. Shah, D., Isah, H., & Zulkernine, F. (2019). Stock market analysis: A review and taxonomy of prediction techniques. *International Journal of Financial Studies*, 7(2). <https://doi.org/10.3390/ijfs7020026>.
- [36]. Shi Y, Eberhart RC. Empirical study of particle swarm optimization; *Proceedings of the 1999 Congress on Evolutionary Computation*; pp. 1950-99.
- [37]. Tang, X. B., Liu, G. C., Yang, J., & Wei, W. (2018). Knowledge-based financial statement fraud detection system: Based on an ontology and a decision tree. *Knowledge Organization*, 45(3), 205-219. <https://doi.org/10.5771/0943-7444-2018-3-205>.
- [38]. V. Vapnik, *The nature of statistical learning theory*, 1995.
- [39]. W.H. Beaver, Financial ratios as predictors of failure, *Journal of Accounting Research* 4 p71- 111. (1966).
- [40]. West and Bhattacharya (2015). Intelligent financial fraud detection: A comprehensive review.
- [41]. *Computers & Security*, 57, 47-66.
- [42]. Y. Jin, R.M. Rejesus, B.B. Little, Binary choice models for rare events data: a crop insurance fraud application, *Applied Economics* Volume 37, Issue 7, p841-848. (2005).
- [43]. Y. Xu, L. Wang, P. Zhong, A rough margin-based v-twin support vector machine, *Neural Comput. Appl.* 21 (2011) 1-11.
- [44]. Yadav, O. P. (2018). Intelligent Grid to Autonomous Cars and Vehicular Clouds in Internet of Vehicles. 7(2), 7-13.
- [45]. Yadav, O. P. (2019). Study on the Performance and Efficiency of Multi-Organize Filtering Framework with A Large Arrangement of SIP Security Based VoIP Domain. 3085(05), 1101- 1106.
- [46]. Yadav, O. P., & Singh, R. P. (2018). INTERNET OF THINGS ( IOT ) SECURITY ISSUE IN WIRELESS SENSOR NETWORK ( WSN ) WITH RADIO FREQUENCY IDENTIFICATION ( RFID ). 1-7.
- [47]. Z.Gao, M.Ye, A framework for data mining-based anti-money laundering research, *Journal of Money Laundering Control* 10 (2), p170-179.

#### AUTHOR'S DETAILS



- Dr. Om Prakash Yadav, Assistant Professor in School of Computer Science and Engineering, Lovely Professor University. He worked ISL Engineering College, Sri Sarada Institute of Science and Technology and Grahmbell P.G. College, Hyderabad, Telangana. He has 16 years of teaching. He has published 16 International and 4 National Journals. He has participated and papers presented in in various National and International conferences. He had received Ph.D. degree from SSSTUMS, Bhopal, M.Tech from IETE, New Delhi and M.C.A from SMU.



- Mr. Ravi Kant Sahu, Assistant Professor, in School of Computer Science and Engineering, Lovely professor University. He has 10 years of teaching experience. He is awarded with M.Tech(CSE-Mobile Computing) from NIT Hamirpur (H.P.) [Gold Medalist] and B.Tech (I.T.) from Uttar Pradesh Technical University degrees. He is Oracle Certified Associate Java Programmer and qualified UGC NET and GATE exams. He has published 15+



**IJAR SCT**  
Impact Factor: 6.252

**IJAR SCT**

ISSN (Online) 2581-9429

International Journal of Advanced Research in Science, Communication and Technology (IJAR SCT)

Volume 2, Issue 1, November 2022

International Journals. and participated in 5 National/International Level/ IEEE conferences.