TELECOMMUNICATION SUBSCRIPTION FRAUD DETECTION USING NEURAL NETWORK

A thesis submitted to the Faculty of Information Technology in partial fulfillment of the requirement for the degree

Master of Science (Intelligent System)

Universiti Utara Malaysia

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ABSTRACT (Bahasa Malaysia)

Setiap tahun, penipuan telekomunikasi menelan kos berbilion dolar daripada keuntungan pelaburan. Jenayah penipuan telekomunikasi kebiasaannya dimotivasikan oleh wang atau sesuatu keadaan yg terdesak tanpa melibatkan nama, atau kedua-duanya. Melalui penggunaan internet, komuniti penyalahgunaan menjadi semakin berkolaborasi, dan akibatnya, mereka menjadi lebih bijak. Oleh itu, aplikasi pengesan penipuan perlu menjadi lebih sofistikated supaya selaras dengan penjenayah. Sebuah sistem untuk menghalang langganan penipuan dalam talian tetap dengan impak yang tinggi ke atas penghantar jarak jauh telah dicadangkan. Sistem tersebut meliputi teknik rangkaian neural, menggunakan kaedah' Multilayer Perceptron' (MLP). Sejumlah 158 sampel data dari Telekom Malaysia Bhd yang telah dikumpulkan, dilatih dan diuji dengan menggunakan model. Modul ramalan membenarkan pengenalpastian penipuan yang berpotensi ketika proses pendaftaran. Analisis data menunjukkan perkaitan yang kuat antara pemalar masukan yang terdiri daripada persamaan, faktor kepastian pengguna, penunjuk kes dan matlamat. Hasil kajian menunjukkan 78% ketepatan ramalan berjaya diperolehi. Daripada keputusan yang diperolehi, dapat disimpulkan bahawa rangkaian neural berpotensi untuk digunakan sebagai pengesan penipuan dalam telemounikasi.

ABSTRACT (English)

Telecommunications fraud costs carriers billions of revenue dollars annually. The telecom fraud criminal is often ingenious and is typically motivated by money, anonymity, or both. Through use of the Internet, the fraud community is becoming more collaborative, and as a result, more ingenious. Because of this, fraud detection applications must become more sophisticated to keep pace with the criminals. A system to prevent subscription fraud in fixed telecommunications with high impact on long-distance carriers is proposed. The system consists of a neural network technique, using Multilayer Perceptron (MLP). A total of 158 data samples from Telecom Malaysia Bhd were collected trained and tested using model. The prediction module allows identifying potential fraudulent customers at the time of subscription. The analysis of the data shows a reasonably strong correlation between the input variable, which consist of severity, user certainty factor, indicator case and target. The result shows that 78% of prediction accuracy has been obtained. From the result that has been produce, neural network has a potential to be used for detecting fraud in telecommunication.

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CHAPTER 1

INTRODUCTION

In this chapter, the first section explains the background of the study that mainly involves the fraud detection and fraud detection technique using Neural Network. Next, the research problem is presented, followed by the objectives, significance of the project and scope for the study. As a conclusion for this chapter, the organization of this study is presented.

1.1 Background

Telecommunication fraud is increasing dramatically each year resulting in loss of a large amount of euros worldwide. The history of telecommunications crime, including several types of fraudulent activities, was reviewed by Collins (1999a, 1999b, 2000). Fraud detection involves identifying fraud as quickly as possible once it has been committed.

The contents of the thesis is for internal user only

REFERENCES

- Barghava, B., Zhong, Y., & Lu, Y. (2003). Fraud formalization and detection. In Y. Kambayashi, M. Mohania, W. Wöb (Eds.), Data Warehousing and Knowledge Discovery, 5th Int. Conference, DaWak 2003, Prague, Czech Republic (pp. 330-339). LCNS Vol. 2737, Springer-Verlag.
- Bolton, R.J., & Hand, D.J. (2002). Statistical fraud detection: A review. *Statistical Science*, 17(3), 235-249.
- Burge, P., & Shawe-Taylor, J. (2001). An unsupervised neural network approach to profiling the behavior of mobile phone users for use in fraud detection. *Journal of Parallel and Distributed Computing*, 61, 915-925.
- Cahill, M. H., Lambert, D., Pinheiro, J.C., & Sun, D. X. (2002). Detecting fraud in the real world. In J. Abello, P. M. Pardalos, M. G. C. Resende (Eds.), *Handbook of Massive Data Sets*, (pp. 913-930). Dordrecht: Kluwer Academic Publishers.
- Chan, P. K., Fan, W., Prodromidis, A. L., & Stolfo, S. J. (1999). Distributed data mining in credit card fraud detection. *IEEE Intelligent Systems*, 14(6), 67-74.
- Collins, M. (1999a). Telecommunications Crime Part 1. *Computer & Security*, 18, 577-586.

- Collins, M. (1999b). Telecommunications Crime Part 2. *Computer & Security*, 18, 683-692.
- Collins, M. (2000). Telecommunications Crime Part 3. Computer & Security, 19, 141-148.
- Cortes, C., Pregibon, D., & Volinsky, C. (2001). Communities of interest. In F. Hoffman,
 D. J. Hand, N. Adams, G. Guimaraes (Eds.), Advanced in Intelligent Data Analysis,
 4th Int. Conference IDA 2001, Cascais, Portugal (pp. 105-114). LCNS Vol. 2189,
 Springer-Verlag.
- Cortes, C., Pregibon, D., & Volinsky, C. (2003). Computational methods for dynamics graphs. *Journal of Computational and Graphical Statistics*, 12(4), 950-970.
- Daskalaki, S., Kopanas, I., Goudara, M., & Avouris, N. (2003). Data mining for decision support on customer insolvency in telecommunications business. *European Journal of Operational Research*, 145, 239-255.
- Ezawa, K.J., & Norton, S.W. (1996). Constructing Bayesian networks to predict uncollectible telecommunications accounts. *IEEE Expert*, 11(5), 45-51.

- Fawcett, T., & Provost, F. (1997a). Combining data mining and machine learning for effective fraud detection. In AI Approaches to Fraud Detection & Risk Management, Workshop Technical Report WS-97-07 (pp. 14-19). AAAI Press, Menlo Park, California.
- Fawcett, T., & Provost, F. (1997b). Adaptive fraud detection. *Data Min. Knowl. Disc*, 1(3), 291-316.
- Fawcett, T. (2002a). Case study: Adaptive fraud detection. In W. Klösgen, & J. Zytkow (eds.), Handbook of Data Mining and Knowledge Discovery (section H1.2.1). Oxford U. Press, UK.
- Fawcett, T., & Provost, F. (2002b). Fraud detection. In W. Klösgen, & J. Zytkow (eds.), Handbook of Data Mining and Knowledge Discovery (section F2). Oxford U. Press, UK.
- Fawcett, T. (2003). ROC graphs: Notes and practical considerations for data mining researchers. HP Labs. Technical Report HPL-2003-4.
- Ghosh, S. Reilly, D.L. Credit card fraud detection with a neural-network. *Decision Support and Knowledge-Based Systems, Proceedings of the Twenty-Seventh Hawaii International Conference*. Volume 3: 621-630, 1994.

- Hansen, J. V., McDonald, J. B., Messier, W. F., & Bell, T. B. A generalized qualitative response model and the analysis of management fraud. *Management Science*, 42(7), 1022-1032, 1996.
- Hoath, P. (1999). What's new in telecoms fraud?. Computer Fraud & Security, 1999(2), 13-19.
- Hoath, P. (1998). Telecoms fraud, the gory details. *Computer Fraud & Security*, 1998(1), 10-14.
- Hong, S. J., & Weiss, S. M. (2001). Advances in predictive models for data mining.

 Pattern Recognition Letters, 22, 55-61.
- Liao, S. H. (2005). Expert systems methodologies and applications a decade review from 1995 to 2004. *Expert Systems with Applications*, 28, 93-103.
- Mozer, M. C., Wolniewicz, R., Grimes, D. B., Johnson, E., & Kaushansky, H. (2000)

 Predicting subscriber dissatisfaction and improving retention in the wireless telecommunications industry. *IEEE Trans. Neural Networks*, 11(3), 690-696.
- M. Taniguchi, M. Haft, J. Hollmen, and V. Tresp. Fraud detection in communication networks using neural and probabilistic methods. In *Pmceedings of the 1998 IEEE*

International Conference in Acoustics, Speech and Signal Pmcessing, volume 2, pages 1241-1244,1998.

Paredes, R.D. (2005). Lessons from the deregulation transition in Chile's local telephony market. *Telecommunications Policy*, 29, 333-350.

Patterson, D.W. (1995). Artificial neural networks: theory and applications. Singapore: Prentice Hall.

Phua, C., Lee, V., Smith, K., & Gayler R. (2005). A comprehensive survey of data mining-based fraud detection research. Available at http://www.bsys.monash.edu.au/ people/cphua/

Provost, F., & Fawcett, T. (1997). Analysis and visualization of classifier performance with nonuniform class and cost distributions. In AI Approaches to Fraud Detection & Risk Management, Workshop Technical Report WS-97-07 (pp. 57-63). AAAI Press, Menlo Park, California.

Rosset, S., Murad, U., Neumann, E., Idan, Y., & Pinkas, G. (1999). Discovery of fraud rules for telecommunications - challenges and solutions. In *Proceedings of the fifth ACM SIGKDD International Conference on Knowledge Discovery and Data Mining*, pp. 409-413.

- Shawe-Taylor, J., Howker, K., & Burge, P. (1999). Detection of fraud in mobile telecommunications. *Information Security Technical Report*, 4(1), 16-28.
- Shawe-Taylor, J., Howker, K., Gosset, P., Hyland, M., Verrelst, H., Moreau, Y., Stoermann, C., & Burge, P. (2000). Novel techniques for profiling and fraud in mobile telecommunications. In P. J. G. Lisboa, B. Edisbury, & A. Vellido (eds.), Business Applications of Neural Networks. The State-of-the-Art of Real World Applications (pp. 113-139). World Scientific, Singapore.
- Stehmann, O. (1995). Network liberalization and developing countries: The case of Chile. *Telecommunications Policy*, 19(9), 667-684.
- Stolfo, S. J., Fan, D. W., Lee W. L., Prodromidis, A. L. & Chan, P. K. (1997) Credit card fraud detection using meta-learning: issues and initial results. In AI Approaches to Fraud Detection & Risk Management, Workshop Technical Report WS-97-07 (pp. 83-90). AAAI Press, Menlo Park, California.
- Wong, B. K., Bodnovich, T. A., & Selvi, Y. (1997). Neural network applications in business: a review and analysis of the literature (1988-95). *Decision Support Systems*, 19, 301-320.
- Wong, B. K., & Monaco, J. A. (1995). Expert system applications in business: a review and analysis of the literature (1977-1993). *Information & Management*, 29, 141-152.

Y. Moreau, B. Preneel, P. Burge, J. Shawe-Taylor, C. Stoermann, and C. Cooke (1997).
Novel techniques for fraud detection in mobile telecommunication networks. In ACTS
Mobile Summit, Grenada, Spain.