

Announces the Ph.D. Dissertation Defense of

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for the degree of Doctor of Philosophy (Ph.D.)

"Addressing Highly Imbalanced Big Data Challenges for Medicare Fraud Detection"

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DEPARTMENT: Electrical Engineering and Computer Science

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ABSTRACT OF DISSERTATION

Addressing Highly Imbalanced Big Data Challenges for Medicare Fraud Detection

Access to affordable healthcare is a nationwide concern that impacts most of the United States population. Medicare is a federal government healthcare program that aims to provide affordable health insurance to the elderly population and individuals with select disabilities. Unfortunately, there is a significant amount of fraud, waste, and abuse within the Medicare system that inevitably raises premiums and costs taxpayers billions of dollars each year. Dedicated task forces investigate the most severe fraudulent cases, but with millions of healthcare providers and more than 60 million active Medicare beneficiaries, manual fraud detection efforts are not able to make widespread, meaningful impact. Through the proliferation of electronic health records and continuous breakthroughs in data mining and machine learning, there is a great opportunity to develop and leverage advanced machine learning systems for automating healthcare fraud detection.

This dissertation identifies key challenges associated with predictive modeling for large-scale Medicare fraud detection and presents innovative solutions to address these challenges in order to provide state-of-the-art results on multiple real-world Medicare fraud data sets. Our methodology for curating nine distinct Medicare fraud classification data sets is presented with comprehensive details describing data accumulation, data pre-processing, data aggregation techniques, data enrichment strategies, and improved fraud labeling. Data-level and algorithm-level methods for treating severe class imbalance, including a flexible output thresholding method and a cost-sensitive framework, are evaluated using deep neural network and ensemble learners. Novel encoding techniques and representation learning methods for high-dimensional categorical features are proposed to create expressive representations of provider attributes and billing procedure codes. Finally, meticulous performance evaluation and statistical analysis is used to highlight common pitfalls related to evaluating predictive models in the context of highly imbalanced big data and to make recommendations for future works in classifying highly imbalanced big data. As such, this dissertation covers the entire machine learning life cycle and provides meaningful contributions in areas related to data preparation, modeling techniques, and performance evaluation.

BIOGRAPHICAL SKETCH

Born in New Jersey, USA B.S., Florida Atlantic University, Boca Raton, Florida 2016 M.S., Florida Atlantic University, Boca Raton, Florida 2019 Ph.D., Florida Atlantic University, Boca Raton, Florida 2022

CONCERNING PERIOD OF PREPARATION & QUALIFYING EXAMINATION Time in Preparation: 2019 - 2022

Qualifying Examination Passed: Fall 2019

Published Papers:

J. M. Johnson and T. M. Khoshgoftaar. "Survey on deep learning with class imbalance". Journal of Big Data, 6(1):27, 2019.

J. M. Johnson and T. M. Khoshgoftaar. "The effects of data sampling with deep learning and highly imbalanced big data." Journal of Information Systems Frontiers, volume 22, pages 1113–1131 (2020).

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J. M. Johnson and T. M. Khoshgoftaar. "Thresholding strategies for deep learning with highly imbalanced big data." In Deep Learning Applications, volume 2. pages 199–227, 2021.

J. M. Johnson and T. M. Khoshgoftaar. "Robust thresholding strategies for highly imbalanced and noisy data." In 2021 20th IEEE International Conference on Machine Learning and Applications (ICMLA), pages 1182–1188. IEEE, 2021.

J. M. Johnson and T. M. Khoshgoftaar. "Medical provider embeddings for healthcare fraud detection." Springer Nature Computer Science Journal, 2(4): 276, 2021.

J. M. Johnson, Robert K. L. Kennedy, and T. M. Khoshgoftaar. "Learning from Highly Imbalanced Big Data with Label Noise." International Journal on Artificial Intelligence Tools (IJAIT). Accepted 2022.

J. M. Johnson and T. M. Khoshgoftaar, "Cost-Sensitive Ensemble Learning for Highly Imbalanced Classification." In 2022 21st IEEE International Conference on Machine Learning and Applications (ICMLA). IEEE, 2022.

J. M. Johnson and T. M. Khoshgoftaar. "Healthcare provider summary data for fraud classification." In 2022 IEEE 23rd International Conference on Information Reuse and Integration for Data Science (IRI), pages 236–242, 2022.

J. M. Johnson and T. M. Khoshgoftaar. "Encoding high-dimensional procedure codes for healthcare fraud detection." Springer Nature Computer Science Journal, 3(5):362, 2022.