

Second Prize Winner

The Impact of Artificial Intelligence on Mortality Modeling and Actuaries Niranjan Rajendran, B.Sc. (Hons)

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INTRODUCTION

Artificial intelligence (AI) is revolutionizing various fields, and actuarial science is no exception. The integration of AI into mortality modeling and forecasting presents unprecedented opportunities and risks that could reshape the role of actuaries and the methodologies they employ. This essay explores the use of AI in actuarial analysis, particularly focusing on the innovative use of nontraditional data sources. The aim is to elucidate how these advancements can enhance the accuracy and reliability of mortality forecasts, while also addressing the inherent challenges and considerations for actuaries.

AI IN ACTUARIAL ANALYSIS OF MORTALITY MODELING AND FORECASTING

Mortality modeling and forecasting are critical components of actuarial science, influencing insurance pricing, pension planning, and public health policy. Traditionally, these models have relied on historical data and statistical techniques to predict future mortality rates. However, the advent of AI offers the potential to significantly enhance these models by incorporating a wider array of data sources and more sophisticated analytical techniques.

AI, particularly machine learning (ML) algorithms, can process vast amounts of data at high speed, identifying patterns and correlations that may not be evident through traditional methods. For instance, neural networks can be trained on historical mortality data to predict future trends with high accuracy. Additionally, AI can incorporate real-time data, such as current health statistics, economic indicators, and environmental factors, providing a more dynamic and responsive approach to mortality forecasting.

NONTRADITIONAL DATA SOURCES IN AI-DRIVEN MORTALITY MODELING

One of the most significant advantages of using AI in mortality modeling is the ability to leverage nontraditional data sources. These include electronic health records (EHRs), social media activity, wearable technology data, and environmental sensors. These data sources can provide real-time, granular insights into factors affecting mortality that traditional datasets might overlook.

• Electronic Health Records (EHRs):

EHRs contain detailed medical histories of individuals, including diagnoses, treatments, and outcomes. By analyzing EHRs, AI can identify patterns related to chronic diseases, medication adherence, and other health-related behaviors that influence mortality. This can enhance the predictive accuracy of mortality models by incorporating individual health trajectories.

• Social Media Activity:

Social media platforms provide a wealth of data on individuals' lifestyles, behaviors, and mental health. Al can analyze these data to detect trends and anomalies that correlate with mortality risks. For example, changes in social media activity patterns might signal mental health issues or lifestyle changes that could impact an individual's mortality risk.

• Wearable Technology Data:

Devices like fitness trackers and smartwatches continuously monitor physiological parameters such as heart rate, physical activity, and sleep patterns. Al can aggregate and analyze this data to detect early signs of health issues, providing a proactive approach to mortality risk assessment.

• Environmental Sensors:

Environmental factors such as air quality, temperature, and pollution levels have a significant impact on public health. Al can integrate data from environmental sensors with health data to model how changes in the environment affect mortality rates. This can be particularly valuable for predicting mortality trends in response to climate change or natural disasters.

RISKS AND CHALLENGES OF UTILIZING AI IN MORTALITY MODELING

While the use of AI and nontraditional data sources offers considerable benefits, it also introduces several risks and challenges that must be carefully managed.

• Data Privacy and Security:

The integration of personal health data and social media information raises significant privacy concerns. Ensuring the security of these data and obtaining informed consent from individuals is paramount. There is a need for robust data governance frameworks that balance the benefits of data utilization with the protection of individual privacy.

• Bias and Fairness:

Al algorithms can inadvertently perpetuate or exacerbate existing biases in mortality modeling. For example, if the training data is not representative of the entire population, the model's predictions may be skewed. Actuaries must ensure that the data used is diverse and representative and that the AI models are tested for fairness and unbiased performance.

• Interpretability and Transparency:

Al models, particularly deep learning techniques, can be complex and opaque, making it difficult to interpret how they arrive at specific predictions. This lack of transparency can be a barrier to gaining trust and acceptance from stakeholders. Actuaries need to develop methods to interpret and explain Al models' outputs clearly and understandably.

• Regulatory Compliance:

The use of AI in mortality modeling must comply with existing and evolving regulatory frameworks. This includes data protection regulations such as the General Data Protection Regulation (GDPR) and industry-specific guidelines. Actuaries must stay informed about regulatory developments and ensure that their use of AI adheres to these requirements.

VALIDATION AND VERIFICATION OF AI MODELS

Given the high stakes associated with mortality predictions, it is crucial to validate and verify AI models rigorously. This involves several steps:

• Data Quality Assessment:

Ensuring the accuracy, completeness, and reliability of the data used to train AI models is the first step in validation. This includes checking for missing values, outliers, and inconsistencies.

- Model Testing and Evaluation: AI models should be tested on separate validation datasets to evaluate their predictive performance. Key metrics such as accuracy, precision, recall, and F1-score should be assessed. Cross-validation techniques can also be used to ensure the model's robustness.
- Sensitivity Analysis:
 Conducting sensitivity analysis helps to understand how changes in input data affect the model's predictions. This can identify potential vulnerabilities and areas where the model may be prone to error.
- Scenario Analysis: Running the AI model under different scenarios, including extreme events and outlier cases, can help assess its reliability and stability under various conditions.
- Peer Review and Auditing: Independent peer reviews and audits of AI models can provide an additional layer of validation. This helps to ensure that the models meet industry standards and best practices.

IMPLICATIONS FOR ACTUARIES AND INSURERS

The integration of AI into mortality modeling and forecasting will inevitably change the role of actuaries. Actuaries will need to develop new skills in data science and AI to effectively leverage these technologies. This includes understanding AI algorithms, data engineering, and the ethical implications of AI use.

• Skill Development:

Actuaries should invest in continuous learning and professional development to stay abreast of advancements in AI and data science. This may involve pursuing additional certifications or degrees in related fields.

- Collaboration with Data Scientists: Actuaries will need to work closely with data scientists and AI experts to develop and validate AI models. This interdisciplinary collaboration can help bridge the gap between traditional actuarial methods and modern AI techniques.
- Ethical Considerations:

Actuaries must be mindful of the ethical implications of using AI in mortality modeling. This includes ensuring fairness, transparency, and accountability in AI models and addressing potential biases and ethical dilemmas.

- Regulatory Compliance: As mentioned earlier, actuaries must stay informed about regulatory developments and ensure that their use of AI complies with relevant laws and guidelines. This includes participating in industry discussions and contributing to the development of new regulations and standards.
- Enhanced Decision-Making: Al can augment the decision-making process by providing more accurate and timely insights into

mortality trends. This can help insurers develop better risk management strategies, price products more accurately, and improve customer service.

CONCLUSION

The application of AI to mortality modeling and forecasting holds immense potential to transform actuarial practice. By leveraging nontraditional data sources and advanced analytical techniques, AI can enhance the accuracy and reliability of mortality predictions. However, the integration of AI also presents significant challenges, including data privacy, bias, interpretability, and regulatory compliance. Actuaries must navigate these challenges carefully, developing new skills and ethical frameworks to harness the benefits of AI responsibly. As the field evolves, continued research and dialogue will be essential to fully realize the potential of AI in actuarial science and ensure its positive impact on mortality modeling and forecasting.

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