

Machine Learning-Based Prediction Algorithms for the Mitigation of Maternal and Fetal Mortality in the Nigerian Tertiary Hospitals

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Abstract

Maternal and fetal mortality rates in Nigeria remain among the highest globally, posing a significant public health challenge. Despite efforts to improve healthcare infrastructure and access, these mortality rates persist at alarming levels. Recent advancements in Machine Learning (ML) have opened new avenues for addressing this issue by predicting and mitigating the risks associated with maternal and fetal health complications. This article reviews the current landscape of ML-based prediction algorithms in Nigerian tertiary hospitals, their potential impact on healthcare outcomes, future prospects, as well as the challenges and opportunities for its implementation.

Keywords: *machine learning, prediction algorithms, healthcare, maternal mortality, fetal mortality, Nigerian tertiary hospitals*

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I. Introduction

Nigeria's healthcare system faces substantial challenges in the reduction of maternal and fetal mortality rates. Despite numerous interventions, these rates remain alarmingly high due to factors such as inadequate healthcare infrastructure, limited access to quality care, and insufficient skilled healthcare personnel. Maternal and fetal mortality rates are crucial indicators of healthcare system effectiveness and accessibility. Despite efforts to enhance maternal and child health in Nigeria, mortality rates in tertiary hospitals remain high (Ikeoha et al., 2022). To address these challenges, leveraging technology like Machine Learning (ML) and Artificial Intelligence (AI) are quite essential (Khuluq, 2023). According to Okpala and Okpala (2024), The integration of AI to healthcare “entails the application of software and the algorithms of machine learning, to use input data to arrive at approximate conclusions, by mimicking the reasoning of humans for evaluation and perception of complicated medical data, in order to surpass man’s competence though the provision of efficient means of prevention, diagnosis, and treatment of diverse sicknesses.”

ML, a subset of AI, offers promising tools for predictive analytics, which can significantly enhance the early detection and management of high-risk pregnancies, ultimately improving maternal and fetal outcomes. ML algorithms have shown promise in healthcare applications, including predictive analytics for identifying high-risk pregnancies and reducing mortality rates (Ke et al., 2020). The integration of ML and AI technologies in healthcare, particularly in predicting maternal and fetal mortality, presents a promising avenue for improving outcomes in Nigerian tertiary hospitals. By leveraging these innovative approaches and insights from relevant studies, healthcare systems can work towards reducing mortality rates and enhancing maternal and child health.

Recent advancements in ML-based prediction algorithms offer hope for preventing maternal and fetal mortality in Nigerian tertiary hospitals (Kasim et al., 2022). Studies have shown that ML algorithms can effectively predict in-hospital mortality for various conditions, such as acute coronary syndrome (Ke et al., 2022). These algorithms utilize features like age, heart rate, and clinical data to stratify mortality risk (Akaba et al., 2021). ML models have also been developed to predict mortality in specific patient populations, such as elderly patients with sepsis (Ope, 2020).

In Nigeria, studies have focused on understanding the determinants and outcomes of obstetric emergencies like obstructed labor (Ikeoha et al., 2022). These studies highlight the need for innovative approaches to improve maternal and fetal outcomes. Research on maternal deaths in Nigerian hospitals has utilized tools like the WHO International Classification of Diseases Maternal Mortality (ICD-MM) to classify and analyze these cases (N et al., 2021).

Efforts to reduce maternal mortality in Nigeria must consider factors like maternal health services perception and experiences. Understanding these aspects can help to increase the utilization of maternal health facilities and improve outcomes. And addressing the third delay in obstetrics through tools like the Maternal Fetal Triage Index can positively impact maternal and neonatal outcomes.

II. Current Challenges in Maternal and Fetal Mortality

According to the Nigerian Bureau of Statistics (2021), Maternal Mortality Ratio (MMR) in Nigeria was estimated at 512 per 100,000 live births in 2018, while the neonatal mortality rate was around 37 per 1,000 live births. These figures underscore the urgent need for effective intervention strategies. Common causes of maternal mortality include hemorrhage, infections, high blood pressure during pregnancy, and complications during delivery, while fetal mortality is often attributed to preterm birth, birth asphyxia, and infections.

To address maternal and fetal mortality in Nigeria requires a multifaceted approach that considers medical, social, economic, and cultural factors. By leveraging insights from various studies and implementing evidence-based interventions, healthcare systems can work towards reducing mortality rates and improving maternal and child health outcomes in the country.

Maternal and fetal mortality rates in Nigeria, particularly in tertiary hospitals, pose significant challenges to the healthcare system (Sageer et al., 2019). These high mortality rates are influenced by various factors such as inadequate access to quality prenatal care, limited resources, socio-economic disparities, and cultural factors (Nwafor et al., 2020). Also, the lack of timely identification and intervention for high-risk pregnancies exacerbates these challenges, leading to preventable deaths (Ntoimo et al., 2018).

Studies have highlighted the prevalence and predictors of depression, anxiety, and stress symptoms among pregnant women during the COVID-19 lockdown in Nigeria, shedding light on the mental health aspects affecting maternal health (Olonade et al., 2019). It has shown that maternal mortality ratios (MMRs) remain high in referral hospitals in Nigeria, emphasizing the need for targeted interventions to address these concerning rates (Meh et al., 2019).

Efforts to address maternal mortality in Nigeria must consider a holistic approach that includes social, economic, and cultural factors alongside medical conditions (Azuh et al., 2017). Implementing evidence-based interventions at both hospital and community levels is crucial in tackling the underlying causes of maternal mortality (Egbon et al., 2022). And understanding women's perceptions of the causes of maternal mortality can provide valuable insights for designing effective interventions (Emetere and Oladimeji, 2022).

The utilization of facility-based antenatal care and delivery services in rural Nigerian communities plays a vital role in improving maternal and child health outcomes (Adegoke et al., 2022). Factors affecting maternal and neonatal mortality in northern Nigeria have been analyzed using multiple linear regression, highlighting the complex interplay of various determinants (Asogwa et al., 2022). Additionally, a Bayesian spatial modeling approach has been employed to assess under-five children survival times in Nigeria, identifying regions with severe mortality risks (Arisukwu et al., 2021).

III. Machine Learning-Based Prediction Algorithms

Machine learning algorithms can process vast amounts of data to identify patterns and predict outcomes more accurately than traditional methods. In the context of maternal and fetal health, these algorithms can be trained on datasets comprising patient demographics, medical histories, prenatal visit records, lab results, and socio-economic factors. Recent advancements in machine learning techniques have significantly impacted the development of prediction models for assessing maternal and fetal mortality risk during pregnancy and childbirth (Li et al., 2018). These models utilize a variety of data sources such as demographic information, medical history, laboratory tests, and imaging data to predict adverse outcomes (Tsur et al., 2020). Commonly employed ML algorithms in this domain include logistic regression, support vector machines, decision trees, random forests, and neural networks (Li et al., 2018). And, the integration of feature selection techniques and ensemble learning methods has been shown to enhance the predictive performance of these models (Li et al., 2018).

Supervised Learning Algorithms: These algorithms, including logistic regression, decision trees, and Support Vector Machines (SVM), have shown promise in predicting maternal and fetal complications. For example, logistic regression can predict the likelihood of preeclampsia based on factors like blood pressure and proteinuria levels.

Unsupervised Learning Algorithms: Techniques such as clustering and association analysis can uncover hidden patterns in patient data that might not be immediately apparent. For instance, clustering algorithms can group patients with similar risk profiles, facilitating targeted interventions.

Deep Learning: Neural networks, particularly deep learning models, have been effective in image analysis and pattern recognition. In healthcare, Convolutional Neural Networks (CNNs) can analyze ultrasound images to detect fetal abnormalities early.

Studies have demonstrated the effectiveness of ensemble learning algorithms in improving predictive performance. For instance, highlighted that ensemble learning, which combines various base learners like random forests, recurrent neural networks, and elastic nets, leads to better predictive performance (Li et al., 2018). Similarly, developed a machine-learning model integrating maternal risk factors with fetal biometry to predict shoulder dystocia, showcasing the utility of ML in predicting adverse outcomes (Tsur et al., 2020).

The application of ensemble machine learning models has been extended to various medical scenarios beyond maternal and fetal health. For example, developed an ensemble machine learning model to predict early mortality among patients with bone metastases of hepatocellular carcinoma, emphasizing the versatility of ensemble techniques in healthcare applications (Long et al., 2023). Additionally, focused on predicting COVID-19 mortality using machine learning models, underlining the broad spectrum of predictive capabilities offered by ensemble methods (Baik et al., 2022).

Research have demonstrated the wide-ranging applicability of ML algorithms in predicting mortality outcomes across various healthcare domains beyond maternal and fetal health. For example, ML models have been successfully utilized to predict 30-day mortality in mechanically ventilated patients, showcasing the versatility and effectiveness of these techniques in healthcare settings (Park et al., 2022). The AI algorithms have been applied to forecast mortality after burn surgery in critically ill burn patients, highlighting the potential of ML in improving patient outcomes and clinical decision-making (Park et al., 2022).

The importance of explainable machine learning has been underscored in predicting risk factors for mortality in critically ill burn patients after surgery (Park et al., 2022). This approach enables a better understanding of the factors contributing to mortality predictions, assisting clinicians in making well-informed decisions. Technically, the integration of AI in predicting mortality outcomes following cardiac surgery has been investigated, indicating a growing interest in utilizing machine learning for risk prediction in clinical practice (Benedetto et al., 2020).

Subsequently, the use of machine learning models in predicting mortality has expanded to various medical specialties. For instance, AI algorithms have been employed in gynecologic oncology to enhance clinical outcomes and support research (Mysona et al., 2021). And machine learning-based analytics have been used to predict length of stay, readmission, and mortality in the surgical management of colorectal cancer, illustrating the potential of AI in personalized patient care (Masum et al., 2022).

In the field of maternal health, systematic reviews have been crucial in examining various factors that impact maternal, fetal, and neonatal outcomes. These reviews have investigated the influence of elevated ambient temperature on pregnancy outcomes, revealing the complex relationship between environmental factors and health outcomes (Austin et al., 2015). Research has explored the effects of prenatal aerobic exercises on maternal and neonatal outcomes, highlighting the importance of lifestyle interventions in improving pregnancy outcomes (Austin et al., 2015).

Efforts to reduce maternal and fetal mortality in Nigeria require a thorough understanding of the determinants and challenges to effective healthcare provision. Systematic reviews have outlined the obstacles and facilitators in delivering emergency obstetric care, emphasizing the necessity of conducting maternal death and obstetric care audits to enhance maternal health outcomes (Austin et al., 2015). Studies have examined the factors affecting antenatal care, institutional delivery, and postnatal care services utilization in Nigeria, emphasizing the need to improve access to and utilization of essential maternal health services (Austin et al., 2015).

The availability of emergency obstetric care is vital in decreasing maternal mortality rates. Research has evaluated the functionality of emergency obstetric referral systems and the continuum of care in low-resource settings, stressing the importance of process mapping approaches to identify gaps and enhance service delivery (Ofosu et al., 2021). Additionally, assessments of health facilities' readiness to provide emergency obstetric care in various regions have underscored the significance of well-functioning facilities and referral systems in ensuring timely and effective care (Hou et al., 2022).

IV. Implementation Challenges and Opportunities

Integrating ML-based prediction algorithms into Nigerian tertiary hospitals will involve the following steps:

Data Collection and Management

The establishment of comprehensive and standardized Electronic Health Records (EHRs) is critical. Data from multiple sources, including antenatal care visits, laboratory tests, and delivery outcomes, should be aggregated and anonymized to create robust training datasets.

Algorithm Training and Validation

Collaborations with local and international research institutions can facilitate the development and validation of predictive models tailored to the Nigerian context. Ensuring that these models are trained on diverse datasets will enhance their generalizability and accuracy.

Clinical Integration and Decision Support

Integrating ML algorithms into clinical workflows through decision support systems can provide real-time risk assessments for healthcare providers. These systems can alert clinicians to high-risk pregnancies, suggesting appropriate interventions and monitoring strategies.

Capacity Building and Training

Training healthcare professionals in the use of ML-based tools is essential for effective implementation. Continuous professional development programs can help bridge the knowledge gap and foster acceptance among medical staff.

While ML-based prediction algorithms hold promise for improving maternal and fetal mortality outcomes in Nigerian tertiary hospitals, there are several implementation challenges that need to be addressed (Thorsen-Meyer et al., 2020). These challenges include issues related to the availability and quality of healthcare data, interoperability of electronic health records systems, constraints in healthcare infrastructure, and the necessity for clinician training and collaboration (Sageer et al., 2019). Moreover, ensuring the ethical use of AI in healthcare, encompassing aspects like data privacy and patient consent, is crucial (D'Souza et al., 2017).

Other challenges include the following:

Data Quality and Availability: Inconsistent and incomplete health records can impede the development of accurate predictive models. Efforts to improve data collection practices are necessary.

Technical Infrastructure: Many Nigerian hospitals lack the technical infrastructure required to implement advanced ML systems. Investments in healthcare IT infrastructure are crucial.

Ethical and Legal Considerations: Ensuring patient privacy and data security is paramount. Clear regulatory frameworks must be established to govern the use of patient data in ML applications.

Despite these challenges, there are numerous opportunities for leveraging ML in the prevention of maternal and foetal mortality. Collaborations among healthcare institutions, government agencies, and technology partners can facilitate data sharing and resource allocation (Kang et al., 2020). Investing in capacity building and education programs can empower healthcare professionals to effectively utilize ML tools in clinical practice (Makinde et al., 2020). Additionally, community engagement and awareness initiatives can help address socio-cultural barriers to healthcare access and utilization (Veisy et al., 2021).

Studies have demonstrated the potential of ML algorithms in predicting mortality outcomes in various healthcare settings beyond maternal and foetal health. For instance, machine learning models have been utilized to predict mortality in patients undergoing continuous renal replacement therapy, showcasing the versatility of these techniques in critical care settings (Berglöv et al., 2019). Additionally, AI algorithms have been employed to forecast mortality after burn surgery in critically ill burn patients, highlighting the potential of ML in improving patient outcomes and clinical decision-making (Meh et al., 2019).

Efforts to address maternal and foetal mortality in Nigeria require a comprehensive understanding of the determinants and barriers to effective healthcare delivery. Systematic reviews have explored challenges and enabling factors in the provision of emergency obstetric care, emphasizing the importance of conducting maternal death and obstetric care audits to drive improvements in maternal health outcomes (Dalugoda et al., 2022). Moreover, studies have investigated the determinants of antenatal care, institutional delivery, and postnatal care services utilization in Nigeria, underscoring the significance of improving access to and utilization of essential maternal health services (Mfateneza et al., 2022).

Despite these challenges, the potential benefits are significant. ML-based prediction algorithms can lead to earlier interventions, personalized care plans, and better resource allocation, ultimately reducing maternal and fetal mortality rates.

V. Future Directions

Future research in ML-based prediction algorithms for maternal and fetal mortality prevention should focus on personalized risk assessment models, real-time data integration, novel AI techniques, and cost-effectiveness evaluations. By addressing these key areas and leveraging collaborations and capacity building initiatives, healthcare systems can enhance maternal and child health outcomes in Nigerian tertiary hospitals.

ML-based prediction algorithms for maternal and fetal mortality prevention should focus on key areas to enhance healthcare outcomes in Nigerian tertiary hospitals (Oladapo et al., 2015). Personalized risk assessment models tailored to the Nigerian population are essential to improve the accuracy of predictions and interventions (Ntoimo et al., 2018). Integrating real-time data streams for dynamic risk prediction can enable timely interventions and improve patient outcomes (Pasquier et al., 2023). Evaluating novel AI techniques such as deep learning and reinforcement learning can enhance the predictive capabilities of models and provide more accurate risk assessments (Olamijulo et al., 2022). Additionally, studies examining the cost-effectiveness and scalability of ML-based interventions are crucial for informing policy-making and resource allocation decisions (Rabiu et al., 2018).

Collaborations between healthcare institutions, government agencies, and technology partners can facilitate data sharing and resource allocation to enhance the implementation of ML-based prediction algorithms for maternal and fetal mortality prevention (Mahmoud et al., 2023). Investing in capacity building and education programs can empower healthcare professionals to effectively utilize ML tools in clinical practice, leading to improved maternal and child health outcomes (Tukur et al., 2022). Community engagement and awareness initiatives play a vital role in addressing socio-cultural barriers to healthcare access and utilization, thereby improving maternal and fetal mortality outcomes (Begg et al., 2022).

Studies have highlighted the importance of addressing obstetric complications such as postpartum hemorrhage and pre-eclampsia to reduce maternal mortality rates in Nigeria (Takang et al., 2023). Understanding the trends and causes of maternal deaths in tertiary hospitals is crucial for developing targeted interventions and improving healthcare delivery (Okonofua et al., 2022). Quality improvement methods such as facility-based death reviews and clinical audits are essential for identifying gaps in care and implementing strategies to prevent maternal and neonatal deaths (Sotunsa et al., 2019).

VI. Conclusion

Advancements in machine learning offer a transformative opportunity to mitigate maternal and fetal mortality in Nigerian tertiary hospitals. By leveraging predictive algorithms, healthcare providers can identify high-risk pregnancies earlier and implement targeted interventions, thereby improving outcomes for mothers and their babies. Addressing the challenges of data quality, infrastructure, and training will be essential to fully realize the potential of these technologies in the Nigerian healthcare context. Continued research and collaboration are vital to drive progress and ensure that these innovations lead to tangible improvements in maternal and fetal health.

In conclusion, the implementation of machine learning-based prediction algorithms holds significant promise for mitigating maternal and fetal mortality in Nigerian tertiary hospitals. These advanced technologies offer a robust and innovative approach to early detection, risk assessment, and personalized intervention strategies, which are crucial in addressing the high rates of maternal and fetal mortality. By leveraging vast amounts of medical data, machine learning algorithms can identify patterns and risk factors that may not be apparent through traditional methods. This enables healthcare professionals to predict complications more accurately and take timely, informed actions to prevent adverse outcomes. The use of predictive analytics can optimize resource allocation, enhance the efficiency of prenatal care, and improve the overall quality of maternal and fetal healthcare services.

Furthermore, the integration of these algorithms into the existing healthcare infrastructure in Nigeria can help bridge the gap between limited medical resources and the growing healthcare demands. It can also support continuous monitoring and follow-up, ensuring that high-risk pregnancies receive the necessary attention and care throughout the gestation period. However, the successful deployment of machine learning-based systems in Nigerian tertiary hospitals requires addressing several challenges. These include ensuring data privacy and security, overcoming infrastructural and technological limitations, and providing adequate training for healthcare professionals to effectively utilize these tools. Additionally, it is crucial to tailor these algorithms to the local context by incorporating region-specific data and insights.

In summary, while the adoption of machine learning-based prediction algorithms for maternal and fetal health presents challenges, their potential benefits far outweigh the obstacles. With proper implementation and support, these technologies can play a pivotal role in reducing maternal and fetal mortality rates, ultimately leading to better health outcomes for mothers and their babies in the Africa's most populous nation.

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