
Predictive Analytics and Neural Networks in Assessing Endodontic Treatment Prognosis: A New Frontier in Digital Dentistry

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Abstract

The advent of digital dentistry has transformed endodontic practice, offering innovative tools to enhance treatment planning and prognosis assessment. Predictive analytics and neural networks have emerged as powerful technologies capable of analyzing complex patient and clinical data to forecast the outcomes of endodontic treatments. By leveraging machine learning algorithms, neural networks can interpret radiographic images, detect subtle pathological changes, and provide data-driven predictions of treatment success. Integrating predictive analytics with neural network models enables clinicians to make more informed decisions, reduce treatment failures, and personalize patient care. Despite the promising potential, challenges such as data quality, algorithm transparency, and clinical adoption remain. Continued research and development in this domain are essential to fully realize the transformative impact of AI-driven predictive tools in endodontics, paving the way for more precise, efficient, and patient-centered care.

Keywords: Predictive analytics, neural networks, endodontic prognosis, digital dentistry, artificial intelligence, treatment outcome prediction

I. Introduction

The field of endodontics has undergone a significant transformation in recent years due to the integration of digital technologies, enabling more precise diagnostics, treatment planning, and outcome assessment. Accurate prognosis is a critical component of endodontic treatment, as it informs clinical decision-making, reduces the risk of treatment failure, and enhances patient satisfaction (Singh, 2022). Traditional prognostic methods rely heavily on clinician experience and interpretation of radiographic findings, which may introduce variability and limit predictive accuracy.

Artificial intelligence (AI), particularly through the use of predictive analytics and neural networks, has emerged as a promising approach to overcome these limitations by analyzing complex datasets and identifying patterns that may not be apparent through conventional evaluation (Kaur, 2021). Neural networks, a subset of machine learning, can process large

volumes of clinical and imaging data to predict treatment outcomes, detect periapical pathology, and optimize retreatment strategies (Chen, Stanley, & Att, 2020).

By integrating predictive analytics with neural network models, digital dentistry is moving toward a more data-driven, personalized approach, enabling clinicians to make evidence-based decisions and improve overall treatment prognosis. This advancement marks a new frontier in endodontic care, emphasizing the role of AI as both a diagnostic and prognostic tool in modern dental practice.

II. Predictive Analytics in Endodontics

Predictive analytics refers to the use of statistical algorithms, machine learning techniques, and data modeling to forecast future outcomes based on historical and current data. In endodontics, predictive analytics has gained significant attention due to its potential to enhance clinical decision-making and improve treatment outcomes. By analyzing patient demographics, clinical findings, radiographic data, and previous treatment records, predictive models can estimate the likelihood of treatment success or failure, allowing clinicians to tailor interventions more precisely (Singh, 2022).

One of the key applications of predictive analytics in endodontics is in assessing the prognosis of primary and retreatment procedures. For example, machine learning algorithms can evaluate the risk factors associated with endodontic failure, such as periapical lesion size, canal morphology, and prior restorative history, providing clinicians with objective evidence to guide treatment planning (Kaur, 2021). Additionally, predictive models can support decision-making regarding whether to attempt retreatment or proceed with extraction, improving patient outcomes while optimizing clinical resources (Chen, Stanley, & Att, 2020).

Furthermore, predictive analytics contributes to personalized endodontic care by identifying patient-specific risk profiles. This allows for early intervention in high-risk cases, monitoring of disease progression, and the selection of appropriate treatment techniques and materials. The integration of large-scale datasets from electronic dental records and imaging systems strengthens the accuracy and reliability of these predictive models, highlighting their role as an essential component of modern digital dentistry (Singh, 2022).

III. Neural Networks and AI Models

Neural networks, a subset of artificial intelligence (AI), have become increasingly significant in digital dentistry due to their ability to model complex, non-linear relationships within large datasets. These computational models mimic the human brain's neural architecture, enabling the identification of patterns and correlations that may not be readily apparent through conventional analysis (Singh, 2022). In endodontics, neural networks are particularly valuable for interpreting diagnostic data, including radiographs, cone-beam computed tomography (CBCT) scans, and clinical records, to predict treatment outcomes and potential complications.

AI models can assist clinicians in detecting periapical lesions, root fractures, or anatomical variations with higher accuracy than traditional methods, thereby enhancing diagnostic reliability (Chen, Stanley, & Att, 2020). Moreover, neural networks can be integrated into clinical decision support systems to guide choices between retreatment and extraction, optimizing patient-specific treatment strategies (Kaur, 2021). The predictive capabilities of these models are strengthened by continuous learning from new patient data, which allows for ongoing refinement of prognostic accuracy.

Despite their advantages, the successful deployment of neural networks in clinical practice depends on high-quality datasets, algorithm transparency, and validation against real-world outcomes. Proper integration into endodontic workflows has the potential to revolutionize treatment planning, improve success rates, and advance personalized care in digital dentistry (Singh, 2022; Chen, Stanley, & Att, 2020).

IV. Integration of Predictive Analytics and Neural Networks

The integration of predictive analytics with neural networks represents a transformative approach in assessing endodontic treatment prognosis. Predictive analytics leverages historical and real-time clinical data, such as patient demographics, tooth morphology, and previous treatment outcomes, to identify patterns that may influence treatment success (Singh, 2022). Neural networks, particularly deep learning models, are capable of processing complex radiographic and clinical datasets to detect subtle features often imperceptible to the human eye, such as early periapical lesions or microstructural variations in root canals (Chen, Stanley, & Att, 2020).

When combined, these technologies provide a synergistic platform for precision dentistry. Predictive analytics can feed structured patient data into neural network models, enhancing their ability to generate accurate prognostic predictions. This integration enables clinicians to stratify risk, anticipate potential complications, and tailor endodontic treatment plans to individual

patients, ultimately improving success rates and reducing the likelihood of retreatment (Kaur, 2021).

Emerging digital tools that merge these methodologies, including AI-driven decision support systems, are already demonstrating improved predictive accuracy compared to conventional diagnostic approaches. By facilitating evidence-based, data-driven decisions, the integration of predictive analytics and neural networks positions digital dentistry at the forefront of personalized, outcome-oriented care (Singh, 2022; Chen, Stanley, & Att, 2020).

Conclusion

The integration of predictive analytics and neural networks in endodontic practice represents a significant advancement in digital dentistry. These technologies enable clinicians to analyze complex datasets, interpret radiographic findings with high precision, and provide evidence-based prognostic assessments, ultimately enhancing treatment outcomes (Singh, 2022). By combining artificial intelligence-driven models with predictive analytics, practitioners can make more informed decisions regarding retreatment or extraction, reducing the risk of clinical failure and improving patient care (Kaur, 2021). While challenges such as data standardization, algorithm transparency, and clinical adoption persist, ongoing research and technological refinement are steadily addressing these limitations (Chen, Stanley, & Att, 2020). Overall, predictive analytics and neural networks hold transformative potential for personalized endodontic care, offering a pathway toward more accurate, efficient, and patient-centered treatment planning in the evolving landscape of digital dentistry.

References

1. Singh, S. (2022). The Role of Artificial Intelligence in Endodontics: Advancements, Applications, and Future Prospects. *Well Testing Journal*, 31(1), 125-144.
2. Kaur, A. (2021). Clinical Decision Support Systems Using AI for Endodontic Retreatment vs. Extraction. *International Journal of Cell Science and Biotechnology*, 10(1), 5-9.
3. Chen, Y. W., Stanley, K., & Att, W. (2020). Artificial intelligence in dentistry: current applications and future perspectives. *Quintessence Int*, 51(3), 248-57.
4. Nafees, Samraiz & Shahzad, Khalid & Sibhai, Imad & Mashayekhi, Yashar & Khan, Sami & Mian, Sana & Zahid, Abra & Mohamed, Khaled & C, Gyanendra & Ibrar,
5. Muhammad & Sarmad Farook Al Hamdani, Mustafa. (2025). Predischarge Dysphagia Measured Using the Eating Assessment Tool-10 (EAT-10) and Its Association With 90-Day Aspiration Pneumonia and Hospital Readmission. *Cureus*. 17. 10.7759/cureus.93643.

6. Govindasamy, Vaisnavy & Mashayekhi, Yashar & Khan, Marium & Olakunori-Ovaga, Oluwaronke & Kumar, Naresh & Zafar, Mahnoor & Mustafa, Sufyan & Soman, Muhammad & Akbar, Muhammad & Arif, Timsal & Karim, Asiya & Murtaza, Muhammad & Bano, Sher. (2025). Association Between Gastrointestinal Symptoms and Anxiety Levels in Patients with Functional Dyspepsia. *Cureus*. 17. 10.7759/cureus.84810.
7. Mashayekhi, Yashar & Ali, Auj & Shafique, Muhammad & Akbari, Tariq & Naz, Falak & Shaukat, Muneeba & Malik, Moeed & Rehmat, Pasbaan & Junaid, Mehreen & Fahim, Mahnoor & Mubarak, Tamer. (2025). Association Between Preoperative Sleep Disturbance and Postoperative Delirium in Adults Undergoing Elective Surgery. *Cureus*. 17. 10.7759/cureus.85114.
8. Mashayekhi, Yashar & Amadi-Livingstone, Chibuchi & Timamy, Abdulmalik & Eish, Mohammed & Attia, Ahmed & Panourgia, Maria & Mital, Dushyant & Pearce, Oliver & Ahmed, Mohamed. (2025). Narrative Review on the Management of Neck of Femur Fractures in People Living with HIV: Challenges, Complications, and Long-Term Outcomes. *Microorganisms*. 13. 1530. 10.3390/microorganisms13071530.
9. Wandala, Abali & Ikram, Naila & Safiullah, Maryam & Soomro, Azeem & Mashayekhi, Yashar & Nawasrah, Amal & Ali, Aisha & Tariq, Nayyab & Malik, Rida & Jamshed, Shehrezad & Fatima, Mishal. (2025). Association Between Sleep Quality and Temporomandibular Disorders (TMD) Symptoms Among Adults: A Cross-Sectional Study. *Cureus*. 17. 10.7759/cureus.87818.
10. Mashayekhi, Yashar & Jadhav, Aneesh & Sarfraz, Minahil & Sachwani, Harmain & Khan, Mujadad & Sultan, Saher & Thorani, Mahek & Ashraf, Mahwish & Mustafa, Imtiaz & Yar, Ahmad. (2025). Role of Serum Magnesium Deficiency in Insulin Resistance Among Overweight and Obese Children: A Meta-Analysis. *Cureus*. 17. 10.7759/cureus.90604.
11. Mashayekhi, Yashar & Iguh, Chinenye & Baba-Aissa, Sara & Iqbal, Mishal & Nidiginti, Tejashree & Jalali, Rabia & Kashmoola, Ali & Abualhab, Mutaz & Niazi, Racha & Shaikh, Ayaan & Polackal, Jerin & Zahid, Ramsha. (2025). Exploring the Prevalence and Symptom Burden of Small Fiber Neuropathy in Patients With Diabetes Using the Small Fiber Neuropathy Symptoms Inventory Questionnaire (SFN-SIQ). *Cureus*. 17. 10.7759/cureus.93548.
12. Mashayekhi, Yashar & Baba-Aissa, Sara & Assefa, Amanuel & Mutamba, Francis & Nur, Aamir & Shahid, Zuhaib & Salimon, Naheemat & Habahbeh, Ahmad & Ali, Niamat & Shandi, Ibrahim & Niazi, Racha & Habib, Fatima. (2025). Depression and Anxiety as Predictors of Quality of Life in Osteoarthritis Patients. *Cureus*. 17. 10.7759/cureus.93872.
13. Shaik, Kamal Mohammed Najeeb. (2025). SDN-based detection and mitigation of botnet traffic in large-scale networks. *World Journal of Advanced Research and Reviews*. 10.30574/wjarr.2025.25.2.0686.

14. Ashraf, M. S., Akuthota, V., Prapty, F. T., Sultana, S., Riad, J. A., Ghosh, C. R., ... & Anwar, A. S. (2025, April). Hybrid Q-Learning with VLMs Reasoning Features. In *2025 3rd International Conference on Artificial Intelligence and Machine Learning Applications Theme: Healthcare and Internet of Things (AIMLA)* (pp. 1-6). IEEE.
15. Arefin, N. T. Z. S. (2025). Future-Proofing Healthcare: The Role of AI and Blockchain in Data Security.
16. Shuvo, M. R., Debnath, R., Hasan, N., Nazara, R., Rahman, F. N., Riad, M. J. A., & Roy, P. (2025, February). Exploring Religions and Cross-Cultural Sensitivities in Conversational AI. In *2025 International Conference on Artificial Intelligence and Data Engineering (AIDE)* (pp. 629-636). IEEE.
17. Arefin, M. A. O. S. (2025). Advancements in AI-Enhanced OCT Imaging for Early Disease Detection and Prevention in Aging Populations.
18. Sultana, S., Akuthota, V., Subarna, J., Fuad, M. M., Riad, M. J. A., Islam, M. S., ... & Ashraf, M. S. (2025, June). Multi-Vision LVMs Model Ensemble for Gold Jewelry Authenticity Verification. In *2025 International Conference on Computing Technologies (ICOCT)* (pp. 1-6). IEEE.
19. Arefin, S., & Zannat, N. T. (2025). Securing AI in Global Health Research: A Framework for Cross-Border Data Collaboration. *Clinical Medicine And Health Research Journal*, 5(02), 1187-1193.
20. Riad, M. J. A., Roy, P., Shuvo, M. R., Hasan, N., Das, S., Ayrin, F. J., ... & Rahman, M. M. (2025, January). Fine-Tuning Large Language Models for Regional Dialect Comprehended Question answering in Bangla. In *2025 IEEE International Students' Conference on Electrical, Electronics and Computer Science (SCEECS)* (pp. 1-6). IEEE.
21. Arefin, N. T. Z. S. (2025). AI vs Cyber Threats: Real-World Case Studies on Securing Healthcare Data.
22. Azmi, S. K. (2025). Voronoi partitioning for secure zone isolation in software-defined cyber perimeters. *Global Journal of Engineering and Technology Advances*, 24(03), 431-441.
23. Shaik, Kamal Mohammed Najeeb. (2025). Secure Routing in SDN-Enabled 5G Networks: A Trust-Based Model. *International Journal for Research Publication and Seminar*. 16. 10.36676/jrps.v16.i3.292.
24. Almazrouei, K. M. K., Kotb, R., Salem, O. A., Oussaid, A. M., Al-Awlaqi, A. M., & Mamdouh, H. (2025). Knowledge, Attitude and Practice towards Pre-Marital Screening and Consultations among a sample of students in Abu Dhabi, the United Arab Emirates: A Cross-Sectional Study.
25. Ojuri, M. A. (2025). Ethical AI and QA-Driven Cybersecurity Risk Mitigation for Critical Infrastructure. *Euro Vantage journals of Artificial intelligence*, 2(1), 60-75.
26. Mansur, S. (2025). AI Literacy as a Foundation for Digital Citizenship in Education. *JOURNAL OF TEACHER EDUCATION AND RESEARCH*, 20(01), 5-12.

27. Rahman, M. M. (2025). Generational Diversity and Inclusion: HRM Challenges and Opportunities in Multigenerational Workforces.
28. Azmi, S. K. (2025). Hypergraph-Based Data Sharding for Scalable Blockchain Storage in Enterprise IT Systems.
29. Prior, M. (2025). The Diaspora: Survival, Sacrifices, and the Misunderstood Heartbeat Of Africa: An analysis of migration, remittances, and identity across Nigeria, Ghana, and Togo. *International Journal of Technology, Management and Humanities*, 11(03), 26-28.
30. Karamchand, G. ZERO TRUST SECURITY ARCHITECTURE: A PARADIGM SHIFT IN CYBERSECURITY FOR THE DIGITAL AGE. *Journal ID*, 2145, 6523.
31. Gupta, N. (2025). The Rise of AI Copilots: Redefining Human-Machine Collaboration in Knowledge Work. *International Journal of Humanities and Information Technology*, 7(03).
32. Sanusi, B. O. (2025). Smart Infrastructure: Leveraging IoT and AI for Predictive Maintenance in Urban Facilities. *SAMRIDDHI: A Journal of Physical Sciences, Engineering and Technology*, 17(02), 26-37.
33. Karamchand, Gopalakrishna & Aramide, Oluwatosin. (2025). AI AND CYBERWARFARE. *Journal of Tianjin University Science and Technology*. 58. 10.5281/zenodo.16948349.
34. Vethachalam, S. (2025). Cybersecurity automation: Enhancing incident response and threat mitigation.
35. Ojuri, M. A. (2025). Quality Metrics for Cybersecurity Testing: Defining Benchmarks for Secure Code. *Well Testing Journal*, 34(S3), 786-801.
36. Lima, S. A., Rahman, M. M., & Hoque, M. I. Leveraging HRM practices to foster inclusive leadership and advance gender diversity in US tech organizations.
37. Sanusi, B. Design and Construction of Hospitals: Integrating Civil Engineering with Healthcare Facility Requirements.
38. Shaik, Kamal Mohammed Najeeb. (2025). Next-Generation Firewalls: Beyond Traditional Perimeter Defense. *International Journal For Multidisciplinary Research*. 7. 10.36948/ijfmr.2025.v07i04.51775.
39. Bilchenko, N. (2025). Fragile Global Chain: How Frozen Berries Are Becoming a Matter of National Security. *DME Journal of Management*, 6(01).
40. Karamchandz, G. (2025). Secure and Privacy-Preserving Data Migration Techniques in Cloud Ecosystems. *Journal of Data Analysis and Critical Management*, 1(02), 67-78.
41. Oni, B. A., Adebayo, I. A., Ojo, V. O., & Nkansah, C. (2025). Insight into Underground Hydrogen Storage in Aquifers: Current Status, Modeling, Economic Approaches and Future Outlook. *Energy & Fuels*.
42. Karamchand, Gopalakrishna & Aramide, Oluwatosin. (2025). AI AND CYBERWARFARE. *Journal of Tianjin University Science and Technology*. 58. 10.5281/zenodo.16948349.

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43. Azmi, S. K. Bott-Cher Cohomology For Modeling Secure Software Update Cascades In Iot Networks.
 44. Lima, S. A., & Rahman, M. M. (2025). Neurodiversity at Work: Hrm Strategies for Creating Equitable and Supportive Tech Workplaces. *Well Testing Journal*, 34(S3), 245-250.
 45. Samuel, A. J. (2025). Predictive AI for Supply Chain Management: Addressing Vulnerabilities to Cyber-Physical Attacks. *Well Testing Journal*, 34(S2), 185-202.
 46. Azmi, S. K. Retrieval-Augmented Requirements: Using RAG To Elicit, Trace, And Validate Requirements From Enterprise Knowledge Bases.
 47. SANUSI, B. O. (2025). LEVERAGING CIVIL ENGINEERING AND DATA ANALYTICS FOR ECONOMIC GROWTH: A CASE STUDY ON SUPPLY CHAIN OPTIMIZATION IN SPORTS FACILITY RENOVATIONS. *MULTIDISCIPLINARY JOURNAL OF ENGINEERING, TECHNOLOGY AND SCIENCES*, 2(1).
 48. Azmi, S. K. (2025). Kirigami-Inspired Data Sharding for Secure Distributed Data Processing in Cloud Environments.
 49. Sachar, D. (2025, May). Enhanced Machine Learning Approaches for Network Intrusion and Anomaly Detection. In 2025 Systems and Information Engineering Design Symposium (SIEDS) (pp. 426-431). IEEE.
 50. Sachar, D. (2025, May). Optimizing Transaction Fraud Detection: A Comparative Study of Nature-Inspired Algorithms for Feature Selection. In 2025 Systems and Information Engineering Design Symposium (SIEDS) (pp. 392-397). IEEE.
 51. Almazrouei, K. M. K., Kotb, R., Salem, O. A., Oussaid, A. M., Al-Awlaqi, A. M., & Mamdouh, H. (2025). Knowledge, Attitude and Practice towards Pre-Marital Screening and Consultations among a sample of students in Abu Dhabi, the United Arab Emirates: A Cross-Sectional Study.
 52. Kumar, K. (2025). Cross-Asset Correlation Shifts in Crisis Periods: A Framework for Portfolio Hedging. *Journal of Data Analysis and Critical Management*, 1(01), 40-51.
 53. Azmi, S. K. Zero-Trust Architectures Integrated With Blockchain For Secure Multi-Party Computation In Decentralized Finance.
 54. Karamchand, G. (2025). AI-Optimized Network Function Virtualization Security in Cloud Infrastructure. *International Journal of Humanities and Information Technology*, 7(03), 01-12.
 55. Gade, S., Kholpe, B. M., Paikrao, U. B., & Kumbhar, G. J. (2025). Enriching redistribution of power in EV Charging Stations through Deep learning. *International Journal of Scientific Research in Modern Science and Technology*, 4(1), 29-45.