

AI and Machine Learning in Dental Diagnostics

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Editorial

Received: 1-Dec-2025, Manuscript No. jds-25-177992; **Editor Assigned:** 3-Dec-2025, Pre QC No. jds-25-177992; **Reviewed:** 17-Dec-2025, QC No. jds-25-177992; **Revised:** 22-Dec-2025, Manuscript No. jds-25-177992; **Published:** 29-Dec-2025, DOI: 10.4172/2320-7949.13.4.012

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Citation: Alejandro Martínez, AI and Machine Learning in Dental Diagnostics. RRJ Dental Sci. 2025.13.012.

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INTRODUCTION

The landscape of dental diagnostics is rapidly evolving with the advent of artificial intelligence (AI) and machine learning (ML) [1]. These advanced computational technologies are transforming how dental professionals detect, diagnose, and plan treatments for oral diseases. Traditional dental diagnostics, primarily dependent on clinical examination and radiographic interpretation by dentists, can be limited by human variability and time constraints. AI and ML offer the potential to enhance accuracy, efficiency, and early detection, ultimately improving patient outcomes and streamlining clinical workflows.

Understanding AI and Machine Learning in Dentistry

AI is a branch of computer science that enables machines to perform tasks that typically require human intelligence. This includes pattern recognition, decision making, and predictive analytics. Machine learning, a subset of AI, involves training algorithms on large datasets so they can learn and improve from experience without explicit programming. In dental diagnostics, AI and ML systems analyze complex data – such as digital radiographs, cone beam computed tomography (CBCT) scans, and clinical records – to recognize patterns indicative of disease [2].

Applications in Dental Diagnostics

Caries Detection

Dental caries (tooth decay) is one of the most common oral diseases worldwide. Early detection is crucial to prevent progression and minimize invasive treatments. AI assisted diagnostic tools analyze radiographic images to detect subtle changes in enamel and dentin that may be overlooked by the human eye.

These systems can highlight suspicious areas and quantify lesion severity, aiding dentists in clinical decision making.

Periodontal Disease Assessment

Periodontal diseases affect the supporting structures of teeth and are a leading cause of tooth loss in adults. AI algorithms can evaluate periodontal bone levels on radiographs and identify patterns of alveolar bone loss associated with disease severity. Some advanced ML models also integrate clinical parameters like probing depths and bleeding indices to provide comprehensive risk assessments [3].

Oral Cancer Screening

Early detection of oral cancer significantly improves prognosis. AI tools have been trained on large image datasets – including intraoral photos and histopathological slides – to differentiate between benign and malignant lesions. These models assist clinicians by flagging suspicious lesions for further investigation, thereby enhancing early referral and intervention.

Temporomandibular Joint (TMJ) and Orthodontic Diagnostics

AI has applications in assessing TMJ disorders and orthodontic conditions. ML models process CBCT scans to detect structural anomalies and joint degeneration. In orthodontics, AI can automate cephalometric landmark identification, reducing analysis time and increasing measurement consistency. Such tools support treatment planning and progress evaluation.

Integration with Electronic Health Records (EHRs)

By integrating AI with EHR systems, dental practices can gain predictive insights into disease risk and treatment outcomes. Predictive models analyze patient history, demographics, and clinical variables to identify individuals at high risk for conditions like caries, periodontal disease, and oral pathology. This supports personalized prevention strategies and proactive care.

Benefits of AI Driven Diagnostics

AI and ML technologies offer several notable benefits:

Improved Accuracy: By reducing human error and variability, AI enhances diagnostic precision, especially in early or subtle disease stages.

Efficiency and Productivity: Automated analysis of radiographs and imaging data saves time for clinicians, allowing them to focus on patient communication and treatment planning.

Consistency: ML models apply the same criteria in every case, providing standardized assessments and minimizing subjective interpretation [4].

Early Detection: AI's ability to recognize patterns beyond human perception supports the identification of diseases at earlier stages, improving intervention success.

Challenges and Limitations

Despite promising advancements, several challenges must be addressed:

Data Quality and Bias

AI models are only as effective as the data on which they are trained. Poor quality images, limited datasets, or unbalanced representation of populations can introduce bias, leading to inaccurate predictions for certain groups.

Integration and Workflow Adoption

Incorporating AI tools into existing dental workflows and software systems can be technically complex. Clinicians require seamless interfaces that support, not hinder, patient care.

Regulatory and Ethical Considerations

AI diagnostic tools must comply with regulatory standards to ensure safety and efficacy. Additionally, ethical concerns arise around data privacy, informed consent, and accountability when AI systems inform clinical decisions.

Trust and Acceptance

Dentists and patients may be hesitant to rely on AI recommendations without clear understanding and transparency of how these systems operate. Building trust requires robust validation studies and clinician education.

Case Examples and Real World Impact

In many dental clinics and research settings globally, AI tools have already demonstrated real world impacts. For instance, deep learning algorithms trained on thousands of radiographic images have achieved diagnostic performance on par with experienced clinicians for caries detection and periodontal assessment. Academic institutions and dental software companies are collaborating to refine these technologies and validate them across diverse populations.

Future Directions

The future of AI in dental diagnostics is bright, with several emerging trends:

Multimodal AI Systems: Integration of data from radiographs, 3D imaging, intraoral scans, and clinical records will enable more holistic diagnostic insights [5].

Real Time Chairside AI: Portable devices equipped with AI could provide instant diagnostic feedback during patient consultations.

Continuous Learning Systems: AI platforms that evolve with new data inputs from diverse clinical settings will enhance performance and generalizability.

Personalized Preventive Care: Predictive models will support tailored preventive plans based on individual risk profiles, shifting focus from treatment to prevention.

Conclusion

AI and machine learning are redefining dental diagnostics by enhancing accuracy, efficiency, and early detection capabilities. These technologies complement clinical expertise, offering powerful tools that support data driven decision making and personalized patient care. While challenges related to data quality, integration, and ethical considerations remain, ongoing research and

technological advancements promise to further integrate AI into everyday dental practice. As AI systems become more sophisticated and widely adopted, they will play a pivotal role in shaping the future of oral healthcare — ultimately improving outcomes for patients around the world.

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