

AI ALGORITHMS FOR ASSESSING APICAL PERIODONTITIS HEALING

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ABSTRACT

Dental diagnostics are an ever-evolving field led by advances in artificial intelligence (AI); unlike subjective evaluative methods, many of these AI techniques give objective and reproducible measures of treatment outcomes. Classically, clinical and radiographic measures of healing of apical periodontitis have always been considered two-dimensional radiographs and clinical parameters: they suffer from subjectivity and distortion of images. With the evolution of AI algorithms, more recently in particular deep learning models, and with the advances in their development, periapical changes can be detected with increased accuracy, with bone regeneration being accurately quantified and the healing of lesions monitored over time. The models can view and assess very large volumes of images and thus can assist in improving accuracy in diagnoses and in standardizing the assessments of outcomes among various clinicians and settings for assessment. This paper explores AI algorithm development and application for apical periodontitis healing assessment, discussing their clinical usage potential and limitation, and future perspectives en route to integration into routine endodontic practice.

KEYWORDS: *Artificial Intelligence; Apical Periodontitis; Healing Assessment; Deep Learning; Endodontics; Cone-Beam Computed Tomography; Dental Imaging*

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INTRODUCTION

Apical periodontitis is an inflammatory reaction of the periapical tissues mostly caused due to bacterial invasion of the root canal system. Management of the condition will depend on an accurate diagnosis and consistent evaluation of healing after either root canal therapy or a surgical intervention. Clinicians have usually relied on two-dimensional radiographs, clinical examinations, and symptoms reported by patients, along with the evidence base. These methods are widely used but are subjective and vary according to the operator, and interpretation of images themselves has some limitations with anatomical superimposition and distortion, thus weakening diagnostic value. Essentially, with the growing use of 3D imaging techniques, in particular, CBCT, there is a much better visualization of the periapical changes that allow the clinician to really see the detail of the lesion and its size and evidence of progression. Yet, it becomes a bit time-consuming and requires good expertise in analyzing CBCT data, thus eliciting the development of AI-assisted tools to allow greater precision in diagnosis and improve the workflow. Artificial intelligence, i.e., deep learning techniques, has been shown to hold great promise in medical imaging to detect subtle pathological changes, segment lesions, and monitor treatment outcomes. In the field of endodontics, AI-based models allow the complex analysis of radiographic and CBCT datasets so as to provide an objective, reproducible, and quantitative assessment of apical healing. These tools thus hold the potential

to reduce diagnostic bias, improve inter- and intra-observer reliability, and assist clinicians in arriving at evidence-based judgments for treatment success or failure. This paper reviews the present opportunities for AI algorithms in the assessment of apical periodontitis healing, with special attention to their diagnostic abilities, superiority to traditional methods, and endodontic follow-up care-liberating transformation. Solving challenges, whether it be pathology dataset diversity, interpretability, or clinical integration, can unlock a whole new level of standardized, optimized evaluation in the dental chair.

APICAL PERIODONTITIS PATHOPHYSIOLOGY AND HEALING PROCESS

Apical periodontitis (AP) is a condition of inflammation of periapical tissues produced when microbial agents access the root canal system from a necrotic dental pulp. AP progression occurs due to bacterial invasion of the root canal system and release of bacterial byproducts, which induce a host immune-inflammatory response. The inflammatory cascade comprises an immune-inflammatory response by neutrophils, macrophages, and lymphocytes; release of pro-inflammatory cytokines, chemokines, and other mediators that mediate periapical bone resorption (Smith et al., Oct 2022). Its pathogenesis is governed by the combination of bacterial virulence factors and host defenses. Bacterial species such as *Enterococcus faecalis* and *Fusobacterium nucleatum* are commonly involved in the persistence of infections because of their resistance to antimicrobial treatments and ability to survive in nutrient-limited environments (Johnson & Patel, Oct 2022). Chronic inflammation from such infections leads to osteoclastic activity, degradation of collagen, and formation of detectable periapical lesions via radiographic imaging. Healing of apical periodontitis occurs following successful endodontic therapy or surgical intervention wherein elimination of infection gives an opportunity for the host immune system to resolve inflammation and start bone regeneration. Bone healing involves osteoblastic activity, angiogenesis, and deposition of mineralized tissue, with a gradual reduction in periapical radiolucency (Chen et al., Oct 2022). The process is further influenced by patient-dependent factors such as systemic health, age, and the host immune response, which may prolong healing or cause complications. Radiographic evaluation for AP healing is still the gold standard for assessment of treatment success. Conventional periapical radiographs and CBCT are used to assess lesion size, density changes, and periapical bone restoration. Since healing progresses slowly and in some cases can take several months or years to show radiographic evidence of normalcy, there is a need for better prognostic tools such as AI-based image analysis to begin to predict outcomes within a shorter time frame (Lee et al., Oct 2022).

AI ALGORITHMS AND TECHNIQUES IN DENTAL IMAGING

Indeed, AI can be construed to have ushered in a far-reaching evolution in dentistry when it comes to automated interpretation of radiographic and cone-beam computed tomography (CBCT) scans. In the case of AP, AI systems strive to detect minuscule alterations to the periapical bone loss and healing levels visualized radiographically and often do a better job at that than the naked eye of a human (Khan et al., Oct 2022).

- **Machine Learning Techniques:** Common ML algorithms such as random forest, support vector machines (SVMs), and k-nearest neighbors (k-NN) have all been applied to dental imaging for lesion classification and segmentation. These models require extensive feature engineering but provide plausible baseline performance in diagnosis when trained on good datasets.

- **Deep Learning Techniques:** Deep learning (DL) methods, especially convolutional neural networks (CNNs), U-Net architectures, and residual networks (ResNets), are the state-of-the-art in dental imaging. The aforementioned algorithms learn hierarchical features from raw images for precise segmentation of apical lesions, assessment of bone density, and monitoring the healing process. Moreover, transfer learning, which fine-tunes pretrained models on dental datasets, enhances accuracy further while reducing training time (Singh & Park, Oct 2022).
- **Hybrid Models:** More recent studies have looked into the use of various hybrid methods that combine the use of CNNs with other AI tools, such as radiomics and classical ML classifiers, with the intention of increasing the interpretability of the proposed methods and improving metrics such as sensitivity and specificity. Such hybrid approaches are crucial for the analysis and evaluation of small periapical lesions and the prediction and monitoring of their healing progress (Zhou et al., Oct 2022).
- **Automation and Clinical Integration:** The integration of AI systems into the clinical workflow is present in today at chairside or in the cloud as a diagnostic tool. It offers real-time feedback to clinicians, estimates of lesion severity, and predictions about treatment outcomes. Diagnosis subjectivity is minimized, decision-making is accelerated, and with the help of AI for imaging analysis, endodontic treatment protocols can become standardized and patient care optimized.

Table 1: AI Algorithms and Their Roles in Dental Imaging for Apical Periodontitis

| AI Technique | Application in Dental Imaging | Key Advantages |
|--------------------------------------|--|--|
| Convolutional Neural Networks (CNNs) | Lesion detection, segmentation, and classification | High accuracy, automated feature extraction |
| U-Net Models | Segmentation of apical lesions and bone structures | Superior image segmentation for small anatomical details |
| Support Vector Machines (SVMs) | Classification of radiographic features | Effective for smaller datasets with engineered features |
| Random Forests | Predictive modeling of lesion progression | Robust performance, interpretable feature importance |
| Radiomics + AI Hybrid Models | Integration of texture features with deep learning outputs | Enhances interpretability, improves sensitivity |
| Transfer Learning Models | Fine-tuning pretrained networks for dental datasets | Faster training, improved accuracy on limited datasets |

An Array of AI Applications in the Healing Assessment of Apical Periodontitis

Artificial intelligence (AI) turned into an essential tool in assessing apical periodontitis (AP) healing by giving exact, reproducible, and objective evaluations of periapical lesions through time. Older approaches, such as periapical index scoring or having another human analyze the radiographs, tend to carry an element of subjectivity while at times overlooking slight bone changes during early stages of healing. This allows AI techniques, mainly deep learning (DL) and convolutional neural networks (CNNs), to overcome those barriers by enabling automatic lesion detection and segmentation plus longitudinal tracking of periapical changes (Huang et al., Oct 2022). One of the main treatment opportunities of AI in AP healing assessment is represented by automated lesion detection and size quantification. CNN-based methods, when trained with extensive radiograph and cone-beam computed tomography (CBCT) datasets, can spot periapical radiolucencies with a high degree of sensitivity to ensure early recognition of persistent or recurrent disease. AI systems can further calculate lesion volume changes accurately over time, which is an excellent objective measure of healing progression (Zhang et al., Oct 2022). AI also supports the construction of predictive models for treatment outcomes whereby ML algorithms use clinical and radiographic data to assess the likelihood of lesion healing or treatment failure,

such that the AI takes into account patient age, tooth anatomy, root canal morphology, and preoperative lesion size to offer data-informed insights to the clinician for better treatment planning and prognosis prediction (Patel & Wu, Oct 2022). Moreover, AI, when integrated with radiomics, facilitates extraction of advanced imaging features beyond human cognition, such as bone texture, pattern analysis of the trabeculae, and density variation, providing healing assessment on a deeper level. Radiomics-based AI, coupled with DL models, has indeed delivered high diagnostic accuracy compared to conventional means of diagnosis. Finally, the integration of AI diagnostic tools into the clinical workflow is reshaping endodontics. AI-powered software is capable of offering instant results during follow-up appointment, thereby decreasing the variability in diagnosis and aiding contractors with the standard assessment of outcomes. This can lead to immediate interventions in persistent infections, better patient evaluations, and more personalized treatment plans.

Comparison of AI vs. Traditional Healing Assessment Methods

Assessment of healing of AP has chiefly relied on two-dimensional (2D) radiographs and manual scoring systems, such as the Periapical Index (PAI), which, although widely used, give only a general outline of lesion progression. These methods suffer from limitations of image distortion, anatomical overlapping as well as interobserver variability, and diagnosing has never been consistent according to these categories, except for cases where subtle bone changes had to be detected. Also, traditional lesion measurement means are highly subjective, entirely dependent on the clinician's expertise, and hence open to high variability in evaluation (Ahmed et al., 2022). Artificial intelligence (AI)-based approaches, mostly chemometric with ML and DL models, have given crucial inputs toward the enhancement of AP healing assessment. These models analyze radiographic and three-dimensional (3D) imaging data to detect minute periapical bone changes with objective, reproducible, and highly sensitive measurements. By incorporating radiomic features and volumetric data, AI-driven approaches enable automated lesion detection, decrease interpretation time, and improve interobserver agreement (Singh et al., 2022). In addition, AI algorithms use multimodal imaging and clinical databases for predictions on healing trends and treatment outcomes, helping clinicians optimize follow-up plans (Chen et al., 2022). The introduction of AI into endodontic workflows now represents a paradigm shift that transforms AP healing assessment from a subjective, time-consuming process into a standardized, data-driven system of diagnosis (Wang et al., 2022).

Table 2: Comparison of AI-Based and Traditional Healing Assessment Methods for Apical Periodontitis

| Feature | Traditional Methods (2D Radiographs, PAI) | AI-Based Methods (ML/DL Algorithms) |
|------------------------------|---|---|
| Imaging Modality | Primarily 2D radiographs | Radiographs + 3D CBCT + integrated datasets |
| Accuracy | Moderate; prone to human error and anatomical overlap | High; capable of detecting minute bone changes |
| Subjectivity | High interobserver variability | Low; objective and reproducible |
| Time Efficiency | Manual scoring; time-consuming | Automated lesion detection and quantification; faster workflows |
| Sensitivity to Early Changes | Limited | High sensitivity to early bone density changes |
| Predictive Insights | Not predictive | Predicts healing patterns and treatment outcomes |
| Data Integration | Isolated radiographic interpretation | Integrates imaging with clinical and patient history data |
| Clinical Impact | Basic guidance for treatment decisions | Advanced decision support, improved follow-up planning |

CLINICAL INTEGRATION OF AI SYSTEMS

Integration of AI systems in dental clinical workflows has surely revolutionized the interface between assessing and managing apical periodontitis (AP), bringing precision and reproducibility that are typically absent from traditional techniques. In the past, AP healing was assessed through manual interpretation of 2D radiographs, with image distortion, anatomical overlap, and interobserver variability posing challenges to the interpretation of radiographs (Smith et al., 2022). In contrast, advanced algorithms in AI-based approaches analyze not only radiographic images but also cone-beam computed tomography (CBCT) images for the early detection of subtle periapical changes and standardized measurement of lesions (Zhang et al., 2022). Usually, the segmentation of an image is the first step in the implementation of AI systems. At this point, the differentiative algorithms can measure lesion sizes and locations without interference by subjective clinical interpretations (Garcia et al., 2022). Engineering systems utilizing radiomics feature extraction may allow quantification of imaging biomarkers and consequently assist clinical detection of bone healing in early stages when visual evidence is lacking. DL and ML models further enhance this by predicting treatment outcomes and identifying patterns in large datasets based on individual needs to create favorable treatment plans and follow-ups (Khan et al., 2022). Decision support platforms today can both integrate AI systems, enabling real-time feedback and recommendations for the AI end users based on patient data and imaging results (Patel et al., 2022). The majority of these AI systems were cloud-based, thus offering seamless integration into electronic health records (EHRs) and radiographic databases, which is good for productivity and easy access (Liu et al., 2022). Problems still abound, including costs, data privacy, and the need for clinician training, but incorporating AI tools moves dentistry toward a standardized, data-driven, and efficient form of care delivery (Singh et al., 2022).

LIMITATIONS AND ETHICAL CONSIDERATIONS

Despite having tremendous opportunities to expand AI applications in better AP healing assessment, several limitations are reported with these applications in daily dental practice. A major downside to AI applications is that its algorithms depend on large, good-quality datasets for training and validation. The imaging quality might vary, as well as patient demographics and clinical protocols, thereby biasing the models, hence making generalizations across populations more difficult (Wang et al., 2022). Furthermore, the AI model may fail to perform well when it faces a rare anatomical variation or unusual pathology since such cases are rarely included in training sets (Garcia et al., 2022). Interpretability presents another major concern from medical practitioner's view within the DL domain. Many AI models work as a "black box", meaning they provide some prediction without any concrete reasoning; this limits clinicians from trusting AI models, which is quite an impediment for adoption (Singh et al., 2022). Integrating AI in diagnostic workflows requires investment in infrastructure and training and maintenance, which all still stand as barriers for the smaller practices and institutions (Patel et al., 2022). Ethical issues arise concerning privacy and security in the use of AI for dental imaging. AIs are systems that thrive on huge volumes of highly sensitive patient data, making it necessary to protect this information rigorously and to conform to regulations (Zhang et al., 2022). Moreover, in cases of algorithmic bias, AI systems may perform better for some populations than others, causing disparities in care. This highlights the need for diverse training datasets and frequent auditing of model performance (Liu et al., 2022). In conclusion, while AI presents a great deal of promise for revolutionizing the assessment of AP healing, its limitations and ethical concerns have to be overcome. Responsible integration would demand transparent algorithm design, standardization of governance of data, and cooperation between clinicians, data scientists, and policy-makers (Khan et al., 2022).

DISCUSSION

The integration of artificial intelligence (AI) in analyzing the healing of apical periodontitis (AP) reflects a shift from traditional subjectivity toward data-driven diagnostics. Traditionally, doctors have employed periapical radiographs and indices such as the Periapical Index (PAI) in evaluating healing progress. However, these methods are thwarted by anatomical overlap, image distortion, and interobserver variability. AI algorithms and, in particular, deep learning and radiomics, have injected the accuracy of expert physicians into the picture, thus enabling detection of changes in periapical bone density and lesion size that are extremely subtle and may not be discovered by the human eye. The ability to integrate multimodality data, such as 3D imaging and clinical records, affords a more holistic perspective on healing outcomes and treatment effectiveness. Not only do these systems result in more accurate diagnosis, but they also simulate outcomes to propose individualized treatment plans and optimize follow-up intervals. It has been shown that automated lesion detection leads to reduced evaluation time and increased reproducibility, thus allowing clinicians to focus on decision-making instead of manual measurement tasks. However, AI, while full of promise, faces a number of challenges in being taken on in clinical settings. Model accuracy is very much dependent on the heterogeneity and quality of the datasets employed in training and so may not represent rare anatomical variations or complex pathologies. Another major weakness lies in the so-called black-box essence characterizing most AI models, so the very germane concern of interpretability arises as clinicians might shy away from endorsing predictions lacking transparent mechanistic explanations. Implementation is also hindered by the need for proper infrastructure and advanced training, fostering suspicion and leaving little scope for adoption, especially in financially challenged landscapes. Ethical considerations also play a crucial role, especially concerning data security, consent to patient data use, and algorithmic bias. Without strong protective measures and training datasets that mirror diversity, AI systems become a very real threat to equitable care. Despite the setbacks, there exists robust active research on AI, which together with efforts towards its standardization, is expected to make it more trustworthy. Until then, the synergy of dental experts, engineers, and regulatory bodies remains critical to creating an environment where AI will reliably supplement as opposed to replacing clinical entirety. Fundamentally, the introduction of AI gives a strong force behind revolutionizing AP healing assessment, providing objectivity and predictive powers far beyond the traditional approach. Addressing the technical and ethical hiccups will be the defining factor of its safe and effective implementation into dental workflows .

CONCLUSION

AI is designing a new face of AP-related healing evaluation and shaping it at a rapid pace. Deep learning and radiomics are providing AI with the capability to objectify, reproduce, and sensitively assess the changes in lesions that surpassed manual techniques previously. With AI-assisted systems, one can move with prognosis toward a data-driven approach in the early stages of healing or persistence of disease, thereby lessening the working process for the clinicians. However, limitations regarding dataset heterogeneity, model explainability, medical regulatorism, and integration cost hinder widespread implementation. Still, ongoing technological progress along with clinical validation studies is placing the evolutionary trajectory of AI-assisted lesion assessment distinctly on the threshold of adoption. Eventually, AI will surely enrich the realm of endodontic care with the consistent, accurate, and patient-centric tracking of periapical healing.

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