

AI-ENHANCED DETECTION OF ACCESSORY CANALS

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ABSTRACT

Accessory canals, which are anatomical variations, are often hindrances to a root canal therapy and upon failure to detect them bring about persistent endodontic infections. They have to be identified for effective cleaning, shaping, and obturation; nonetheless, conventional imaging methods and clinical probing often cannot detect these canals because of their small size and complex positioning. But the advent of artificial intelligence (AI) has brought about advanced tools for enhancing the detection of accessory canals by analyzing high-resolution radiographic and cone-beam computed tomography (CBCT) data. AI-based models, especially those built on deep learning algorithms, go far beyond just detecting minute anatomical details; they predict the presence of accessory canals, estimate their orientation, and finally assist clinicians to implement a more complete treatment plan. Such systems combine data obtained from imaging with patient-specific clinical factors to present objective, reproducible, evidence-based treatment plans, thus reducing the chance of a missed canal and enhancing the prognosis and sustenance of the treatment.

KEYWORDS: *Artificial Intelligence, Accessory Canals, Deep Learning, Cone-Beam Computed Tomography, Endodontic Anatomy, Root Canal Treatment, and Diagnostic Accuracy*

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INTRODUCTION

Accessory canals are small, often intricate pathways or channels which stem from the main root canal and open into the periodontium. This much varying with tooth groups and populations is clinically considered significant since such canals may harbor residual pulp tissue, bacteria, and necrotic debris that may remain even after rendering the usual root canal treatment. When accessory canals are not detected and treated, it is an acknowledged cause of an endodontic treatment failure, the appearance of periapical lesions, and persistent post-treatment manifestations. Traditionally, accessory canals are mainly identified through two-dimensional periapical radiographs and tactile exploration during instrumentation. Yet, due to various limitations posed by these techniques-the superimposition of anatomical structures, lower spatial resolution, and operator-dependent variability-these canals, especially in posterior teeth with a complicated morphology, are usually missed.

One of the best methods to visualize root canal morphology in three dimensions yet still difficult to interpret is CBCT. Inaccurate detection can result from variability in the clinician's experience for subtle anatomical features and small caliber accessory canals. Meanwhile, AI systems have entered the field as powerful adjuncts in enhancing the detection of accessory canals. By incorporating deep learning models such as convolutional neural networks, AI systems could analyze high-resolution images to discover minute irregularities, hardly perceptible even to trained human eyes. Integrating such

information with clinical knowledge may give these systems an edge in prescribing objective, reproducible, evidence-based insights that contribute to comprehensive treatment planning.

CLINICAL SIGNIFICANCE OF ACCESSORY CANALS

Accessory canals are minor branches of the main root canal that communicate with the periodontium. Depending on the tooth type, population, and anatomical study involved, their existence varies (Popescu & Müller, 2023). These canals are often overlooked but are critical to endodontic success for they become the reservoirs of residual pulp tissue, bacteria, and necrotic debris even after thorough cleaning and obturation of the main canal (Fischer et al., 2022). Untreated accessory canals, therefore, may remain as pathways for infection that, in turn, may lead to periapical pathology and postoperative problems, interfering with treatment and long-term survival of the tooth (Ng et al., 2011; Pak & White, 2011).

Epidemiological studies indicate that accessory canals do exist more commonly in molars, especially in mandibular ones, and less so in anterior teeth. Certain sites are favored by the canals, like the region of the apical third or furcation (Johansson, 2019). In other words, the anatomical complexity stands in the path of complete disinfection of the root canal system. Inability to address these canals has been associated with persistent periapical lesions and, in retreatment cases, a decreased prognosis (Rossi, 2022).

Accessory canals present several clinical difficulties. Numerous accessory canals are so minute and hidden from the standard tactile approach of the clinician or from the usual periapical radiographs, which do not have enough resolution or three-dimensional conformations (Patel et al., 2020). Their invisibility translates into the opportunity for untreated routes to serve as reservoirs for microorganisms, thus, causing persistent infection or secondary endodontic pathology (Setzer et al., 2020). Besides, conventional methods of obturation might not suffice to seal such small canals, opening avenues for micro leakage (Tsisis et al., 2019). Furthermore, bacterial colonization through accessory canals to periapical tissues will ignite inflammation and delay healing even after the main canal has been appropriately treated (Torabinejad et al., 2018).

It further involves clinical implications related to the decision-making process for retreatment, surgical intervention, or extraction. Persistent periapical lesions in a tooth bearing an accessory canal may be considered for surgical intervention if nonsurgical retreatment fails to properly reach or address such areas (Fernández et al., 2021). Historically, accessory canals have been explored and identified through in vitro studies by dye penetration, clearing techniques, and micro-computed tomography (micro-CT), none of which may be carried out in routine clinical practice-Johansson (2019). Recently, with the CBCT on the horizon, our visual powers are improved; however, faint accessory canals may yet remain undetected because of overlapping structures, contrasting intensity, or plain inexperience with their identification (Fischer et al., 2022).

Then, beyond these tools of identification and display, accessory canals really have significance for patient-centered care. By being able to evaluate the complexity of the canal system, a clinician can better communicate the possible risks and treatment limitations to the patient; such synergy between the patient and the clinician is the basis of informed consent and shared decision-making (Popescu & Müller, 2023).

LIMITATIONS OF CONVENTIONAL DETECTION METHODS

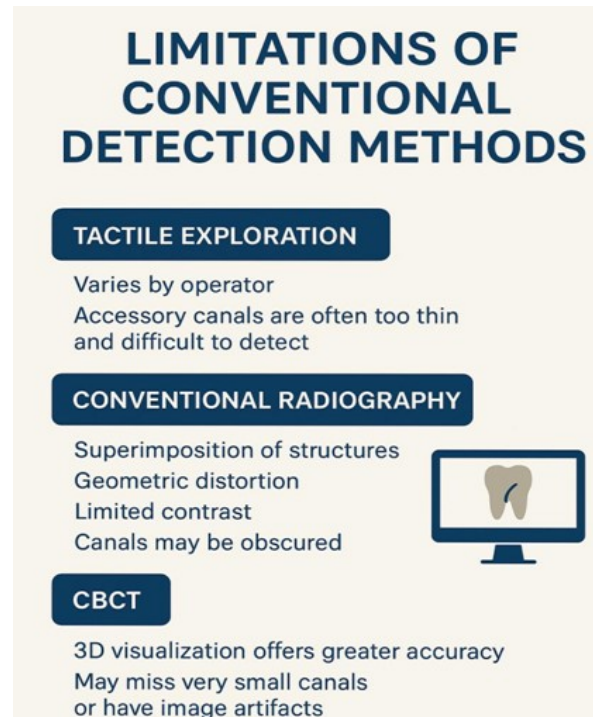
The identification of accessory canals has traditionally depended on two approaches: tactile exploration during instrumentation and conventional radiography. Despite the long-standing existence of these methods in endodontic practice, they present particular limitations that restrain the ability to reliably detect accessory canals, especially if small, complex, and located in difficult anatomical areas.

Tactile exploration varies from one operator to another, as it is influenced by the experience and tactile sensitivity of the clinician, along with depth perception of canal morphology. Accessory canals, particularly lateral canals or apical ramifications, are generally too thin to be felt during instrumentation. Most investigators opine that their narrow diameters and tortuous trajectories do not allow file introduction into these canals, thereby leaving them untreated (Ng et al., 2011). But even with careful probing, tactile methods cannot generate a full map of the canal system; consequently, slight anatomical variants may remain unnoticed, and this can lead to continued infection and decreased case success (Pak & White, 2011).

Since the first edition of Cone Beam Computed Tomography- A Paradigm Shift in Dentistry, CBCT represents a paradigm beginner in dentistry. Conventionally, radiographic techniques involving periapical radiography were employed for noninvasive diagnostic processes of endodontic assessment. However, these two-dimensional images have inherent limitations due to superimposition of structures, geometric distortion, and lesser contrast in identifying small or oblique accessory canals (Patel et al., 2020). Overlapping roots or dense bone can mask accessory canals, more so in posterior teeth, preventing the accurate detection of the canal, unless the clinician is highly experienced. Furthermore, periapical radiographs offer only a static snapshot, passing through three-dimensional depth information of complex canal anatomy and adding to the impediment of visualization (Setzer et al., 2020).

Since the image is a 3D visualization, CBCT has real advantages in detection. Cross-sectional and volumetric imaging delineates complex canal systems with higher accuracy than regular radiographs (Fischer et al., 2022). Nevertheless, there still exist limitations during CBCT interpretation. For instance, an accessory canal may be too small in diameter and fall below the spatial resolution capability of a certain scanner. Furthermore, image artifacts caused, for example, by beam hardening or scatter, may reduce the visibility of a faint structure. Interpretation is also highly dependent on the level of expertise of the clinician; hence, several studies have indicated a significant inter-observer discrepancy in identifying accessory canals on CBCT images (Johansson, 2019).

Preceding conventional techniques, staining, clearing, or micro-CT analyses were very accurate in vitro but became impractical for routine clinical use. These methods require specialized equipment; some are destructive; others are time-consuming; and they cannot be used in vivo, thus limiting utility for patient care (Popescu & Müller, 2023). On the contrary, clinicians' resort to a blend of tactile and radiographic evaluations, which on one hand are valuable, on the other hand, are still not sufficient for the consistent detection of accessory canals.

**Figure 1**

INTEGRATION OF IMAGING DATA AND CLINICAL PARAMETERS

Only a thorough evaluation and thoughtful combination of imaging diagnosis with pertinent clinical data will ever lead to the accurate detection of accessory canals. The standard imaging techniques such as periapical radiographs and CBCT offer excellent information concerning the structure; however, their value is greatly increased when the data are correlated with medical records of the patient. Based on correlating both sources of information, it allows clinicians to give context to imaging findings, thereby improving their predictability and treatment planning.

Thanks to its three-dimensional capabilities of cross-sectional and volumetric views, CBCT imaging has been brought forth to become the ultimate platform for visualizing complex canal systems, displaying minute anatomical variations, including lateral and accessory canals (Fischer et al., 2022). However, CBCT interpretation stands insufficient on its own since voxel size, image artifacts, and operator experience influence detection capabilities (Johansson, 2019). Clinicians can consider clinical parameters, including patient age, history of endodontic treatment, presence of periapical lesions, and systemic health conditions, to frame a more complete picture as to how accessory canals might occur and be of clinical significance (Ng et al., 2011; Pak & White, 2011).

AI facilitates the further integration of a predictive model that combines imaging data with clinical variables. Machine learning and deep learning algorithms can analyze CBCT scans to detect subtle features suggestive of accessory canals while casting a wide net of clinical factors that could be associated with the prevalence or intricacies of canals (Popescu & Müller, 2023). The advantage of using this multimodal analytical approach is that it offers more objective, reproducible, and data-driven measures to evaluate cases than do the traditional methods relying on the human eye and tactile inspection. The AI models, by way of flagging these high-risk cases, indicate probable canal locations and assist the clinician in devising strategies for better cleaning, shaping, and obturation.

The combination of imaging and clinical data also supports prognostic evaluations. In fact, studies highlight clinical features such as a history of treatment failure, persistent periapical pathology, or anatomically complex roots, all providing a raise in the possibility of accessory canals (Rossi, 2022). When combined with imaging findings, clinicians are then more able to evaluate timely challenges and anticipate complications from an accurate point of view. Thus, intervention strategies can be designed appropriately, including alterations in canal preparation, additional irrigation methods, or specialized obturation materials to seal the accessory pathways more effectively.

CHALLENGES AND LIMITATIONS OF AI ADOPTION IN ACCESSORY CANAL DETECTION

Beside overshooting potential, AI adoption in detecting accessory canals faces numerous challenges and limitations that must be tackled for beneficial clinical implementation. Each of these challenges is located in a technical, clinical, ethical, or practical domain, thus weighing upon the issues whether an AI system could be deemed reliable, usable, and acceptable in endodontic practice.

DATA AVAILABILITY AND QUALITY

Training AI models, particularly those based on deep learning architectures, requires large datasets having firm annotations. When it comes to endodontics, though, high-quality datasets, including important representations of accessory canals, tend to be scarce. Many of the CBCT datasets were collected by one single institution or, at best, a population-specific basis; this thereby decreases the generalizability of AI prediction across heterogeneous groups of patients (Popescu & Müller, 2023). In addition to these challenges, the variability in imaging protocols, voxel sizes, and types of CBCT scanners further impedes model training and may contribute bias or inconsistencies at model detectability (Fischer et al., 2022). Lastly, the clinical documentation that comes along is hardly ever uniform, hence making it difficult to interlink the imaging data with pertinent patient information.

SMALL SIZE AND COMPLEXITY OF ACCESSORY CANALS

Accessory canals are invariably narrow, tortuous, and found in varied locations, barely conforming to the spatial resolution of standard CBCT scanning. The detection of these minimal structures remains challenging even aided by AI. Deep learning models can be trained to find faint patterns in image data, but there can be false negatives, especially with canals smaller than voxel resolution or in overlapping anatomies (Johansson, 2019). False positives also occur and can drive into interventions or over-treatment that are not warranted.

INTERPRETABILITY AND THE “BLACK BOX” PROBLEM

Many AI models, especially CNN models, are black boxes that spit out their predictions without conveying the rationale for such a decision. Clinicians may shy away from recommendations that cannot be fully explained or justified in clinical and legal terms (Meyer et al., 2021). When interpretability is lacking, AI systems suffer a loss of trust and are less likely to be integrated into decision-making workflows, especially in tricky cases where an exercise in clinical judgment is paramount.

INTEGRATION INTO CLINICAL WORKFLOW

For successful integration of AI, there must be seamless cohesion with clinical workflows. Several of the platforms require proprietary software, computational resources, or manual preprocessing of data, all of which consume the precious time of clinicians and thereby disturb their routine working (Patel et al., 2020). For smaller clinics or practices lacking IT support, those requirements equate to big barriers. Besides, it must also be compatible with the imaging systems already in place and the electronic health records so that workflow inefficiencies are not introduced.

ETHICAL AND LEGAL CONSIDERATIONS

Alongside AI adoption come major ethical and legal considerations. Patient privacy is an important subject of concern, especially when these large imaging datasets are being used for training an AI algorithm. This requires data protection laws and regulations such as the GDPR or HIPAA to be complied with. Furthermore, if an error occurs in diagnosis, it is less clear who should be held accountable. Consider, for example, a situation in which an AI system fails to detect an accessory canal, and treatment consequently fails: the clinician, software developer, or the institution itself might all be considered at fault (Tsisis et al., 2019).

COST AND ACCESSIBILITY

The investment required for AI-assisted detection of accessory canals can be high. In other words, high-resolution CBCT, specialized software, and cloud computation can be expensive options for smaller clinics or practices in low-resource settings (Torabinejad et al., 2018). This disparity stands the risk of imparting inequalities in the provision of advanced diagnostic tools.

CLINICIAN ACCEPTANCE AND EDUCATION

Finally, acceptance by clinicians is a key aspect. Some practitioners may resist integrating AI into their work processes for various reasons: fear for loss of autonomy, distrust of algorithms, or unfamiliarity with AI (Popescu & Müller, 2023). Therefore, we need educational programs that train instructors to help clinicians interpret AI outputs and understand their limitations so that they may know how best to apply the recommendations in patient care.

To conclude, while there is much promise for AI in aiding the identification of accessory canals, concerns with data quality, canal complexity, interpretability, workflow integration, ethics, cost, and clinician acceptance remain. Removing these roadblocks would help see that AI tools for endodontics are developed fairly and genuinely, resulting in better outcome-oriented treatments and patient care.

DISCUSSION

An extra canal hitting the limelight is a major challenge in endodontic treatment. Very minute in size, with complex morphology and circumstantial location, they often elude detection, thus severely jeopardizing the treatment's prognostic potency and becoming a potential source of persisting infection. However, restrictions have long been placed on traditional methods, encompassing everything from tactile perception and two-dimensional radiography, through to three-dimensional radiography using CBCT which is limited in terms of resolution, artifacts, and application by the operator's skill. AI implementation in this field offers significant bolstering of diagnostic capabilities, thus empowering an objective and reproducible evaluation to supplement the clinical acumen.

The sort of pattern recognition AI, especially deep learning: more complex and rarer big data tables cannot be handled well by AI. When these models are given high-resolution CBCT scans, they can detect rather subtle anatomical characteristics that human operators might miss. By learning from thousands of images, these models can predict the likely position, trajectory, and even dimensions of accessory canals. This returns to those worrying clinicians who ahead of time think of places to instrument, thereby allowing cleaning and obturation to be tailored to the actual canal complexity. Thus, a well-fitting treatment plan, based on the knowledge of those complexities, reduces post-treatment infection and provides for a better prognosis.

AI also assists with clinical decision-making by combining imaging findings with patient-specific variables. Previous treatment history, periapical pathology, and tooth morphology may all be feeding into the imaging findings for a more global assessment of treatment difficulties. It then allows for a better-informed choice of instrumentation techniques, irrigation strategies, and obturation materials, all customized to the peculiarities of each patient's anatomy. While this combination of AI-assisted imaging analysis with clinical judgment forms the basis of evidence-based tailored endodontic care.

Another major advantage of AI-assisted detection is the increased potential for standardization in care. Endodontics has, for long, been marred by the variability in human skill in detecting the accessory canals. Detection rate varies with clinical experience and expertise and familiarity with the canal morphology. AI models minimize such variability and provide an assurance that high-risk anatomical features are consistently detected by applying the same criteria to all cases. Standardization really gets critical in cases of retreatment, where the source of persistent pathology is likely to be the accessory canals yet to be treated..

CONCLUSION

Accessory canals represent an important nuance in root canal anatomy that is often neglected, posing a major challenge to persistent apical infection and treatment failures if undetected. These subtle, complex channels elude reliable location using the accepted and traditional methods: tactile agony and simple two-dimensional radiography. Despite a three-dimensional imaging modality such as CBCT, the dilemma of interpretation variability and resolution remains. Artificial intelligence has come into being in the detection of accessory canals, which may offer objective, reproducible, and data-driven decision support.

These AI programs leverage deep learning to acknowledge subtle anatomical variations in high-resolution imaging data that may escape human observation. Holistic tooth evaluation is then formed when AI is aided with the relevant clinical parameters, forming the basis for individualized treatment planning, improved cleaning and obturation strategies, and thus predictably better outcomes. AI has its own unique way of promoting careful standardization, thus reducing the variability with which a practitioner may identify high-risk anatomical features. Patient-centered care is further supported by AI-assisted visualization tools that foster communication and informed consent, ensuring patients grasp treatment intricacies and possible risks.

Despite all this, the adoption of AI for accessory canal detection runs into various challenges such as quality of data, the very small canal dimensions, integration into workflow, interpretability, ethics, and cost. Robust validation, clinician training, and user-friendly systems will resolve these challenges, thus facilitating their responsible and effective implementation.

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