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| **Source:** | | TG-Dental Topic Drivers | | |
| **Title:** | | Att.1 – TDD update (TG-Dental) | | |
| **Purpose:** | | Discussion | | |
| **Contact:** | | Falk Schwendicke TG-Dental Charité – Universitätsmedizin Berlin, Germany | | Email: falk.schwendicke@charite.de |
| **Contact:** | | Joachim Krois TG-Dental Charité – Universitätsmedizin Berlin, Germany | | Email: joachim.krois@charite.de |
| **Contact:** | | Tarry Singh TG-Dental deepkapha.ai, Netherlands | | Email: [tarry.singh@deepkapha.ai](mailto:tarry.singh@deepkapha.ai) |

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| **Abstract:** | This topic description document (TDD) specifies a standardized benchmarking for AI-based Dental Diagnostics and Digital Dentistry. It covers all scientific, technical, and administrative aspects relevant for setting up this benchmarking (and follows the template structure defined in document FGAI4H-J-105). The creation of this TDD is an ongoing iterative process until it is approved by the Focus Group on AI for Health (FG-AI4H) as deliverable No. 10.17. This draft will be a continuous input- and output document. |

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| **Change notes:** | Version 4.2 (submitted as FGAI4H-M-010-A01 for e-meeting M; remote)   * Onboarding of new contributors to the topic group * Expanding on section 3 by structuring subtopics along the clinical care pathway and along subfields of dentistry.   Version 4.1 (submitted as FGAI4H-L-010-A01 for e-meeting L; remote)   * Onboarding of new contributors to the topic group * Inserting figure 1 that visualizes the thematic structure of subtopics along the clinical care pathway and along subfields of dentistry.   Version 4.0 (submitted as FGAI4H-K-010-A01 for e-meeting K; remote)   * Updating the template structure as defined in document FGAI4H-J-105   Version 3.0 (submitted as FGAI4H-J-010-A01 for e-meeting J; remote)   * Updates for section 2, AI4H Topic group.   + Expanding list of contributors (adding information on Dr Robert Gaudin and Dr Akhilanand Chaurasia) * Adding section 3.5.2 Tooth level metrics for segmentation tasks   Version 2 (submitted as FGAI4H-I-010-A01 for e-meeting I in Geneva)   * Update of TDD abstract * Formatting of the document to emphasis sections not yet populated with text * Updates for section 2, AI4H Topic group.   + Expanding list of contributors (adding information on Tarry Singh and deepkapha.ai)   + Defining tools for communication   + Setting up biweekly calls   + Updating the current status   Version 1 (submitted as FGAI4H-H-010-A01 for meeting H in Brasilia)  This document served as initial draft for the topic description document (TDD) of the topic group TG-Dental, which is concerned with the standardized benchmarking of AI for dental diagnostics and digital dentistry.  The focus for the first draft was on the sections 1. Introduction and 3. Methods.   * Introduction   + Relevance   + Impact   + Existing work * Method   + Anatomical structures   + Pathologies   + Data sets and format   + AI Output Data Structure   + Metrics |

**Contributors**

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| --- | --- |
| Falk Schwendicke Charité – Universitätsmedizin Berlin Germany | Email: [falk.schwendicke@charite.de](mailto:falk.schwendicke@charite.de) |
| Joachim Krois Charité – Universitätsmedizin Berlin Germany | Email: [joachim.krois@charite.de](mailto:joachim.krois@charite.de) |
| Tarry Singh deepkapha.ai Netherlands | Email: [tarry.singh@deepkapha.ai](mailto:tarry.singh@deepkapha.ai) |
| Prof. Jae-Hong Lee Wonkwang University College of Dentistry, Daejeon Korea | Email: [ljaehong@gmail.com](mailto:ljaehong@gmail.com) |
| Dr. Akhilanand Chaurasia King George Medical University, Lucknow India | Email: [chaurasiaakhilanand49@gmail.com](mailto:chaurasiaakhilanand49@gmail.com) |
| Dr. Robert André Gaudin  Charité – Universitätsmedizin Berlin Germany | Email: [robert-andre.gaudin@charite.de](mailto:robert-andre.gaudin@charite.de) |
| Sergio Uribe Riga Stradins University, Riga, Latvia Universidad Austral de Chile, Valdivia, Chile | Email: [sergio.uribe@rsu.lv](mailto:sergio.uribe@rsu.lv) |
| Hossein Mohammad-Rahimi Research Assistant, Computer Engineering Department, Sharif University of Technology Iran | Email: [ramtin.rhm@gmail.com](mailto:ramtin.rhm@gmail.com) |
| Janet Brinz Universität Regensburg  Germany | Email: [janet.brinz@stud.uni-regensburg.de](mailto:janet.brinz@stud.uni-regensburg.de) |
| Anahita Haiat  University of Western Australia Australia | Email: [anahita.haiat@gmail.com](mailto:anahita.haiat@gmail.com) |
| Gürkan Ünsal Near East University, Experimental Health Sciences Research Center, Artificial Intelligence in Medicine Research Group Turkey | Email: [gurkanunsal@aol.com](mailto:gurkanunsal@aol.com) |
| Nielsen Santos Pereira Private Sector  Brazil | Email: [nielsen.spereira@gmail.com](mailto:nielsen.spereira@gmail.com) |
| Ulrike Kuchler Associate Professor, Oral Surgery Vienna | Email: [ulrike.kuchler@meduniwien.ac.at](mailto:ulrike.kuchler@meduniwien.ac.at) |
| Shankeeth Vinayahalingam Radboudumc the Netherlands | Email: [shankeeth.vinayahalingam@radboudumc.nl](mailto:shankeeth.vinayahalingam@radboudumc.nl) |
| Balazs Feher Medical University of Vienna | Email: [balazs.feher@meduniwien.ac.at](mailto:balazs.feher@meduniwien.ac.at) |

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**FG-AI4H Topic Description Document**

**Topic group - Dental Diagnostics and Digital Dentistry (TG-Dental)**

# Introduction

The topic group Dental Diagnostics and Digital Dentistry (TG-Dental) focusses on diagnostics in dentistry, dental and oral medicine, and associated disciplines (oral and maxillofacial surgery, orthodontics, dental and oral traumatology). Specifically, it deals with the following (not exhaustive) ICD-10 code headings:

* K00 Disorders of tooth development and eruption
* K01 Embedded and impacted teeth
* K02 Dental caries
* K03 Other diseases of hard tissues of teeth
* K04 Diseases of the pulp and periapical tissues
* K05 Gingivitis and periodontal diseases
* K06 Other disorders of the gingiva and edentulous alveolar ridges
* K07 Dento-facial anomalies including malocclusion
* K08 Other disorders of the teeth and supporting structures
* K09 Cysts of oral origin, not elsewhere classified
* K10 Other diseases of jaws
* K11 Diseases of the salivary glands
* K12 Stomatitis and related lesions
* K13 Other diseases of the lip and oral mucosa
* K14 Diseases of the tongue
* ???? Disorders in the maxillary sinus

Diagnostics include the detection, assessment and prognosis of and on these conditions, as well as diagnostics on patients' risk to experience a condition or its progression. Diagnostics also includes the (image or non-image) assessment of anatomic structures or physiologic functions for the purpose of both detection of conditions, but also treatment planning for various therapies (e.g., implantology).

Dental conditions, like caries, periodontitis or tooth loss, are among the most prevalent diseases of humankind, affecting up to 98% of a population. Direct treatment costs due to dental diseases globally were estimated at US $298 billion annually, corresponding to an average of 4.6% of global health expenditure. The burden emanating from oral diseases is comparable to that from diabetes or cardiovascular diseases. The majority of the world's 1.6 million dentists are based in Europe and the Americas, such that 69% of the world's dentists serve 27% of the global population. Africa has only 1% of the global workforce. The overall workforce in dentistry exceeds 10 million worldwide. Diagnostics in dentistry largely relies on dentists diagnosing diseases via a combination of dental history taking, clinical investigation and imaging as well, if required, further physical, (bio)chemical or microbiological assessments.

AI will help to (1) improve the accuracy of each of these individual tasks, (2) allow the integration of different data with higher effectiveness than the individual can do this, (3) without ease also longitudinally assess these data, compare them over time, and hence allow predictions, (4) reduce the reliance of diagnosis making from the dentist, expanding the scope of dental auxiliary staff, thereby increasing the access and efficiency of dental services, and (5) enable patients and healthy individuals to better participate into their dental health experience and management. AI will pave the way to a more personalized, precise, preventive and participatory dentistry for more people worldwide. It has the potential to aid in overcoming current ineffective, expensive care models.

Expected impact of benchmarking: Benchmarking is expected to yield more robust models and algorithms, with initially lower accuracy compared with current validation strategies (largely in-sample). Benchmarking is further expected to allow transparent comparisons of different models and algorithms.

This topic description document specifies the standardized benchmarking for Dental Diagnostics and Digital Dentistry systems. It serves as deliverable No. 10.17 of the ITU/WHO Focus Group on AI for Health (FG-AI4H).

# About the FG-AI4H topic group on Dental Diagnostics and Digital Dentistry

The introduction highlights the potential of a standardized benchmarking of AI systems for Dental Diagnostics and Digital Dentistry to help solving important health issues and provide decision-makers with the necessary insight to successfully address these challenges.

To develop this benchmarking framework, FG-AI4H decided to create the TG-Dental at the meeting G in New Delhi, 13-15 November 2019.

FG-AI4H assigns a topic driver to each topic group (similar to a moderator) who coordinates the /collaboration of all topic group members on the TDD. During FG-AI4H meeting G in in New Delhi, 13-15 November 2019 Prof. Dr. Falk Schwendicke and Dr. Joachim from Charité – Universitätsmedizin Berlin Germany nominated as topic driver for the TG-Dental.

## Documentation

This document is the TDD for the TG-Dental. It introduces the health topic including the AI task, outlines its relevance and the potential impact that the benchmarking will have on the health system and patient outcome, and provides an overview of the existing AI solutions for Dental Diagnostics and Digital Dentistry. It describes the existing approaches for assessing the quality of Dental Diagnostics and Digital Dentistry systems and provides the details that are likely relevant for setting up a new standardized benchmarking. It specifies the actual benchmarking methods for all subtopics at a level of detail that includes technological and operational implementation. There are individual subsections for all versions of the benchmarking. Finally, it summarizes the results of the topic group's benchmarking initiative and benchmarking runs. In addition, the TDD addresses ethical and regulatory aspects.

The TDD will be developed cooperatively by all members of the topic group over time and updated TDD iterations are expected to be presented at each FG-AI4H meeting.

The final version of this TDD will be released as deliverable "DEL 10.17 Dental Diagnostics and Digital Dentistry (TG-Dental)." The topic group is expected to submit input documents reflecting updates to the work on this deliverable (Table 1) to each FG-AI4H meeting.

Table 1 – Topic group output documents

|  |  |
| --- | --- |
| **Number** | **Title** |
| FGAI4H-M-010-A01 | Latest update of the Topic Description Document of the TG-Dental |
| FGAI4H-M-010-A02 | Latest update of the Call for Topic Group Participation (CfTGP) |
| FGAI4H-M-010-A03 | The presentation summarizing the latest update of the Topic Description Document of the TG-Dental |

The working version of this document can be found in the official topic group SharePoint directory.

* https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/tg/SitePages/TG-Dental.aspx

## Status of this topic group

The following subsections describe the update of the collaboration within the TG-Dental for the official focus group meetings.

* Work on this document
* Work on the benchmarking software
* Progress with data acquisition, annotation, etc.
* Overview of the online meetings including links to meeting minutes
* Relevant insights from interactions with other working groups or topic groups
* Partners joining the topic group
* List of current partners
* Relevant next steps

### Status update for meeting H

After the establishment of the TG-Dental at meeting G the "Call for Topic Group Partizipation Document (CfTGP)" was drafted and the "Topic Description Document (TDD)" was updated with a focus on the sections 1. Introduction and 3. Methods.

* Introduction
  + Relevance
  + Impact
  + Existing work
* Method
  + Anatomical structures
  + Pathologies
  + Data sets and format
  + AI Output Data Structure
  + Metrics

Further, other academic groups in dentistry that focus on machine learning and deep learning but as well on ethics, public health and economics were contacted.

### Status update for meeting I

The TDD abstract was updated and the TDD was reviewed to emphasize sections not yet populated with text.

In addition, the list of contributors was expanded and Tarry Singh, from deepkapha.ai, Netherlands, as well as Prof Jae-Hong Lee from Wonkwang University, Korea, joined the TG-Dental. The CfTGP was updated accordingly. The contributors of the TG-Dental defined tools for communication and set up bi-weekly calls via the zoom chat room of FGAI4H (<https://itu.zoom.us/my/fgai4h>).

Prof Schwendicke initiated consultations with key opinion leaders from SPIRIT/CONSORT-AI in order to prepare the dissemination in key dental journals. Further, the contributors of the TG-Dental started the drafting of a series of scientific articles on dental AI methods, opportunities and challenges in leading dental journals.

### Status update for meeting J

Two new contributors joined the TG-Dental, Dr. Robert Gaudin from Charite – Universitätsmedizin Berlin, Germany and Dr. Akhilanand Chaurasia from King George Medical University, Lucknow, India.

The TG-Dental contributors drafted a manuscript on *Artificial intelligence for dental image analysis: A Guide for Authors and Reviewers and* shared it for discussion with FGAI4H members.

The TDD was updated in particular on section 4.5, Metrics, where the concept of computing the tooth-based confusion matrix for dental image segmentation tasks was described. This allows the calculation of tooth level metrics for segmentation tasks, which in turn is useful to compare computer vision metrics to clinical data that is widely accessible in clinical studies and trials.

### Status update for meeting K

For meeting K the TDD structure was adapted as defined in document FGAI4H-J-105.

The contributors of TG-Dental in addition with Thomas Wiegand (Chair of the ITU/WHO FGAI4H) and Sergio Uribe (President of the e-Oral Health Network of the IADR) drafted the manuscript *Artificial intelligence in dental research: A Checklist for Authors and Reviewers.* This manuscript highlights the challenge that the number of studies employing artificial intelligence in dentistry is growing fast but that the majority of studies suffer from limitations in planning, conduct and reporting, resulting in low robustness and applicability. The manuscript provides a consented checklist for authors, reviewers and readers of AI studies in dental research. The initial draft of the checklist and an explanatory document were derived and discussed among the members of IADR's e-oral network and the ITU/WHO focus group "Artificial Intelligence for Health (AI4H)". The checklist was consented by 27 group members via an e-Delphi process.

In addition, members of TG-Dental contributed to the systematic review and meta-analysis of deep learning for cephalometric landmark detection. From 318 identified records, 19 studies (published 2017-2020), all employing convolutional neural networks, mainly on 2-D-lateral cephalometric radiographs (n=15), using data from publicly available datasets (n=12) and testing the detection of a mean of 30 (SD: 25; median: 19; min.-max.: 7-93) landmarks were included. The reference test was established by two (n=11), 1 (n=4), 3 (n=3) or a set of experts (n=1). Risk of bias was high and applicability concerns detected for most studies. The mean (95% confidence interval) deviation of predicted from reference landmarks was 2.1 (95% CI: 1.2-3.0) mm, while the mean % of landmarks detected with 0.05). The analysis concluded that DL shows relatively high accuracy for detecting landmarks on cephalometric imagery; however, only few studies directly compared DL with clinicians or sufficiently reflected on the dimension of deviation (vertical versus sagittal). The overall body of evidence is consistent, but of limited robustness. The insights of this study will be incorporated into the TDD and inform the state-of-the-art of AI-based cephalometric landmark detection.

### Status update for meeting L

Three new contributors joined the TG-Dental, Janet Brinz, final year dentistry student at Universität Regensburg, Germany; Sergio Uribe, PhD, DDS, specialist in Oral & Maxillofacial Radiology, leading researcher, Bioinformatics Research Unit Riga Stradins University, Riga, Latvia and Associate Professor Universidad Austral de Chile, Valdivia, Chile; and Hossein Mohammad-Rahimi a General Dentist (DDS) graduated from SBMU, Tehran, Iran, Research Assistant, Computer Engineering Department, Sharif University of Technology, Tehran, Iran and

founder of "DeepBites" research group, which applies AI approaches to dentistry.

The topic group members initiated a discussion on the structure of AI in dentistry and developed a figure that thematically structures AI subtopics along the clinical care pathway and along subfields of dentistry (see Figure 1).

In addition, a systematic review of AI applications in dentistry was initialized resulting in over eligible 190 studies, with nearly all of them focusing on either image analysis or shallow machine learning. These studies showed a high risk of bias. Various accuracy estimates and more than 25 different use cases were found. A preliminary conclusion indicates limited consistency and comparability of the studies but reporting quality increased over time. Notably, outcomes beyond accuracy score are scarce and robustness and generalizability remains unclear.

### Status update for meeting M

Seven new contributors joined the TG-Dental. Anahita Haiat, a final year dental student- University of Western Australia, Perth, Australia holding a Master of Engineering and a Bachelor of Mathematics. Jaisri Thoppay, President of the Center for Integrative Oral Health inc. Winter Park, Florida, USA. Nielsen Pereira, a dentist and entrepreneur. Gürkan Ünsal, a dentomaxillofacial radiologist at Near East University, Faculty of Dentistry and member of the Artificial Intelligence in Medicine Research Group at Near East University, DESAM Institute. Ulrike Kuchler, an Associate Professor at the Department of Oral Surgery, Medical University of Vienna, Vienna, Austria. Balazs Feher a junior scientist and lecturer in oral surgery at Medical University of Vienna, Vienna, Austria. Shankeeth Vinayahalingam a PhD student in Oral & Maxillofacial Surgery at Radboudumc Nijmegen, the Netherlands and a Master AI student at Radboud University, Nijmegen, the Netherlands. Figure 1 was revised and updated according to new members joining the group. These new members were assigned to different subdomains according to their expertise and interest. In addition an "infrastructure"-column was added. In this column members are referenced that are in particular interested in technical aspects of the collaboration.

Section 3 and in particular the subtopic descriptions have been significantly advanced.

The TG-Dental successfully published the FGAI4H output document "Artificial intelligence in dental research: A checklist for authors and reviewers (FGAI4H-M-004-A01)". The checklist for dental AI studies was produced as a collaboration of experts from the International Association for Dental Research (IADR), the E-oral Health Network and the ITU/WHO Focus Group on AI for Health. Given that the number of studies employing artificial intelligence (AI), specifically machine and deep learning, is growing fast and that the majority of studies suffer from limitations in planning, conduct and reporting, resulting in low robustness, reproducibility and applicability, this document provides a consented checklist on planning, conducting and reporting of AI studies for authors, reviewers and readers in dental research.

The TG-Dental further conceptualized their first multi-center benchmark study. The goal is to include approximately 10 centers/clinics/practices from all over the world. It is expected that differences in cohort characteristics (age and gender), the dental status (e.g. DMFT) and technical conditions (x-ray machine, image characteristics, etc), among others will be present. In a collaborative effort different DL models will be trained on the data of different centers (or different subsets, etc) and predict on hold-out test sets from the center the model was trained on but as well on centers representing other populations. We would expect differences in those results. We will evaluate the statistical significance of those differences but as well implement inference models and methods of model explainability to evaluate which of the metadata (cohort, clinical and technical) can be associated with those differences.

In addition, this benchmark study was accepted as one more use case for the FG-AI4H Trial Audits 2.0 project. This project is ongoing and for the team formation phase very good progress was reported. In a next phase the "Audit Methodology Design" will be further developed (see Section 3.B under [Trial Audit 2.0 Playbook](https://docs.google.com/document/d/1YPMy2oFc8vr5mGlQ_zVft2Mm31fiRdgHr25kEc2g37Y/edit?usp=sharing)).

## Topic group participation

The participation in both, the Focus Group on AI for Health and in a TG is generally open to anyone (with a free ITU account). For this TG, the corresponding 'Call for TG participation' (CfTGP) can be found here:

* https://www.itu.int/en/ITU-T/focusgroups/ai4h/Documents/tg/CfP-TG-Dental.pdf

Each topic group also has a corresponding subpage on the ITU collaboration site. The subpage for this topic group can be found here:

* https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/tg/SitePages/TG-Dental.aspx

For participation in this topic group, interested parties can also join the regular online meetings. The meeting is scheduled for every 2nd Thursday at 08:15 CET. For all TGs, the link will be the standard ITU-TG 'zoom' link:

* <https://itu.zoom.us/my/fgai4h>

All relevant administrative information about FG-AI4H—like upcoming meetings or document deadlines—will be announced via the general FG-AI4H mailing list [fgai4h@lists.itu.int](mailto:fgai4h@lists.itu.int).

All TG members should subscribe to this mailing list as part of the registration process for their ITU user account by following the instructions in the 'Call for Topic Group participation' and this link:

* <https://itu.int/go/fgai4h/join>

Regular FG-AI4H workshops and meetings proceed about every two months at changing locations around the globe or remotely. More information can be found on the official FG-AI4H website:

* <https://itu.int/go/fgai4h>

# Topic description

This section contains a detailed description and background information of the specific health topic for the benchmarking of AI in Dental Diagnostics and Digital Dentistry and how this can help to solve a relevant 'real-world' problem.

Topic groups summarize related benchmarking AI subjects to reduce redundancy, leverage synergies, and streamline FG-AI4H meetings. However, in some cases different subtopic groups can be established within one topic group to pursue different topic-specific fields of expertise. The TG-Dental currently has no subtopics. Future subtopics for [SUBTOPIC NAME] might be introduced.

*Topic Driver: Topic groups typically begin* ***without*** *subtopics. Please write a few lines indicating future subtopics that might become relevant. Once you have defined subtopics, their focus/mandate should be explained in this section.*

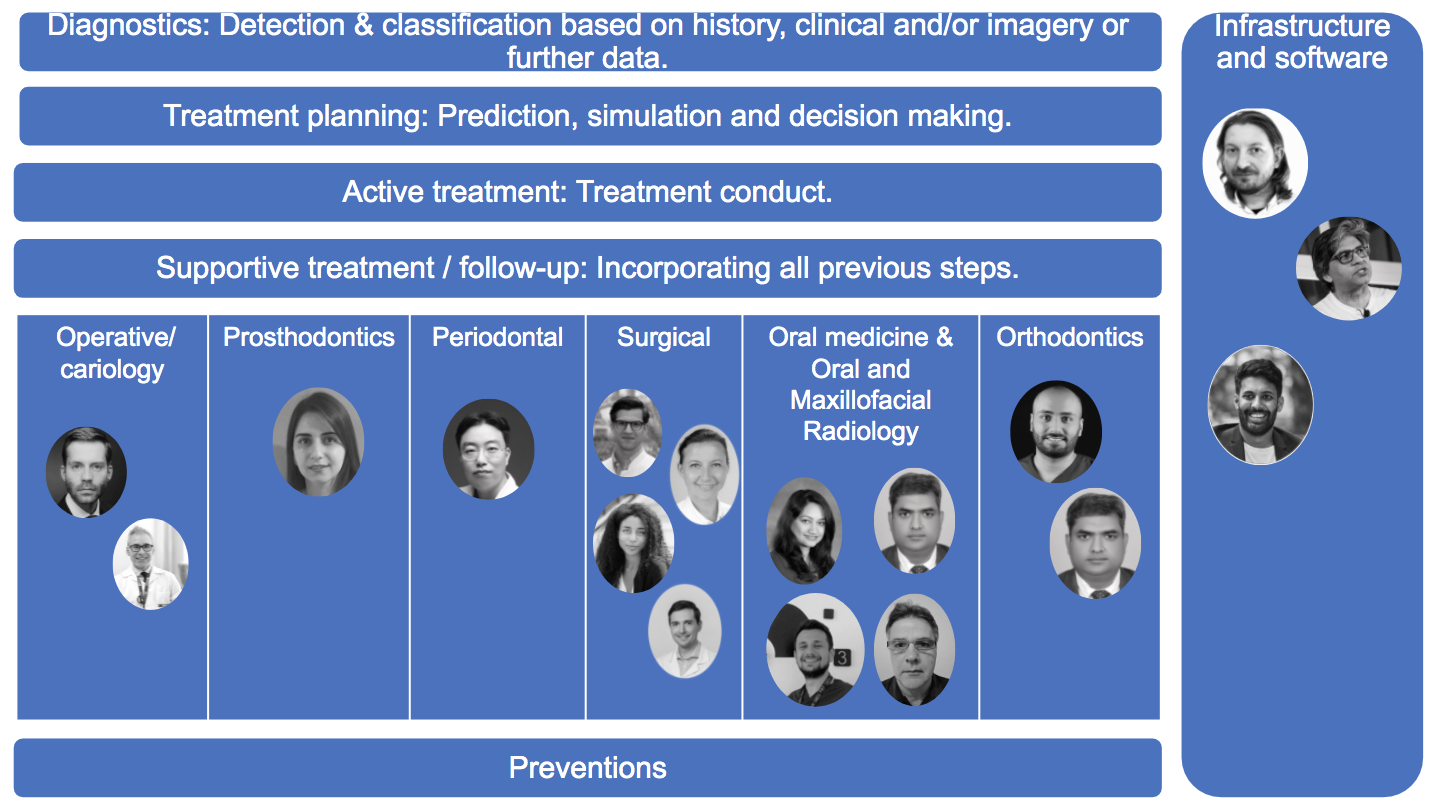


Figure 1 – Thematic structure of subtopics along the clinical care pathway and along subfields of dentistry

## Subtopic - TEMPLATE [A]

### Definition of the AI task

This section provides a detailed description of the specific task the AI systems of this TG are expected to solve. It is *not* about the benchmarking process (this will be discussed more detailed in chapter 4). This section corresponds to [DEL03](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B7997F2C1-5A1D-4409-B2A0-CBC4E9CE8CDA%7D&file=DEL03.docx&action=default) *"AI requirements specifications*," which describes the functional, behavioural, and operational aspects of an AI system.

* What is the AI doing?
* What kind of AI task is implemented (e.g., classification, prediction, clustering, or segmentation task)?
* Which input data are fed into the AI model?
* Which output is generated?

### Current gold standard

This section provides a description of the established gold standard of the addressed health topic.

* How is the task currently solved without AI?
* Do any issues occur with the current gold standard? Does it have limitations?
* Are there any numbers describing the performance of the current state of the art?

### Relevance and impact of an AI solution

This section addresses the relevance and impact of the AI solution (e.g., on the health system or the patient outcome) and describes how solving the task with AI improves a health issue.

* Why is solving the addressed task with AI relevant?
* Which impact of deploying such systems is expected (e.g., impact on the health system, overall health system cost, life expectancy, or gross domestic product)?
* Why is benchmarking for this topic important (e.g., does it provide stakeholders with numbers for decision-making; does it simplify regulation, build trust, or facilitate adoption)?

### Existing AI solutions

This section provides an overview of existing AI solutions for the same health topic that are already in operation. It should contain details of the operations, limitations, robustness, and the scope of the available AI solutions. The details on performance and existing benchmarking procedures will be covered in chapter 6.

* Description of the general status and the maturity of AI systems for the health topic of your TG (e.g., exclusively prototypes, applications, and validated medical devices)
* Which are the currently known AI systems and their inputs, outputs, key features, target user groups, and intended use (if not discussed before)? This can also be provided as a table.
* What are the common features found in most AI solutions that might be benchmarked?
* What are the relevant metadata dimensions characterizing the AI systems in this field and with relevance for reporting (e.g., systems supporting offline functions, availability in certain languages, and the capability to process data in a specific format)?
* Description of existing AI systems and their scope, robustness, and other dimensions.

## Operative and Cariology (PI: [falk.schwendicke@charite.de](mailto:falk.schwendicke@charite.de), [sergio.uribe@gmail.com](mailto:sergio.uribe@gmail.com))

### Diagnostic and treatment pathways

Dental caries is the most prevalent and ubiquitous non-communicable disease affecting humankind today. It was first understood to be an infectious disease, requiring removal of all plaque (biofilm) from the teeth or from affected carious hard tissues (specific plaque hypothesis). This concept, while always debated (i.e. each biofilm being cariogenic under certain conditions; non-specific plaque hypothesis), was later modified, suggesting that the mere presence of biofilm is not sufficient for the pathogenesis of caries, but that an overlapping interaction between the host/teeth, substrate and microbiota is needed. Despite being 'infected' (or rather, contaminated) with cariogenic bacteria, a cavitated carious lesion will not develop without a cariogenic diet. Marsh introduced the ecological plaque theory. The microbial composition of the biofilm is stable unless 'environmental perturbations' occur which can affect microbial homeostasis, leading to dysbiosis. With respect to caries, diet (mainly free sugars), oral hygiene and salivary factors are the contributing drivers of dysbiosis, leading to a shift in the microbiota towards acidogenic and aciduric microorganisms. The initial (naïve) dental biofilm is influenced by both hereditary and environmental factors, but, as time goes on, the types and proportions of micro-organisms acquired early in life are modified by environmental influences. Currently, the extended ecological plaque hypothesis is accepted to explain the pathogenesis of caries.

This pathogenesis of dental caries involves organic acids, the by-product of microbial metabolism of dietary free sugars. As the pH of the biofilm decreases, it reaches a point where the biofilm fluid at the surface of the tooth is under-saturated with respect to tooth mineral and dissolution occurs to maintain equilibrium. Initially, the dissolution occurs at the surface of the tooth, but, if conditions persist, and the lesion becomes more extensive, minerals from deeper in the enamel (and subsequently dentine) will be lost. While the pathogenesis of dental caries is mainly explained by environmental factors, dental caries has a genetic component, with influencing factors including enamel quality and quantity, immune response, dietary preferences and salivary characteristics. In summary, dental caries is a disease characterized by a process of demineralization of the dental hard tissues, caused by frequent free sugars exposure to the dental biofilm, which shifts the ecological balance towards a cariogenic dysbiosis. For dentin and root caries, cleavage of collagen by bacterial or dentinal enzymes follows early mineral loss and contributes to the loss of the hard tissue.

The former, "traditional" management of the caries process and carious lesions, was influenced by an understanding that caries was a purely infectious disease and could be managed invasively/restoratively by removing all demineralized and 'contaminated' tissue. This was grounded in (1) A lack of understanding that the caries process and carious lesions are separate, but related; (2) the incorrect understanding that once a lesion had established and the tooth was "infected", eradication of microbiota was needed; (3) the erroneous concept that lesion progression was inevitable; and, (4) the fact that the majority of carious lesions dentists encountered in the past were truly "decayed", i.e. cavitated dentine lesions. Therefore, the professional education of dental *surgeons* concentrated on mechanistic surgical procedures instead of that of dental *physicians* who manage the disease, and remuneration systems incentivized such invasive/restorative therapies.

Nowadays, and building on evidence accrued over several decades, it is clear that (1) The caries process can be controlled by modifying the patient's caries risk/susceptibility, depending on his/her adherence to behavioural modifications and not only by intervening operatively on carious lesions, yet success/understanding of behavioural interventions on caries control has been limited; (2) The caries process and carious lesions can be managed without removing microorganisms, but by rebalancing the dysbiosis within the tooth surface biofilm and arresting those within the depths of the tissues; (3) Active (progressing) lesions can be inactivated, and progression is usually slow for most lesions; (4) In many high-income countries the spectrum of carious lesions has been and is shifting, especially in younger people, as there are now more non-cavitated lesions being detected.

Hence, the conventional restorative/invasive approach towards managing the caries process and carious lesions is not grounded in current understanding of the disease and it is also not appropriate for managing the broad spectrum of lesions found in many individuals (from very early to large cavitated). It should also be considered that subsequent interventions on restorations are often necessary. This is classically known as the "restorative death spiral". Given these alternative arguments, there is consensus that invasive/restorative interventions alone are not beneficial for managing the caries process and lesions in all situations. Instead, invasive/restorative interventions represent a late stage in the management puzzle, repairing the gross tissue damage and restoring form, function, aesthetics and cleansability, thereby allowing to control the risk of future loss of function. Invasive strategies may also be used to approach acute caries lesions. Invasive/restorative interventions are an important and relevant tool, but they should be complemented by other (non- or micro-invasive) management strategies.

We distinguish three levels of invasiveness to classify intervention strategies for "treating" existing carious lesions. They are based on the degree of tissue removal associated with each strategy: *Non-invasive* strategies do not remove dental hard tissue and involve, for example, fluorides and other chemical strategies for controlling mineral balance, biofilm control measures and dietary control. *Micro-invasive* strategies remove the dental hard tissue surface at the micrometre level, usually during an etching step, such as sealing or infiltration techniques. *Invasive* strategies remove gross dental hard tissue, such as through use of hand excavators, rotary instruments or other devices. In most cases, this process is associated with the placement of restorations.

For each pathway of management, regardless if it is non-, micro- or invasive, caries detection is needed. Moreover, agreed principles for when to use different intervention strategies have been outlined: Lesion activity, cavitation and cleansability are the main factors to be considered to determine intervention thresholds. Inactive lesions do not usually require any treatment (in some cases, restorations might be placed for reasons of form, function, aesthetics); active lesions do. Non-cavitated carious lesions should be managed non- or micro-invasively, as should cavitated carious lesions which are cleansable. Cavitated carious lesions which are not cleansable usually require invasive/restorative management, also to restore form, function and aesthetics of the tooth. In specific circumstances, mixed interventions may be applicable. On occlusal surfaces, cavitated lesions confined to enamel and non-cavitated lesions radiographically extending deep into dentine (middle or inner dentine third, D2/3) may be exceptions to that rule. On proximal surfaces, cavitation is usually hard to assess tactile-visually. Hence, radiographic lesion depth is used to determine the likelihood of cavitation. Lesions extending radiographically into the middle or inner third of the dentine (D2/3) can be assumed to be cavitated, while those restricted to the enamel (external or middle enamel third, E1/2) are usually not cavitated. For lesions radiographically extending into the outer third of the dentine (D1), cavitation status is unclear. These lesions should be managed as if they were non-cavitated unless otherwise indicated. Individual decisions and clinical judgment should consider factors modifying the described intervention thresholds. Comprehensive diagnostics are the basis for systematic decision-making on when to intervene in the caries process and on existing carious lesions. Patients should be comprehensively informed about treatment options and should provide informed consent accordingly.

### Definition of the AI task

A wide range of AI tasks along these pathways of risk assessment, detection, diagnostics, treatment planning, conduct and re-evaluation are conceivable. Currently, these entail

* Caries risk assessment
* Caries detection on
* photos or videos
* bitewing radiographs
* peri-apical radiographs
* panoramic radiographs
* Cone Beam CT (CBCT)
* rare imagery like OCT, NILT
* Caries lesion depth assessment and progression prediction
* Treatment decision-support (decision between management levels and options)

### Current gold standard

In vitro, the gold standard is established using destroying preparation and analytic techniques like histology, micro-radiography or (non-destructively) micro-CT. For clinical gathered data, this is not available, and clinical assessment or, for imagery, expert opinion are needed instead, if possible triangulation is used. For example, routine data relies on one expert in the clinical setting assessing the tooth using criteria like those one outlined elsewhere ([https://pubmed.ncbi.nlm.nih.gov/‌17518963/](https://pubmed.ncbi.nlm.nih.gov/17518963/)).

If only imagery is available, one or a range of experts annotate each image independently, providing an instance label (e.g. caries present), a detection label (e.g. a bounding box around an area) or a segment label (e.g. a pixel-wise area) of interest. Unifying such fuzzy labels is now a matter of choice: (1) For instance-based annotations, majority voting schemes are often applied (e.g. 3 or more out of 5 experts need to agree that a pathology is present, otherwise it is assumed not to be present; in some cases, images where the uncertainty is especially high are discarded). (2) It is also possible to combine these annotations with a "master" review process where one or more master experts are reviewing the labels, created by a range of other human annotators, in order to correct them. Hence the annotations can be unionized based on a state of higher information (as all annotations can be seen by the master reviewer) and on the underlying assumption that the master's experience leads to high-quality final labels. (3) Also, it is possible to discuss uncertain cases by an expert board, which would not be able to jointly assess all images from a dataset given time and resource constraints but can assess the small minority of unclear ("edge") cases to come to a consensus there.

### Relevance and impact of an AI solution

Caries risk assessment and detection suffer from poor accuracy, grounded in either a limited set of predictor variables with poor predictive performance being available, existing prediction models being insufficiently calibrated to individual settings, populations or people, or experts' limited diagnostics accuracy in routine settings further being modified by their experience. AI support along these tasks promises to increase the accuracy of assessment, prediction and detection/localization or to at least achieve an expert-level status for a range of tasks. It may further reduce the effort for doing so (efficiency gain), increase safety and contain costs. Notably, as AI will, for a long time, remain a support tool, the final impact (benefit/harm) will largely depend on the user decision after employing AI support.

### Existing AI solutions

In research, a recent review (unpublished) identified 48 studies on caries detection using AI in PubMed. AI for caries lesion localization or progression prediction have not been found, the same applies to decision support. The detection studies were mainly employing radiographs; the minority used photographs or other imagery.

Commercially, to our knowledge only few applications are available for caries detection. The only application which has undergone confirmatory assessment (CE/FDA) is dentalXrai Pro (dentalXrai GmbH, Berlin), which detects caries on panoramic images and bitewings. Other manufacturers (dentiAI, Toronto, Pearl AI, California, Videahealth, Boston) claim to perform caries detection (on panoramic images - dentiAI, bitewings - Pearl, periapical images - Videahealth), but have so far not received full-scale regulatory approval. No solutions for other use cases are known to us at this point.

## Prosthodontics ([anahita.haiat@gmail.com](mailto:anahita.haiat@gmail.com))

Prosthodontics involves the diagnosis and treatment of the oral function, appearance and health of patients with clinical conditions related to missing or deficient teeth or maxillofacial tissue. The scope of prosthodontic treatment can range from a single indirect restoration to full mouth rehabilitation using fixed or removable, tooth or implant-supported prosthesis, and guidelines such as the Prosthodontic Diagnostic Index (PDI) have been established to assess the complexity of clinical cases and aid in formulating appropriate treatment plans.

### Definition of the AI task

Numerous AI tasks are envisaged to enhance prosthodontic diagnostics, treatment planning and conduct. These comprise the following:

* Decision support to determine suitable treatment plans, for example to determine suitable types of prosthesis and restorations by integrating all the patient-related factors such as 3D assessment of mandibular movements, occlusal loading, and periodontal and endodontic status of existing dentition.
* Integration of AI into the various stages of existing digital workflows such as taking impressions, restoration design and fabrication to enhance accuracy
* Assistance robotic arms to aid the dentist with accurate execution of preparation designed by an AI software.

### Current gold standard

At present, planning for prosthodontic treatment draws on the clinical judgment and experience of the clinician, however agreement on the majority of cases is lacking. The fabrication of most types of prosthesis is now done via the CAD/CAM digital workflow. Preparation of teeth is done by dentists, however even after years of training and clinical practice, this is still a very challenging and error-prone task and imprecision can compromise the final outcome.

### Relevance and impact of an AI solution

Errors during diagnosis, treatment planning and execution of treatment can lead to poor outcomes such as occlusal imbalances, damage to the stomatognathic system, early fracture or failure of restorations, and suboptimal aesthetics. By implementing AI through the various stages of prosthodontic treatment, it is possible to significantly improve predictability of outcomes.

### Existing AI solutions

A recent literature search identified the following applications of AI in prosthodontic research:

1. Prosthesis shade selection:
2. Shade selection for metal-ceramic crowns
3. Coloration of maxillofacial prosthesis
4. Automation of prosthesis design:
   1. Implant-supported monolithic zirconia crowns
   2. Removable partial dentures
   3. Identification of tooth preparation finishing line
5. Assistance in tooth preparation:
   1. Tooth preparation using laser ablation robotic arms
6. Treatment planning support:
   1. Decision support for extraction of compromised teeth
7. Prediction of treatment outcome and prognosis of prosthesis:
   1. Prediction of debonding of CAD-CAM composite resin crowns

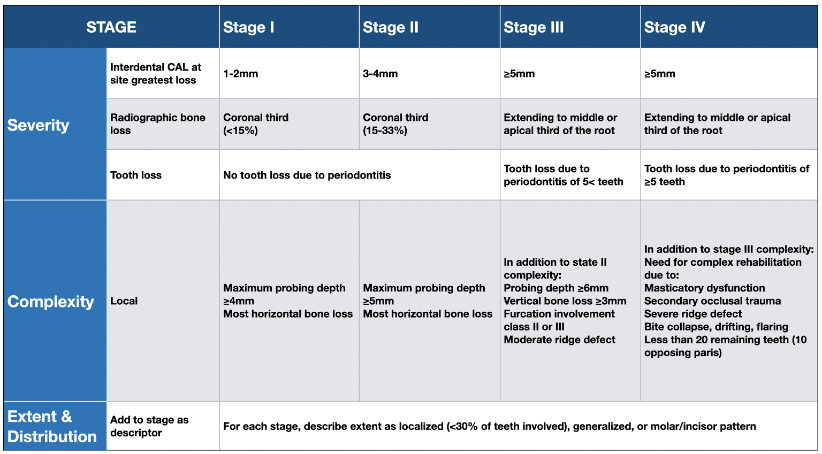
To the best of our knowledge, these applications have not yet undergone regulatory approval and are not available for commercial use.

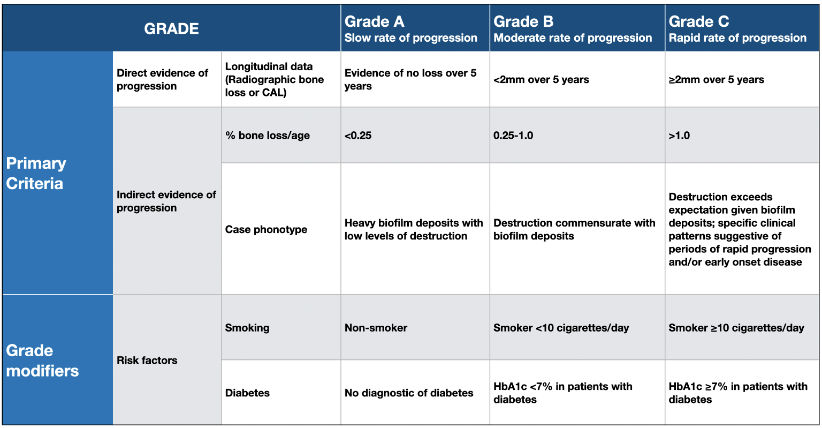
## Periodontal ([ljaehong@gmail.com](mailto:ljaehong@gmail.com))

### Diagnostic and treatment pathways

For periodontal disease, especially periodontitis, the severity, complexity of management, extent, rate of progression, and risk factors are classified based on staging (I–IV) and grading (A-C) systems. Due to the highly heterogeneous clinical manifestations and inheritance patterns, dental professionals tend to conduct several diagnostic analyses based on both stage and grade to determine the accurate diagnosis. Several risk factors that may influence the periodontal condition, such as smoking, diabetes mellitus, obesity, specific genetic factors, physical activity, nutrition, and stress, are also systematically and comprehensively examined for a more detailed analysis.

A broad range of periodontitis-related treatment modalities is available, including chemotherapy, resective treatment, periodontal regenerative treatment, periodontal plastic surgery for gingival augmentation, occlusal treatment, preprosthetic periodontal treatment, and extraction and implant surgery. There is no single effective treatment approach for periodontitis. One treatment procedure may be appropriate and cost-effective for one site of the mouth, while another approach may be suitable for other sites.





### Definition of the AI task

AI systems play a key role in the management of periodontal diseases. AI systems provide a variety of diagnostic and detection methods for periodontal diseases, including dental plaque, calculus, gingivitis, and periodontitis. The use of unsupervised or supervised AI systems will aid in the accurate diagnosis and prediction of periodontal diseases. Periodontal disease-related AI can be applied in the following areas: 1) dental plaque and calculus detection, 2) gingivitis detection, and 3) periodontitis detection.

AI can serve as a second opinion. AI interpretations are based on the experience of professional radiologists and periodontists who participated in setting the AI algorithms. AI training relies on deep learning, especially convolutional neural network algorithms based on thousands of dental periapical and panoramic radiographic images, and continuous training occurs as more data becomes available. Moreover, the classification task is mainly used for periapical images, and the segmentation task is used for panoramic images. Automatic detection of ROIs for pathologies such as alveolar bone loss is also used to provide a detailed diagnosis of the periodontal condition.

### Current gold standard

Clinical diagnostic analyses are generally based on the signs and symptoms of gingival inflammation and periodontal tissue destruction. Common diagnostic tools are used, including clinical and dental radiographic examinations. Typically, the presence of dental plaque, calculus, clinical attachment loss, tooth mobility, bleeding, and pus discharge is detected by clinicians using a periodontal probe, and sometimes, with the aid of panoramic and periapical radiographic images. However, these assessment methods are not only inconvenient but also unreliable and inconsistent. Although several attempts have been made to address or reduce their disadvantages, they still do not provide substantial advantages over existing conventional diagnostic methods, in terms of cost, time, and standardization.

### Relevance and impact of an AI solution

Periodontal disease is a major cause of tooth loss in adults and is one of the most important oral diseases contributing to the global chronic disease burden. More than half of the world's population has periodontal disease. Periodontal disease is the sixth most prevalent inflammatory disease. Globally, 11% of adults have severe and advanced periodontitis.

Periodontal disease is closely related to various sociodemographic factors (including age, sex, household income, insurance status, and residence area) and comorbidities (including diabetes mellitus, obesity, cardiovascular disease, rheumatoid arthritis, metabolic syndrome, and adverse pregnancy outcomes). Therefore, early detection and accurate diagnosis are very important, not only for personal health but also for reducing the social burden.

AI is particularly useful for the standardized diagnosis of periodontal disease, which is relevant to board-certified periodontists and general dentists. In many cases, periodontal disease can be underdiagnosed due to a lack of experience, skill, time, and cost. A more rapid, accurate, and reliable periodontal diagnosis based on AI algorithms will enable early intervention and enhanced prevention and treatment outcomes. In particular, due to the visualizations of the AI second opinion, the acceptance of earlier detection and treatment or additional examinations will be improved among periodontally compromised patients.

### Existing AI solutions

Currently, the direct use of most AI systems in clinical practice is limited. Most systems are still being used for research and investigational purposes only. For example, there are AI systems, e.g., Denti.AI, that interpret periodontal disease being serviced online but these have not yet been used in clinical practice. Recently, several companies and related research institutes have begun developing automated diagnostics for dental radiographs based on AI. Hence, it is expected that AI products will be available for use in clinical practice soon.

The AI tools or solutions available for the diagnosis of periodontal diseases are as follows:

* Diagnocat: Diagnocat is an AI solution that uses CBCT scans, full mouth series, or panoramic and periapical radiographic images. Diagnocat assists dental professionals in the diagnosis of various oral conditions: periodontal, conservative, endodontic, prosthetic, implant, and surgical. The limitation of this tool is that it mainly focuses on three-dimensional radiographic image analyses.
* OralCam: OralCam is a smartphone application that helps users examine common oral conditions (including dental caries, periodontal disease, and dental calculus). Although it has the advantage of simple self-diagnosis, its accuracy is too low for use in clinical practice. Moreover, diagnosis is mainly confined to the labial surfaces of teeth.

## Surgical ([robert-andre.gaudin@charite.de](mailto:robert-andre.gaudin@charite.de) and [brinz.janet@gmail.com](mailto:brinz.janet@gmail.com))

Extensive treatment planning and preparation is essential for the successful treatment of most complex surgeries in complex patients. At present, much planning is still done by hand, but digital, semi-automatic planning is becoming increasingly common in dental surgery, especially in implantology. Even in the largest surgical segment of oral and maxillofacial surgery, automated planning is indispensable to simplify the surgeon's life. For example bony reconstruction of the face after resection of extensive tumors is a particular challenge in facial surgery. There are already commercially available software tools for virtual planning of these procedures. However, these are not automated and so complicated to use that they are rarely used directly by surgeons and require cumbersome and lengthy communication with service providers. AI in this field has the potential to revolutionize and simplify these complicated processes. It would reduce the margin for error as it goes through fewer hands in the planning process. Due to the limitations in the length of this document, we will only focus on the examples mentioned above, automated implant planning and automated bone reconstruction.

The following data is required before designing a comprehensive treatment guide:

1. Personal details
2. Diagnosis
3. Para-clinical assessment
4. Facial/Intraoral Photographs
5. OPG Radiographs
6. CBCT or CT scan of the head
7. Intraoral scan

### Current diagnostic and treatment pathways

Implantology: Detailed diagnostics and precise surgical planning with a drilling template are the most important steps to perform a proper implantation.

In addition to clinical examinations (assessment of remaining teeth, height and width of the jaw, mucosal assessment ), radiographic imaging in the form of a Panoramic radiograph or a CBCT is crucial to provide information about the actual bone volume. Based on this information right now in the most advanced practice the surgeon can determine the length, width and the angle of the implant through an implant planning software. With this information the dental laboratory will 3D print a drilling template. Furthermore the surgeon can insert the implant with the conventional steps of the mucosal incision and drilling.

Reconstruction of bony defects: If the defect of the lower jaw, for example, is too large, it must be reconstructed with bone. For this purpose, the fibula graft for example is used in oral and maxillofacial surgery. The reconstruction is calculated and planned by a service provider on the basis of the size of the defect as well as the 3D image dataset of the jaw and the desired fibula. Finally, the service provider sends a sawing template to the surgeon, who can then use it to shape the bone with millimeter precision. This is a very reliable procedure with good results

### Definition of the AI task

The goal would be to use artificial intelligence to automatically process the given image data and generate a surgical proposal. The virtual planning and the 3D-printing of the surgical template could happen all in house.

Prediction models could be used to approximate implant success rates whilst taking into consideration bone structure, bone density and implant system therefore aiding pre-surgical decision making.

### Existing AI solutions

The Application of AI Models in planning of 3D printed implant templates and Cad Cam cutting guides is currently still limited to research and needs further development and evaluation before clinical utilization. Tasks include predicting success and designing patient specific implants/templates and optimizing the fit, length and diameter. The accuracy of the model is dependent on the task.

## Oral Medicine and Maxillofacial Radiology ([chaurasiaakhilanand49@gmail.com](mailto:chaurasiaakhilanand49@gmail.com))

Detection and diagnosis of oral lesions is of crucial importance in dental practices because early detection significantly improves prognosis. Diagnosis and classification of bone diseases (cysts, tumours). Till date the majority of oral diagnostics are based on manual workouts and personalized perceptions. AI has potential to revolutionize the diagnostic accuracy and prediction of disease course, outcomes and prognosis.

1. Diagnosis and classification of systemic diseases based on oral manifestations
2. Diagnosis and classification of periodontal disease
3. Use of immunologic parameters to diagnose aggressive periodontitis.
4. Classification of halitosis based on identification of periodontal pathogens in saliva.
5. Predicting recurrence of aphthous ulcers.
6. Diagnosis and classification of Head and Neck Syndromes

### Temporomandibular joint disorders (TMD)

1. Use of magnetic resonance imaging to determine the prognosis of TMD.
2. Use of screening questions to aid in treatment of TMD.
3. Identification of subgroups of internal derangement of the temporomandibular joint.
4. Classification of TMD on the basis of clinical sign and symptoms
5. Automated TMD diagnostics

### Oral Cancer

1. Diagnosis and classification of Oral precancers and cancer
2. Prognosis of oral cancer considering lesion histology and genetic characteristics.
3. Assessment of hyper-nasal speech following treatment of oral/or pharyngeal cancer.
4. Risk assessment for oral cancer.(Prediction)
5. Prediction of Cervical lymph node metastasis of oral squamous cell carcinoma

### Masticatory Muscle diseases

1. AI based Classification of masticator muscle disorders
2. Prediction of masticator muscle disorders on the basis of chewing pattern

### AI Based Differential Diagnostic Systems for oral diseases based on Clinical and radiographic features

### Oral and Maxillofacial Radiology

Artificial intelligence is being studied in relation to various topics in the field of maxillofacial radiology. Radiographic imaging is very useful for diagnostic purposes and to ensure proper treatment. Maxillofacial radiologists as professionals understand the basic principles and characteristics of radiographic imaging and interpret them in terms of various diseases and continue to play an important role in artificial intelligence-related research.

Radiographs of Interests for AI based Diagnostics in oral and maxillofacial radiology are-

**a. Conventional Radiographs**

* Periapical
* Bitewings
* Panoramic Radiograph
* Lateral Cephalogram
* Hand and Wrist Radiograph

**b. Advanced Imaging**

* Plane CBCT, 3D CBCT scans
* CT Scan
* MR sans

### Radiographic diagnosis

In the field of OMF radiology, artificial intelligence can be used for the diagnosis of a broad range of oral diseases, dental caries, periodontal disease, osteosclerosis, odontogenic cysts and tumours, diseases of the maxillary sinus or temporomandibular joints**,** vertical root fractures, periapical pathosis and cysts and tumours of jaw bone. Detection of the filling or filled teeth (for prediction of the index DMFT - Decayed, Missing and Filled Permanent Teeth)

**Interpretation of radiographic lesions and automated interpretation of dental radiographs**

* Caries detection: detection and characterization of proximal caries
* Diagnosis of vertical root fractures on CBCT images of endodontically treated and intact teeth.
* To stage tooth development.
* Computer based digital subtraction imaging
* Computer-assisted image analysis is useful to visualize and evaluate the bone architecture directly from the dental panoramic radiograph.
* 3-dimensional orthodontics visualization using patient models and OPGs.
* Bone density evaluation to predict osteoporosis using OPGs.
* Automatic segmentation of mandibular canal.
* Forensic dental imaging: Personal Identification System Using Dental Panoramic Radiograph
* Dental biometric

### Radiographic analysis

Image analysis using artificial intelligence in dentistry has been applied to various tasks-

* Tooth segmentation or localization,
* Bone quality (osteoporosis,Osteopenia) assessment
* Cephalometric landmark localization.
* Automated bone age assessment studies by setting interest in the hands and wrists on radiographs
* Classification of dental implant systems.

### Forensic dentistry

For age and sex determination in forensic odontology, Dentalradiology based methods being non-invasive and accurate are beneficial over conventional methods like sectioning or extracting teeth. They can therefore be used for both living and deceased individuals.

* Automated age estimation based on anatomical landmarks and tooth structures
* Automated sex estimation based on anatomical landmarks and tooth structures
* Identification of deceased based on dentition and tooth structures

## Orthodontics ([chaurasiaakhilanand49@gmail.com](mailto:chaurasiaakhilanand49@gmail.com), [ramtin.rhm@gmail.com](mailto:ramtin.rhm@gmail.com))

### Diagnostic and treatment pathways

Orthodontic treatments aim at correcting the malocclusion and repositioning the dentition in relation to the craniofacial structures in the most harmonious way. In order to accomplish a successful orthodontic treatment, it is necessary to conduct a patient interview, an examination, and gather proper records with the purpose of a comprehensive diagnosis. Clinical evaluations have to be systematic and organized so the clinician does not override any of the diagnostic items that include patient questionnaires and interviews, as well as an extra-oral and intra-oral clinical examination and para-clinical assessments. This can then result in an appropriate treatment plan and desirable treatment outcome. The following data is required before designing a comprehensive treatment plan:

1. Personal details
2. Chief complaint
3. Medical history
4. Dental History
5. TMJ assessment
6. Clinical examination
7. Para-clinical assessment
   * 1. Facial/Intraoral Photographs
     2. OPG Radiographs
     3. Lateral Cephalometry
     4. Study Casts
     5. Cone Beam Computed Tomography
     6. Hand and Wrist Radiograph

Due to the limitations in the length of this document, we go straight to the para-clinical data modalities since they are more applicable in the AI context.

A) Facial/Intraoral Photographs

Following items can be evaluated on facial or intraoral photographs:

1. The shape of head (Mesocephalic, Dolichocephalic, and Brachycephalic)
2. Facial form (Mesoprosopic, Euryprosopic, Leptoprosopic)
3. Facial symmetry
4. Facial proportions
5. Facial profile
6. Facial divergence
7. Anterior-posterior tooth/jaw relations
8. Lips examination
9. Nose examination
10. Chin examination
11. Nasolabial angle
12. Palatal evaluation
13. Frenal attachments
14. Overjet
15. Overbite
16. Transverse mal relations
17. Arch form
18. Smile analysis
19. General aesthetic evaluation

B) OPG Radiographs

It is necessary to assess general conditions regarding patients' dentition, before treatment planning. The following items can be evaluated on OPG radiographs:

* + - 1. Absence and presence of teeth, supernumerary teeth
      2. Evaluation of dentition age
      3. Evaluation of root formation
      4. Status of deciduous teeth
      5. General oral health (caries, previous restoration, etc)
      6. Presence of any pathologies

C) Lateral cephalograms

The main data modality used in orthodontics is lateral cephalograms. Main dental and skeletal orthodontics problems can be diagnosed through lateral cephalograms. Diagnosis in lateral cephalograms are based on the anatomic landmarks and their quantitative relations. The main landmarks are shown in Figure 2.

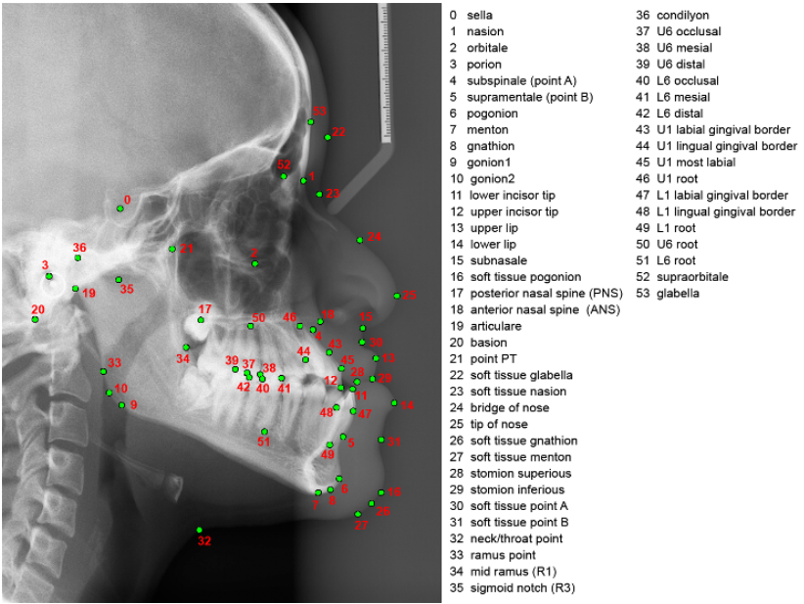


Image from "https://personalpages.manchester.ac.uk/staff/claudia.lindner/default\_files/curr\_project\_ceph.htm"

Figure 2 – Landmarks for Lateral cephalograms

Main orthodontics problem analysis, based on landmarks, are as listed in Table 2.

Table 2 – Eight clinical measurement methods for classification of anatomical types

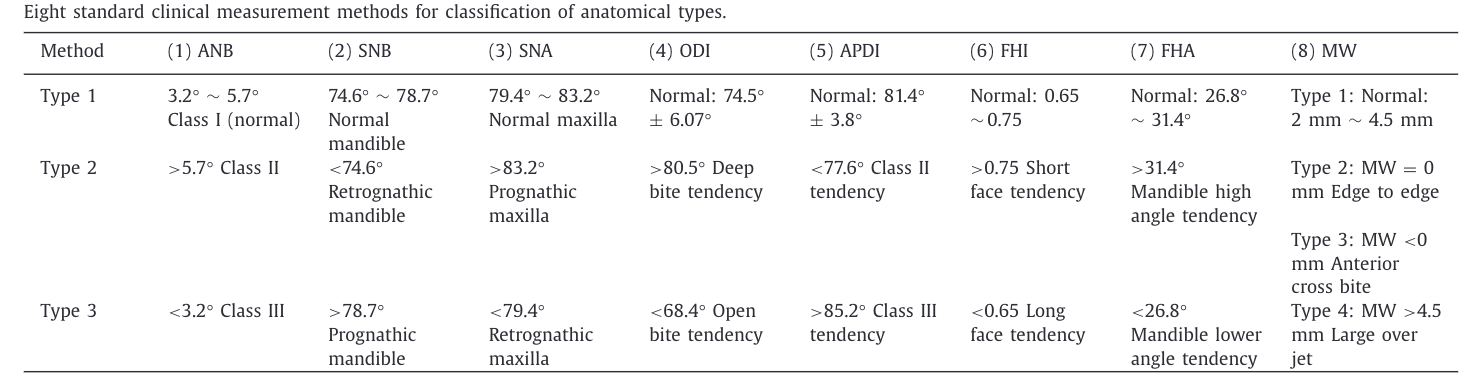


Image from "Wang et al, A benchmark for comparison of dental radiography analysis algorithms. *Medical image analysis,* 2016"

Other than the mentioned items, skeletal maturation degree can be evaluated on lateral cephalograms. Determination of skeletal maturation degree is necessary for treatment timing and reaching a desirable treatment outcome. The most common classification approach is based on the cervical vertebrae maturation degree as indicated in Figure 3.

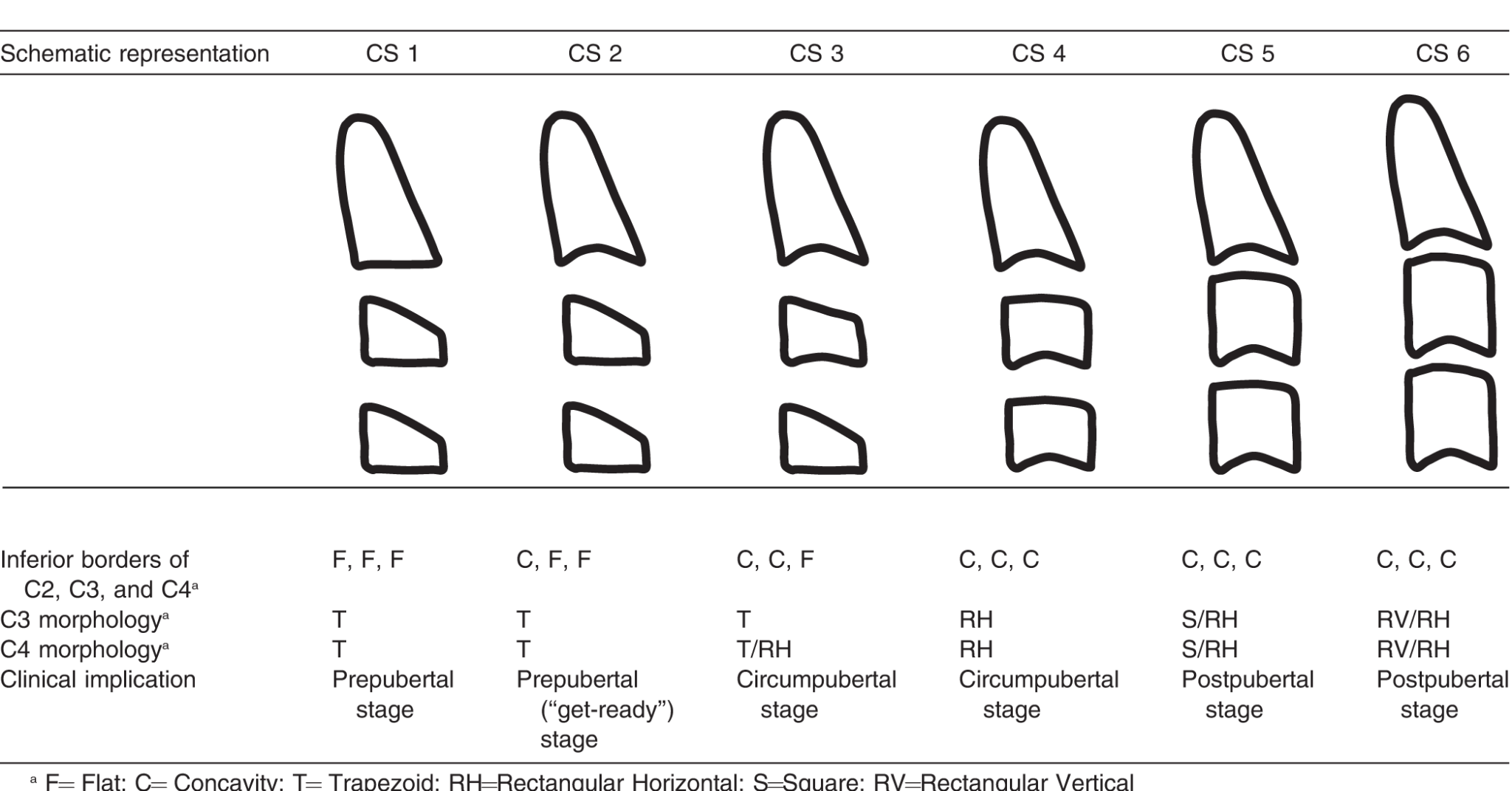


Image from "McNamara et al, The cervical vertebral maturation method: A user's guide, *The Angle Orthodontist*, 2018"

Figure 3 – Most common classification of skeletal maturation degree

*D) Study Casts:*

Following items can be evaluated on orthodontics study casts:

* + - 1. Occlusion
      2. Space analysis
      3. Arch form
      4. Arch symmetry
      5. Curve of spee
      6. Arch width

*E) Cone Beam Computed Tomography (CBCT)*:

Nowadays, CBCTs are becoming more popular in orthodontics. In clinical orthodontics, CBCT can provide volumetric information in three-dimensional form. As a result of advancements in CBCT technology, compared to conventional radiography, CBCT provides much more accurate and reliable information regarding diagnosis, treatment, and follow-up. CBCTs can be used in the following cases:

Evaluation of impacted and transposed teeth

Evaluation of root resorption of adjacent teeth

Evaluation of affected temporomandibular joints contributing to malocclusion

Air-way morphology analysis

Midpalatal suture maturation evaluation

* + - 1. Orthognathic surgery treatment planning

1. Locating the best temporary anchorage site for skeletal anchorage

*F) Hand and Wrist Radiograph*

Hand and Wrist radiograph have been used to predict the correct bone age (skeletal maturity) that can be compared with the patient's chronological age.

### Definition of the AI task

Various AI tasks can be defined in each modality for the purpose of enhancing orthodontic diagnosis and treatment planning. Due to the limitations in the length of this document, we only mention only those which already been addressed in the literature:

* + - 1. Facial/Intraoral Photographs
         1. Landmark detection
         2. Facial attractiveness score (as regression task)
         3. Facial asymmetry and distortion (as a classification task) \*

1. OPG Radiographs
2. All the AI tasks in dentistry can be beneficial in case of orthodontics, e.g. tooth numbering, caries detection, etc
3. Lateral cephalometrics
4. Landmark detection
5. Lateral cephalometric analysis (as a classification task) \*
6. Cervical maturation degree (as a classification task) \*
7. Study Casts
8. Arch form predication (as a classification task) \*
9. Cone Beam Computed Tomography
10. Maxillary suture assessment (as a segmentation task)

**\* NOTE –** These tasks can be defined in the context of AI in three different approaches:

1. One-stage automatic approach:   
   The AI model decides on its prediction as an end-to-end model, the model will be fed with the raw image and the diagnostic class will be the model's output.
2. Two-stage automatic approach:   
   The AI model is trained to detect landmarks. Then through some quantitative calculations between landmarks, they predict the diagnostic class.
3. Semi-automatic approach:   
   The AI model is trained on manually extracted data to predict the diagnostic class, e.g., some studies defined landmarks around cervical vertebrae and fed their models with features based on the relations between these landmarks. However, this approach has many shortcomings, and the practitioners cannot use all the advantages of AI in their practice.

### Current gold standard

As it can be seen, most orthodontics diagnoses are based on clinical and paraclinical evaluation. Almost in all of the AI tasks mentioned, there are no definitive gold standards and diagnoses are made by orthodontists. There is even very low inter/intra-observer agreement in some cases. To encounter this challenge, most diagnostic criteria are defined based on quantitative relations based on landmarks (angle, proportions, distance, *etc*). However, this approach is very time-consuming and prone to the error itself.

### Relevance and impact of an AI solution

In the diagnosis and evaluation of orthodontic problems, AI can be effective. Orthodontic diagnosis and treatment planning for malocclusion can be a very difficult and time-consuming procedure. New practitioners often have trouble making these decisions since they are traditionally made on the basis of clinical experience. AI approaches might be helpful for solving these problems. Cephalometric analysis, for example, is frequently used to detect skeletal and dental anomalies in orthodontics. Manually locating cephalometric landmarks requires a lot of time and is prone to error. The use of deep learning models for landmark detection has achieved outstanding results in recent studies.

Precision medicine is one of the most important applications for AI-based systems in orthodontics. To enhance treatment outcomes, researchers are considering customized treatment approaches based on the characteristics of each patient. Orthodontic and precision medicine advancements have led to breakthroughs in the development of customized treatment approaches. Precision orthodontics has been reported to be the next paradigm shift in orthodontic treatments. AI is capable of enabling and enriching precision medicine approaches based on data.

### Existing AI solutions

In research, a recently published scoping review identified 49 studies that used AI approaches in orthodontics. It has been reported that the number of publications in this field is growing. See Figure 4.

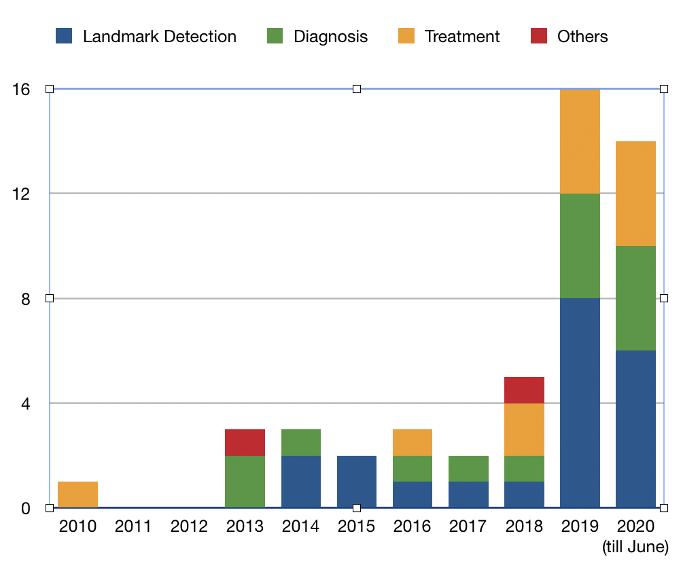


Image from "Mohammad-rahimi et al, Machine learning and orthodontics, current trends and the future opportunities: A scoping review. *American Journal of Orthodontics and Dentofacial Orthopedics*, 2021"

Figure 4 – Survey of studies on AI approaches in orthodontics

Most of the studies reported the AI model's application on the lateral cephalometry landmark detection and they reached a good performance here. However, we are at the very beginning of orthodontics utilizing AI diagnosis and decision making.

In industry, there are a few companies trying to provide AI solutions in orthodontics. They can be divided into the following categories:

1. Diagnostic assistant tools
   1. WeDoCeph: This company provides a tool for cephalometric tracing analysis on 2D lateral or PA X-rays
   2. CephX: This company provides a tool for cephalometric tracing analysis on 2D lateral cephalometry and 3D CBCTs. It also can do airway volume analysis on 3D CBCTs.
2. Patient management and monitoring
3. Dental monitoring: Engage new patients using AI-generated reporting and smile simulations.
4. AI-based aligners
5. 3D Predict
6. Soft Smile

## Ethical considerations

The rapidly evolving field of AI and digital technology in the fields of medicine and public health raises a number of ethical, legal, and social concerns that have to be considered in this context. They are discussed in deliverable [DEL01](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B0505B020-362C-45B2-94BF-215D2EBBD8F5%7D&file=DEL01.docx&action=default) "*AI4H ethics considerations,"* which was developed by the working group on "Ethical considerations on AI4H" (WG-Ethics). This section refers to DEL01 and should reflect the ethical considerations of the TG-Dental.

* What are the ethical implications of applying the AI model in real-world scenarios?
* What are the ethical implications of introducing benchmarking (having the benchmarking in place itself has some ethical risks; e.g., if the test data are not representative for a use case, the data might create the illusion of safety and put people at risk)?
* What are the ethical implications of collecting the data for benchmarking (e.g., how is misuse of data addressed, is there the need for an ethics board approval for clinical data, is there the need for consent management for sharing patient data, and what are the considerations about data ownership/data custodianship)?
* What risks face individuals and society if the benchmarking is wrong, biased, or inconsistent with reality on the ground?
* How is the privacy of personal health information protected (e.g., in light of longer data retention for documentation, data deletion requests from users, and the need for an informed consent of the patients to use data)?
* How is ensured that benchmarking data are representative and that an AI offers the same performance and fairness (e.g., can the same performance in high, low-, and middle-income countries be guaranteed; are differences in race, sex, and minority ethnic populations captured; are considerations about biases, when implementing the same AI application in a different context included; is there a review and clearance of 'inclusion and exclusion criteria' for test data)?
* What are your experiences and learnings from addressing ethics in your TG?

# Existing work on benchmarking

This section focuses on the existing benchmarking processes in the context of AI and Dental Diagnostics and Digital Dentistry for quality assessment. It addresses different aspects of the existing work on benchmarking of AI systems (e.g., relevant scientific publications, benchmarking frameworks, scores and metrics, and clinical evaluation attempts). The goal is to collect all relevant learnings from previous benchmarking that could help to implement the benchmarking process in this topic group.

## Subtopic [A]

*Topic driver: If there are subtopics in your topic group, describe the existing work on benchmarking for the first subtopic [A] in this section. If there are no sub-topics, you can remove the "Subtopic" outline level, but - of course - you need to keep the subsections below!*

### Publications on benchmarking systems

While a representative comparable benchmarking for Dental Diagnostics and Digital Dentistry does not yet exist, some work has been done in the scientific community assessing the performance of such systems. This section summarizes insights from the most relevant publications on this topic. It covers parts of the deliverable DEL07 *"AI for health evaluation considerations,"* [DEL07\_1](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B565EEC0A-D755-41C8-AC68-37B4C38C953F%7D&file=DEL07_1.docx&action=default) *"AI4H evaluation process description,"* [DEL07\_2](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B58679341-C738-40F0-A822-3AC2B24DD09F%7D&file=DEL07_2.docx&action=default) *"AI technical test specification*,*"* [DEL07\_3](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BA3088882-F82B-493B-B1C5-49CFF0EEEFA8%7D&file=DEL07_3.docx&action=default) *"Data and artificial intelligence assessment methods (DAISAM),"* and [DEL07\_4](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BB846B260-373A-41FC-A892-EE5BBCFE3CF8%7D&file=DEL07_4.docx&action=default) *"Clinical Evaluation of AI for health"*.

* What is the most relevant peer-reviewed scientific publications on benchmarking or objectively measuring the performance of systems in your topic?
* State what are the most relevant approaches used in literature?
* Which scores and metrics have been used?
* How were test data collected?
* How did the AI system perform and how did it compare the current gold standard? Is the performance of the AI system equal across less represented groups? Can it be compared to other systems with a similar benchmarking performance and the same clinically meaningful endpoint (addressing comparative efficacy)?
* How can the utility of the AI system be evaluated in a real-life clinical environment (also considering specific requirements, e.g., in a low- and middle-income country setting)?
* Have there been clinical evaluation attempts (e.g., internal and external validation processes) and considerations about the use in trial settings?
* What are the most relevant gaps in the literature (what is missing concerning AI benchmarking)?

### Benchmarking by AI developers

All developers of AI solutions for Dental Diagnostics and Digital Dentistry implemented internal benchmarking systems for assessing the performance. This section will outline the insights and learnings from this work of relevance for benchmarking in this topic group.

* What are the most relevant learnings from the benchmarking by AI developers in this field (e.g., ask the members of your topic group what they want to share on their benchmarking experiences)?
* Which scores and metrics have been used?
* How did they approach the acquisition of test data?

### Relevant existing benchmarking frameworks

Triggered by the hype around AI, recent years have seen the development of a variety of benchmarking platforms where AIs can compete for the best performance on a determined dataset. Given the high complexity of implementing a new benchmarking platform, the preferred solution is to use an established one. This section reflects on the different existing options that are relevant for this topic group and includes considerations of using the assessment platform that is currently developed by FG-AI4H and presented by deliverable [DEL07\_5](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B8BFCFF21-3908-4BAD-AB9C-9814EB3F9B36%7D&file=DEL07_5.docx&action=default) *"FG-AI4H assessment platform"* (the deliverable explores options for implementing an assessment platform that can be used to evaluate AI for health for the different topic groups).

* Which benchmarking platforms could be used for this topic group (e.g., EvalAI, AIcrowd, Kaggle, and CodaLab)?
* Are the benchmarking assessment platforms discussed, used, or endorsed by FG-AI4H an option?
* Are there important features in this topic group that require special attention?
* Is the reporting flexible enough to answer the questions stakeholders want to get answered by the benchmarking?
* What are the relative advantages and disadvantages of these diverse solutions?

## Subtopic [B]

*Topic driver: If there are subtopics in your topic group, describe the existing work on benchmarking for the second subtopic [B] in this section using the same subsection structure as above. (If there are no sub-topics, you can remove the "Subtopic" outline level.)*

# Benchmarking by the topic group

This section describes all technical and operational details regarding the benchmarking process for the Dental Diagnostics and Digital Dentistry AI task including subsections for each version of the benchmarking that is iteratively improved over time.

It reflects the considerations of various deliverables: [DEL05](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B2012357A-941E-44BD-B965-370D7829F52C%7D&file=DEL05.docx&action=default) *"Data specification"* (introduction to deliverables 5.1-5.6), [DEL05\_1](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B19830259-F63B-42D4-A408-48C854D6C124%7D&file=DEL05_1.docx&action=default)*"Data requirements"* (which lists acceptance criteria for data submitted to FG-AI4H and states the governing principles and rules), [DEL05\_2](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B25141F77-E59A-45F1-B081-185C2194FE67%7D&file=DEL05_2.docx&action=default) *"Data acquisition"*, [DEL05\_3](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B05D8938E-BC2A-4A62-BCB0-1FD46AA72235%7D&file=DEL05_3.docx&action=default) *"Data annotation specification"*, [DEL05\_4](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BF267A95C-4C5B-4D63-A135-58AF487C3AD3%7D&file=DEL05_4.docx&action=default) *"Training and test data specification"* (which provides a systematic way of preparing technical requirement specifications for datasets used in training and testing of AI models), [DEL05\_5](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B71FE8B9D-ACB3-48CE-AA3F-136409B550A4%7D&file=DEL05_5.docx&action=default) *"Data handling"* (which outlines how data will be handled once they are accepted), [DEL05\_6](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B5C95327E-96A5-4175-999E-3EDB3ED147C3%7D&file=DEL05_6.docx&action=default) *"Data sharing practices"* (which provides an overview of the existing best practices for sharing health-related data based on distributed and federated environments, including the requirement to enable secure data sharing and addressing issues of data governance), [DEL06](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BF5967277-90C8-4252-A0B9-43A5692F35E2%7D&file=DEL06.docx&action=default) *"AI training best practices specification"* (which reviews best practices for proper AI model training and guidelines for model reporting), [DEL07](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B47E77197-F87B-49F4-80B3-2DD949A5F185%7D&file=DEL07.docx&action=default)*"AI for health evaluation considerations"* (which discusses the validation and evaluation of AI for health models, and considers requirements for a benchmarking platform), [DEL07\_1](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B565EEC0A-D755-41C8-AC68-37B4C38C953F%7D&file=DEL07_1.docx&action=default) *"AI4H evaluation process description"* (which provides an overview of the state of the art of AI evaluation principles and methods and serves as an initiator for the evaluation process of AI for health), [DEL07\_2](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B58679341-C738-40F0-A822-3AC2B24DD09F%7D&file=DEL07_2.docx&action=default) *"AI technical test specification"* (which specifies how an AI can and should be tested *in silico*), [DEL07\_3](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BA3088882-F82B-493B-B1C5-49CFF0EEEFA8%7D&file=DEL07_3.docx&action=default) *"Data and artificial intelligence assessment methods (DAISAM)"* (which provides the reference collection of WG-DAISAM on assessment methods of data and AI quality evaluation), [DEL07\_4](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BB846B260-373A-41FC-A892-EE5BBCFE3CF8%7D&file=DEL07_4.docx&action=default)*"Clinical Evaluation of AI for health"* (which outlines the current best practices and outstanding issues related to clinical evaluation of AI models for health), [DEL07\_5](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B8BFCFF21-3908-4BAD-AB9C-9814EB3F9B36%7D&file=DEL07_5.docx&action=default) *"FG-AI4H assessment platform"* (which explores assessment platform options that can be used to evaluate AI for health for the different topic groups), [DEL09](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B3E940987-8D75-44B8-85E4-F0E475964F15%7D&file=DEL09.docx&action=default) *"AI for health applications and platforms"* (which introduces specific considerations of the benchmarking of mobile- and cloud-based AI applications in health), [DEL09\_1](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B1A2EC8D5-53CA-4C8C-9B09-B61CA6F428C5%7D&file=DEL09_1.docx&action=default) *"Mobile based AI applications,"* and [DEL09\_2](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B3B5A31DE-D3B1-4EC1-A261-2C2E19F73810%7D&file=DEL09_2.docx&action=default) *"Cloud-based AI applications"* (which describe specific requirements for the development, testing and benchmarking of mobile- and cloud-based AI applications).

## Subtopic [A]

*Topic driver: Please refer to the above comments concerning subtopics.*

The benchmarking of Dental Diagnostics and Digital Dentistry is going to be developed and improved continuously to reflect new features of AI systems or changed requirements for benchmarking. This section outlines all benchmarking versions that have been implemented thus far and the rationale behind them. It serves as an introduction to the subsequent sections, where the actual benchmarking methodology for each version will be described.

* Which benchmarking iterations have been implemented thus far?
* What important new features are introduced with each iteration?
* What are the next planned iterations and which features are they going to add?

### Benchmarking version [Y]

This section includes all technological and operational details of the benchmarking process for the benchmarking version [Y] (latest version, chronologically reversed order).

#### Overview

This section provides an overview of the key aspects of this benchmarking iteration, version [Y].

* What is the overall scope of this benchmarking iteration (e.g., performing a first benchmarking, adding benchmarking for multi-morbidity, or introducing synthetic-data-based robustness scoring)?
* What features have been added to the benchmarking in this iteration?

#### Benchmarking methods

This section provides details about the methods of the benchmarking version [Y]. It contains detailed information about the benchmarking system architecture, the dataflow and the software for the benchmarking process (e.g., test scenarios, data sources, and legalities).

##### Benchmarking system architecture

This section covers the architecture of the benchmarking system. For well-known systems, an overview and reference to the manufacturer of the platform is sufficient. If the platform was developed by the topic group, a more detailed description of the system architecture is required.

* How does the architecture look?
* What are the most relevant components and what are they doing?
* How do the components interact on a high level?
* What underlying technologies and frameworks have been used?
* How does the hosted AI model get the required environment to execute correctly? What is the technology used (e.g., Docker/Kubernetes)?

##### Benchmarking system dataflow

This section describes the dataflow throughout the benchmarking architecture.

* How do benchmarking data access the system?
* Where and how (data format) are the data, the responses, and reports of the system stored?
* How are the inputs and the expected outputs separated?
* How are the data sent to the AI systems?
* Are the data entries versioned?
* How does the lifecycle for the data look?

##### Safe and secure system operation and hosting

*From a technical point of view, the benchmarking process is not particularly complex. It is more about agreeing on something in the topic group with potentially many competitors and implementing the benchmarking in a way that cannot be compromised. This section describes how the benchmarking system, the benchmarking data, the results, and the reports are protected against manipulation, data leakage, or data loss. Topic groups that use ready-made software might be able to refer to the corresponding materials of the manufacturers of the benchmarking system.*

This section addresses security considerations about the storage and hosting of data (benchmarking results and reports) and safety precautions for data manipulation, data leakage, or data loss.

In the case of a manufactured data source (vs. self-generated data), it is possible to refer to the manufacturer's prescriptions.

* Based on the architecture, where is the benchmarking vulnerable to risk and how have these risks been mitigated (e.g., did you use a threat modelling approach)? A discussion could include:
* Could someone access the benchmarking data before the actual benchmarking process to gain an advantage?
* What safety control measures were taken to manage risks to the operating environment?
* Could someone have changed the AI results stored in the database (your own and/or that of competitors)?
* Could someone attack the connection between the benchmarking and the AI (e.g., to make the benchmarking result look worse)?
* How is the hosting system itself protected against attacks?
* How are the data protected against data loss (e.g., what is the backup strategy)?
* What mechanisms are in place to ensure that proprietary AI models, algorithms and trade-secrets of benchmarking participants are fully protected?
* How is it ensured that the correct version of the benchmarking software and the AIs are tested?
* How are automatic updates conducted (e.g., of the operating system)?
* How and where is the benchmarking hosted and who has access to the system and the data (e.g., virtual machines, storage, and computing resources, configurational settings)?
* How is the system's stability monitored during benchmarking and how are attacks or issues detected?
* How are issues (e.g., with a certain AI) documented or logged?
* In case of offline benchmarking, how are the submitted AIs protected against leakage of intellectual property?

##### Benchmarking process

This section describes how the benchmarking looks from the registration of participants, through the execution and resolution of conflicts, to the final publication of the results.

* How are new benchmarking iterations scheduled (e.g., on demand or quarterly)?
* How do possible participants learn about an upcoming benchmarking?
* How can one apply for participation?
* What information and metadata do participants have to provide (e.g., AI autonomy level assignment (IMDRF), certifications, AI/machine learning technology used, company size, company location)?
* Are there any contracts or legal documents to be signed?
* Are there inclusion or exclusion criteria to be considered?
* How do participants learn about the interface they will implement for the benchmarking (e.g., input and output format specification and application program interface endpoint specification)?
* How can participants test their interface (e.g., is there a test dataset in case of file-based offline benchmarking or are there tools for dry runs with synthetic data cloud-hosted application program interface endpoints)?
* Who is going to execute the benchmarking and how is it ensured that there are no conflicts of interest?
* If there are problems with an AI, how are problems resolved (e.g., are participants informed offline that their AI fails to allow them to update their AI until it works? Or, for online benchmarking, is the benchmarking paused? Are there timeouts?)?
* How and when will the results be published (e.g., always or anonymized unless there is consent)? With or without seeing the results first? Is there an interactive drill-down tool or a static leader board? Is there a mechanism to only share the results with stakeholders approved by the AI provider as in a credit check scenario?
* In case of online benchmarking, are the benchmarking data published after the benchmarking? Is there a mechanism for collecting feedback or complaints about the data? Is there a mechanism of how the results are updated if an error was found in the benchmarking data?

#### AI input data structure for the benchmarking

This section describes the input data provided to the AI solutions as part of the benchmarking of Dental Diagnostics and Digital Dentistry. It covers the details of the data format and coding at the level of detail needed to submit an AI for benchmarking. This is the only TDD section addressing this topic. Therefore, the description needs to be complete and precise. This section does *not* contain the encoding of the labels for the expected outcomes. It is only about the data the AI system will see as part of the benchmarking.

* What are the general data types that are fed in the AI model?
* How exactly are they encoded? For instance, discuss:
  + The exact data format with all fields and metadata (including examples or links to examples)
  + Ontologies and terminologies
  + Resolution and data value ranges (e.g., sizes, resolutions, and compressions)
  + Data size and data dimensionality

#### AI output data structure

Similar to the input data structure for the benchmarking, this section describes the output data the AI systems are expected to generate in response to the input data. It covers the details of the data format, coding, and error handling at the level of detail needed for an AI to participate in the benchmarking.

* What are the general data output types returned by the AI and what is the nature of the output (e.g., classification, detection, segmentation, or prediction)?
  + How exactly are they encoded? Discuss points like:
    - The exact data format with all fields and metadata (including examples or links to examples)
    - Ontologies and terminologies
* What types of errors should the AI generate if something is defective?

#### Test data label/annotation structure

*Topic driver: Please describe how the expected AI outputs are encoded in the benchmarking test data. Please note that it is essential that the AIs never access the expected outputs to prevent cheating. The topic group should carefully discuss whether more detailed labelling is needed. Depending on the topic, it might make sense to separate between the best possible output of the AI given the input data and the correct disease (that might be known but cannot be derived from the input data alone). Sometimes it is also helpful to encode acceptable other results or results that can be clearly ruled out given the evidence. This provides a much more detailed benchmarking with more fine-grained metrics and expressive reports than the often too simplistic leader boards of many AI competitions.*

While the AI systems can only receive the input data described in the previous sections, the benchmarking system needs to know the expected correct answer (sometimes called 'labels') for each element of the input data so that it can compare the expected AI output with the actual one. Since this is only needed for benchmarking, it is encoded separately. The details are described in the following section.

* What are the general label types (e.g., expected results, acceptable results, correct results, and impossible results)?
* How exactly are they encoded? Discuss points like:
  + The exact data format with all fields and metadata (including examples or links to examples)
  + Ontologies and terminologies
* How are additional metadata about labelling encoded (e.g., author, data, pre-reviewing details, dates, and tools)?
* How and where are the labels embedded in the input data set (including an example; e.g., are there separate files or is it an embedded section in the input data that is removed before sending to the AI)?

#### Scores and metrics

*Topic drivers: This section describes the scores and metrics that are used for benchmarking. It includes details about the testing of the AI model and its effectiveness, performance, transparency, etc. Please note that this is only the description of the scores and metrics actually used in* ***this*** *benchmarking iteration. A general description of the state of the art of scores and metrics and how they have been used in previous work is provided in section 3.*

Scores and metrics are at the core of the benchmarking. This section describes the scores and metrics used to measure the performance, robustness, and general characteristics of the submitted AI systems.

* Who are the stakeholders and what decisions should be supported by the scores and metrics of the benchmarking?
* What general criteria have been applied for selecting scores and metrics?
* What scores and metrics have been chosen/defined for robustness?
* What scores and metrics have been chosen/defined for medical performance?
* What scores and metrics have been chosen/defined for non-medical performance?
  + Metrics for technical performance tracking (e.g., monitoring and reporting when the performance accuracy of the model drops below a predefined threshold level as a function of time; computational efficiency rating, response times, memory consumption)
* What scores and metrics have been chosen/defined for model explainability?
* Describe for each aspect
  + The exact definition/formula of the score based on the labels and the AI output data structures defined in the previous sections and how they are aggregated/accumulated over the whole dataset (e.g., for a single test set entry, the result might be the probability of the expected correct class which is then aggregated to the average probability of the correct class)
  + Does it use some kind of approach for correcting dataset bias (e.g., the test dataset usually has a different distribution compared to the distribution of a condition in a real-world scenario. For estimating the real-world performance, metrics need to compensate this difference.)
  + What are the origins of these scores and metrics?
  + Why were they chosen?
  + What are the known advantages and disadvantages?
  + How easily can the results be compared between or among AI solutions?
  + Can the results from benchmarking iterations be easily compared or does it depend too much on the dataset (e.g., how reproducible are the results)?
* How does this consider the general guidance of WG-DAISAM in [DEL07\_3](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BA3088882-F82B-493B-B1C5-49CFF0EEEFA8%7D&file=DEL07_3.docx&action=default) "Data and artificial intelligence assessment methods (DAISAM)"?
* Have there been any relevant changes compared to previous benchmarking iterations? If so, why?

#### Test dataset acquisition

Test dataset acquisition includes a detailed description of the test dataset for the AI model and, in particular, its benchmarking procedure including quality control of the dataset, control mechanisms, data sources, and storage.

* How does the overall dataset acquisition and annotation process look?
* How have the data been collected/generated (e.g., external sources vs. a process organized by the TG)?
* Have the design goals for the benchmarking dataset been reached (e.g., please provide a discussion of the necessary size of the test dataset for relevant benchmarking results, statistical significance, and representativeness)?
* How was the dataset documented and which metadata were collected?
  + Where were the data acquired?
  + Were they collected in an ethical-conform way?
  + Which legal status exists (e.g., intellectual property, licenses, copyright, privacy laws, patient consent, and confidentiality)?
  + Do the data contain 'sensitive information' (e.g., socially, politically, or culturally sensitive information; personal identifiable information)? Are the data sufficiently anonymized?
  + What kind of data anonymization or deidentification has been applied?
  + Are the data self-contained (i.e., independent from externally linked datasets)?
  + How is the bias of the dataset documented (e.g., sampling or measurement bias, representation bias, or practitioner/labelling bias)?
  + What additional metadata were collected (e.g., for a subsequent detailed analysis that compares the performance on old cases with new cases)? How was the risk of benchmarking participants accessing the data?
* Have any scores, metrics, or tests been used to assess the quality of the dataset (e.g., quality control mechanisms in terms of data integrity, data completeness, and data bias)?
* Which inclusion and exclusion criteria for a given dataset have been applied (e.g., comprehensiveness, coverage of target demographic setting, or size of the dataset)?
* How was the data submission, collection, and handling organized from the technical and operational point of view (e.g., folder structures, file formats, technical metadata encoding, compression, encryption, and password exchange)?
* Specific data governance derived by the general data governance document (currently [F-103](https://www.itu.int/en/ITU-T/focusgroups/ai4h/Documents/FGAI4H-F-103-DataPolicy.pdf) and the deliverables beginning with [DEL05](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B2012357A-941E-44BD-B965-370D7829F52C%7D&file=DEL05.docx&action=default))
* How was the overall quality, coverage, and bias of the accumulated dataset assessed (e.g., if several datasets from several hospitals were merged with the goal to have better coverage of all regions and ethnicities)?
* Was any kind of post-processing applied to the data (e.g., data transformations, repackaging, or merging)?
* How was the annotation organized?
  + How many annotators/peer reviewers were engaged?
  + Which scores, metrics, and thresholds were used to assess the label quality and the need for an arbitration process?
  + How have inter-annotator disagreements been resolved (i.e., what was the arbitration process)?
  + If annotations were part of the submitted dataset, how was the quality of the annotations controlled?
  + How was the annotation of each case documented?
  + Were metadata on the annotation process included in the data (e.g., is it possible to compare the benchmarking performance based on the annotator agreement)?
* Were data/label update/amendment policies and/or criteria in place?
* How was access to test data controlled (e.g., to ensure that no one could access, manipulate, and/or leak data and data labels)? Please address authentication, authorization, monitoring, logging, and auditing
* How was data loss avoided (e.g., backups, recovery, and possibility for later reproduction of the results)?
* Is there assurance that the test dataset is undisclosed and was never previously used for training or testing of any AI model?
* What mechanisms are in place to ensure that test datasets are used only once for benchmarking? (Each benchmarking session will need to run with a new and previously undisclosed test dataset to ensure fairness and no data leakage to subsequent sessions)

#### Data sharing policies

This section provides details about legalities in the context of benchmarking. Each dataset that is shared should be protected by special agreements or contracts that cover, for instance, the data sharing period, patient consent, and update procedure (see also [DEL05\_5](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B71FE8B9D-ACB3-48CE-AA3F-136409B550A4%7D&file=DEL05_5.docx&action=default) on *data handling* and [DEL05\_6](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B5C95327E-96A5-4175-999E-3EDB3ED147C3%7D&file=DEL05_6.docx&action=default) on *data sharing practices*).

* Which legal framework was used for data sharing?
* Was a data sharing contract signed and what was the content? Did it contain:
  + Purpose and intended use of data
  + Period of agreement
  + Description of data
  + Metadata registry
  + Data harmonization
  + Data update procedure
  + Data sharing scenarios
    - Data can be shared in public repositories
    - Data are stored in local private databases (e.g., hospitals)
  + Rules and regulation for patients' consent
  + Data anonymization and de-identification procedure
  + Roles and responsibilities
    - Data provider
    - Data protection officer
    - Data controllers
    - Data processors
    - Data receivers
* Which legal framework was used for sharing the AI?
* Was a contract signed and what was the content?

#### Baseline acquisition

The main purpose of benchmarking is to provide stakeholders with the numbers they need to decide whether AI models provide a viable solution for a given health problem in a designated context. To achieve this, the performance of the AI models needs to be compared with available options achieving the same clinically meaningful endpoint. This, in turn, requires data on the performance of the alternatives, ideally using the same benchmarking data. As the current alternatives typically involve doctors, it might make sense to combine the test data acquisition and labelling with additional tasks that allow the performance of the different types of health workers to be assessed.

* Does this topic require comparison of the AI model with a baseline (gold standard) so that stakeholders can make decisions?
* Is the baseline known for all relevant application contexts (e.g., region, subtask, sex, age group, and ethnicity)?
* Was a baseline assessed as part of the benchmarking?
* How was the process of collecting the baseline organized? If the data acquisition process was also used to assess the baseline, please describe additions made to the process described in the previous section.
* What are the actual numbers (e.g., for the performance of the different types of health workers doing the task)?

#### Reporting methodology

*After the benchmarking, the next step is to describe how the results are compiled into reports that allow stakeholders to make decisions (e.g., which AI systems can be used to solve a pre-diagnosis task in an offline –field –clinic scenario in central America). For some topic groups, the report might be as simple as a classical AI competition leader board using the most relevant performance indicator. For other tasks, it could be an interactive user interface that allows stakeholders to compare the performance of the different AI systems in a designated context with existing non-AI options. For the latter, statistical issues must be carefully considered (e.g., the multiple comparisons problem). Sometimes, a hybrid of prepared reports on common aspects are generated in addition to interactive options. There is also the question of how and where the results are published and to what degree benchmarking participants can opt in or opt out of the publication of their performance.*

This section discusses how the results of the benchmarking runs will be shared with the participants, stakeholders, and general public.

* What is the general approach for reporting results (e.g., leader board vs. drill down)?
* How can participants analyse their results (e.g., are there tools or are detailed results shared with them)?
* How are the participants and their AI models (e.g., versions of model, code, and configuration) identified?
* What additional metadata describing the AI models have been selected for reporting?
* How is the relationship between AI results, baselines, previous benchmarking iterations, and/or other benchmarking iterations communicated?
* What is the policy for sharing participant results (e.g., opt in or opt out)? Can participants share their results privately with their clients (e.g., as in a credit check scenario)?
* What is the publication strategy for the results (e.g., website, paper, and conferences)?
* Is there an online version of the results?
* Are there feedback channels through which participants can flag technical or medical issues (especially if the benchmarking data was published afterwards)?
* Are there any known limitations to the value, expressiveness, or interpretability of the reports?

#### Result

This section gives an overview of the results from runs of this benchmarking version of your topic. Even if your topic group prefers an interactive drill-down rather than a leader board, pick some context of common interest to give some examples.

* When was the benchmarking executed?
* Who participated in the benchmarking?
* What overall performance of the AI systems concerning medical accuracy, robustness, and technical performance (minimum, maximum, average etc.) has been achieved?
* What are the results of this benchmarking iteration for the participants (who opted in to share their results)?

#### Discussion of the benchmarking

This section discusses insights of this benchmarking iterations and provides details about the 'outcome' of the benchmarking process (e.g., giving an overview of the benchmark results and process).

* What was the general outcome of this benchmarking iteration?
* How does this compare to the goals for this benchmarking iteration (e.g., was there a focus on a new aspect to benchmark)?
* Are there real benchmarking results and interesting insights from this data?
  + How was the performance of the AI system compared to the baseline?
  + How was the performance of the AI system compared to other benchmarking initiatives (e.g., are the numbers plausible and consistent with clinical experience)?
  + How did the results change in comparison to the last benchmarking iteration?
* Are there any technical lessons?
  + Did the architecture, implementation, configuration, and hosting of the benchmarking system fulfil its objectives?
  + How was the performance and operational efficiency of the benchmarking itself (e.g., how long did it take to run the benchmarking for all AI models vs. one AI model; was the hardware sufficient)?
* Are there any lessons concerning data acquisition?
  + Was it possible to collect enough data?
  + Were the data as representative as needed and expected?
  + How good was the quality of the benchmarking data (e.g., how much work went into conflict resolution)?
  + Was it possible to find annotators?
  + Was there any relevant feedback from the annotators?
  + How long did it take to create the dataset?
* Is there any feedback from stakeholders about how the benchmarking helped them with decision-making?
  + Are metrics missing?
  + Do the stakeholders need different reports or additional metadata (e.g., do they need the "offline capability" included in the AI metadata so that they can have a report on the best offline system for a certain task)?
* Are there insights on the benchmarking process?
  + How was the interest in participation?
  + Are there reasons that someone could not join the benchmarking?
  + What was the feedback of participants on the benchmarking processes?
  + How did the participants learn about the benchmarking?

#### Retirement

*Topic driver: describe what happens to the benchmarking data and the submitted AI models after the benchmarking.*

This section addresses what happens to the AI system and data after the benchmarking activity is completed. It might be desirable to keep the database for traceability and future use. Alternatively, there may be security or privacy reasons for deleting the data. Further details can be found in the reference document of this section [DEL04](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BC68833D1-9B31-4E8E-8A4A-3939D7DEA56F%7D&file=DEL04.docx&action=default) "*AI software lifecycle specification"* (identification of standards and best practices that are relevant for the AI for health software life cycle).

* What happens with the data after the benchmarking (e.g., will they be deleted, stored for transparency, or published)?
* What happens to the submitted AI models after the benchmarking?
* Could the results be reproduced?
* Are there legal or compliance requirements to respond to data deletion requests?

### Benchmarking version [X]

This section includes all technological and operational details of the benchmarking process for the benchmarking version [X].

*Topic driver: Provide details of previous benchmarking versions here using the same subsection structure as above.*

## Subtopic [B]

*Topic driver: If there are subtopics in your topic group, please provide the details about the benchmarking of the second subtopic [B] here using the same subsection structure as above (please refer to earlier comments – in red fonts - concerning subtopics).*

# Overall discussion of the benchmarking

This section discusses the overall insights gained from benchmarking work in this topic group. This should not be confused with the discussion of the results of a concrete benchmarking run (e.g., in 4.2.11).

* What is the overall outcome of the benchmarking thus far?
* Have there been important lessons?
* Are there any field implementation success stories?
* Are there any insights showing how the benchmarking results correspond to, for instance, clinical evaluation?
* Are there any insights showing the impact (e.g., health economic effects) of using AI systems that were selected based on the benchmarking?
* Was there any feedback from users of the AI system that provides insights on the effectiveness of benchmarking?
  + Did the AI system perform as predicted relative to the baselines?
  + Did other important factors prevent the use of the AI system despite a good benchmarking performance (e.g., usability, access, explainability, trust, and quality of service)?
* Were there instances of the benchmarking not meeting the expectations (or helping) the stakeholders? What was learned (and changed) as a result?
* What was learned from executing the benchmarking process and methodology (e.g., technical architecture, data acquisition, benchmarking process, benchmarking results, and legal/contractual framing)?

# Regulatory considerations

*Topic Driver: This section reflects the requirements of the working group on* [***Regulatory considerations on AI for health (WG-RC)***](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/wg/SitePages/WG-RC.aspx) *and their various deliverables. It is* ***NOT requested to re-produce regulatory frameworks****, but to show the regulatory frameworks that have to be applied in the context of your AIs and their benchmarking (****2 pages max****).*

For AI-based technologies in healthcare, regulation is not only crucial to ensure the safety of patients and users, but also to accomplish market acceptance of these devices. This is challenging because there is a lack of universally accepted regulatory policies and guidelines for AI-based medical devices. To ensure that the benchmarking procedures and validation principles of FG-AI4H are secure and relevant for regulators and other stakeholders, the working group on *"*[*Regulatory considerations on AI for health"* *(WG-RC)*](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/wg/SitePages/WG-RC.aspx) compiled the requirements that consider these challenges.

The deliverables with relevance for regulatory considerations are [DEL02](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BF2F46A99-7457-4BC8-81A3-0E1E63D6072A%7D&file=DEL02.docx&action=default) *"AI4H regulatory considerations"* (which provides an educational overview of some key regulatory considerations), [DEL02\_1](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B6AF7C004-8BCE-4151-9F44-45F041A1EB1D%7D&file=DEL02_1.docx&action=default) *"Mapping of IMDRF essential principles to AI for health software",* and[DEL02\_2](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7B1ED0D4D1-876C-4A0F-AEF7-06D3F445F5E6%7D&file=DEL02_2.docx&action=default) *"Guidelines for AI based medical device (AI-MD): Regulatory requirements"* (which provides a checklist to understand expectations of regulators, promotes step-by-step implementation of safety and effectiveness of AI-based medical devices, and compensates for the lack of a harmonized standard). [DEL04](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BC68833D1-9B31-4E8E-8A4A-3939D7DEA56F%7D&file=DEL04.docx&action=default) identifies standards and best practices that are relevant for the "*AI software lifecycle specification*.*"* The following sections discuss how the different regulatory aspects relate to the TG-Dental.

## Existing applicable regulatory frameworks

Most of the AI systems that are part of the FG-AI4H benchmarking process can be classified as *software as medical device* (SaMD) and eligible for a multitude of regulatory frameworks that are already in place. In addition, these AI systems often process sensitive personal health information that is controlled by another set of regulatory frameworks. The following section summarizes the most important aspects that AI manufacturers need to address if they are developing AI systems for Dental Diagnostics and Digital Dentistry.

* What existing regulatory frameworks cover the type of AI in this TDD (e.g., MDR, FDA, GDPR, and ISO; maybe the systems in this topic group always require at least "MDR class 2b" or maybe they are not considered a medical device)?
* Are there any aspects to this AI system that require additional specific regulatory considerations?

## Regulatory features to be reported by benchmarking participants

In most countries, benchmarked AI solutions can only be used legally if they comply with the respective regulatory frameworks for the application context. This section outlines the compliance features and certifications that the benchmarking participants need to provide as part of the metadata. It facilitates a screening of the AI benchmarking results for special requirements (e.g., the prediction of prediabetes in a certain subpopulation in a country compliant to the particular regional regulatory requirements).

* Which certifications and regulatory framework components of the previous section should be part of the metadata (e.g., as a table with structured selection of the points described in the previous section)?

## Regulatory requirements for the benchmarking systems

The benchmarking system itself needs to comply with regulatory frameworks (e.g., some regulatory frameworks explicitly require that all tools in the quality management are also implemented with a quality management system in place). This section outlines the regulatory requirements for software used for benchmarking in this topic group.

* Which regulatory frameworks apply to the benchmarking system itself?
* Are viable solutions with the necessary certifications already available?
* Could the TG implement such a solution?

## Regulatory approach for the topic group

*Topic Driver: Please select the points relevant for your type of AI and the corresponding benchmarking systems. If your AIs and your benchmarking are not a medical device, this might be quite short.*

Building on the outlined regulatory requirements, this section describes how the topic group plans to address the relevant points in order to be compliant. The discussion here focuses on the guidance and best practice provided by the [DEL02](https://extranet.itu.int/sites/itu-t/focusgroups/ai4h/_layouts/15/WopiFrame.aspx?sourcedoc=%7BF2F46A99-7457-4BC8-81A3-0E1E63D6072A%7D&file=DEL02.docx&action=default) *"AI4H regulatory considerations."*

* Documentation & Transparency
  + How will the development process of the benchmarking be documented in an effective, transparent, and traceable way?
* Risk management & Lifecycle approach
  + How will the risk management be implemented?
  + How is a life cycle approach throughout development and deployment of the benchmarking system structured?
* Data quality
  + How is the test data quality ensured (e.g., the process of harmonizing data of different sources, standards, and formats into a single dataset may cause bias, missing values, outliers, and errors)?
  + How are the corresponding processes document?
* Intended Use & Analytical and Clinical Validation
  + How are technical and clinical validation steps (as part of the lifecycle) ensured (e.g., as proposed in the IMDRF clinical evaluation framework)?
* Data Protection & Information Privacy
  + How is data privacy in the context of data protection regulations ensured, considering regional differences (e.g., securing large data sets against unauthorized access, collection, storage, management, transport, analysis, and destruction)? This is especially relevant if real patient data is used for the benchmarking.
* Engagement & Collaboration
  + How is stakeholder (regulators, developers, healthcare policymakers) feedback on the benchmarking collected, documented, and implemented?

## Dentition Stages

The project of teeth identification, can be divided in three separate Phases: **Permanent Dentition; Milk Dentition and Mixed Dentition**.

At this time, only panoramic radiograph from patients with Permanent Dentition shall be included and that's because it will be an enormous task to identify any teeth and if we try to identify, besides the permanent teeth, the milk teeth and germs too, that shall be very difficult for the training set and the model.

So, in Phase One - **Permanent Dentition**, we focus on Permanent teeth identification; in Phase Two - **Milk Dentition,** once we have already been able to identify the permanent teeth, so we can focus on milk teeth identification; and in Phase Three - **Mixed Dentition**, once we have already been able to identify the permanent and the milk teeth, so we can focus on the germs identification.

## Imaging Acquisition Format

Regarding the radiographic or tomographic or photographic image acquisition, I suggest to use the International standard DICOM (Digital Imaging and Communications in Medicine) for the communication and management of medical imaging information and related data, from NEMA (National Electrical Manufacturers Association). The use of DICOM is a safety guarantee for the image and the information embedded in the "dcm'' file. It shall be noted that all major x-ray equipment manufacturers in the dental market, such as Instrumentarium, Kavo, Soredex, Carestream Dental, Plameca, Vatech, NewTom and others are already DICOM compatible, directly or indirectly (with a bridge software).

## Anonymization or deidentification

Regarding the anonymization or deidentification of the DCIOM files from each image, it can be used with a free Python Library – Pydicom (<https://pydicom.github.io/> ). It is a very nice tool for that purpose, it can be directed to each of the many DICOM tags (ex. Name, ID, Gender, Date of Birth, Referring Dentist, Modality, etc). After this step, anonymization, the image can be converted to any image file format, such as "jpeg", "png", "bmp" or nay other, for the training set or the model. ( \* This is a technical issue that, probably, shall not be pointed out by a dentist like myself, so this is just a mention!)

## Direct DICOM-to-CNN

Recently Dr.Salim Kanoun and Wendy Revailler, have developed a Python Library, DICOM-to-CNN (<https://github.com/salimkanoun/Dicom-To-CNN> -<https://dicom-to-cnn.readthedocs.io/en/latest/> ) that allows to process DICOM files directly for training and inference purpose, without the conversion step. (\* This is a technical issue that, probably, shall not be pointed out by a dentist like myself, so this is just a mention!)

# References

*Topic driver: Add the bibliography here.*

*Topic driver: If you include figures in this document, please use the following MS Word format/style (otherwise the figure won't be included in the table of figures).*

# Annex A: Glossary

This section lists all the relevant abbreviations, acronyms and uncommon terms used in the document.

|  |  |  |
| --- | --- | --- |
| **Acronym/Term** | **Expansion** | **Comment** |
| TDD | Topic Description Document | Document specifying the standardized benchmarking for a topic on which the FG AI4H Topic Group works. This document is the TDD for the Topic Group on dental diagnostics and digital dentistry (TG-Dental) |
| TG | Topic Group |  |
| WG | Working Group |  |
| FGAI4H | Focus Group on AI for Health |  |
| AI | Artificial intelligence |  |
| ITU | International Telecommunication Union |  |
| WHO | World Health Organization |  |
| DEL | Deliverable |  |
| CfTGP | Call for topic group participation |  |
| AI4H | Artificial intelligence for health |  |
| IMDRF | International Medical Device Regulators Forum |  |
| MDR | Medical Device Regulation |  |
| ISO | International Standardization Organization |  |
| GDPR | General Data Protection Regulation |  |
| FDA | Food and Drug administration |  |
| SaMD | Software as a medical device |  |
| AI-MD | AI based medical device |  |
| LMIC | Low-and middle-income countries |  |
| GDP | Gross domestic product |  |
| API | Application programming interface |  |
| IP | Intellectual property |  |
| PII | Personal identifiable information |  |
| […] |  |  |

# Annex B: Declaration of conflict of interests

In accordance with the ITU transparency rules, this section lists the conflict-of-interest declarations for everyone who contributed to this document. Please see the guidelines in FGAI4H-F-105 "ToRs for the WG-Experts and call for experts" and the respective forms (Application form & Conflict of interest form).

**Company/Institution/Individual XYZ**

A short explanation of the company's area of activity and how the work on this document might benefit the company and/or harm competitors. A list of all people who contributed to this document on behalf of this company and any personal interest in this company (e.g., shares).

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