

#### Comparative analysis of the diagnostic accuracy of an artificial intelligence software and dental practitioners in assessing caries (using the CPO-D index) and periapical lesions on panoramic radiographs

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#### Introduction

#### From screening



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Determine the sensitivity, specificity, predictive values, accuracy, F1 score and area under the ROC curve (AUC) of **WeDiagnostix** in its Sensitive (WD-S) and Optimal (WD-O) modes. Compare the performance of each mode with that of three dentists with ≥ 5 years of clinical practice. Assess inter-observer agreement (Cohen's κ) among the clinicians.

Dental caries remains the most prevalent noncommunicable disease worldwide, affecting approximately 2.5 billion individuals (Kassebaum et al., 2015 ; Guerreiro et al., 2024). Human interpretation of orthopantomograms (OPGs) shows considerable inter-observer variability, with k values rarely exceeding 0.60 (Landis & Koch, 1977). Convolutional neural networks currently achieve receiver-operating-characteristic (ROC) areaunder-the-curve (AUC) values above 0.90 in dentistry (Schwendicke et al., 2019). However, gaps persist regarding sample representativeness and algorithmic bias (Ezhov et al., 2021). The equivalence—or superiority—of artificial intelligence (AI) compared with experienced

### to confirmation:

Al modes WD-S

### & WD-O match

dentists'

diagnostic

accuracy

# Results and Conclusions

Table 1: Main results of the different modes compared with the established gold standard.

Lesion Mode Sensitivity Cific F1 AUC

observers still needs to be confirmed in a Portuguese sample.

## Materials and Methods

**Study design:** Observational, comparative study evaluating—on panoramic radiographs—the diagnostic accuracy for dental caries (CPO-D index) and apical lesions achieved by an artificialintelligence (AI) software versus dentists with more than five years' experience. Lesion **Sample:** 200 adult OPGs (September–December 2024), quality Grade 1, selected from 780 examinations performed at the Egas Moniz High University Clinic. confidence Gold standard: Consensus of the three evaluators ( $\kappa = 0.75$ , "substantial"). **Al system:** WeDiagnostix v23.10 operated in two modes—WD-S (sensitive) and WD-O (optimal). **Dentist's** final decision



	Lesion	Mode	Sensitivity	cific ity	F1	AUC
	Caries	WD-S	85,3 %	88,7 %	0,49	0,870
	Caries	WD-O	55,5 %	<b>99,0</b> %	0,66	0,773
	Apical Lesion	WD-S	<b>88,2</b> %	95,9 %	0,31	0,920
	Apical Lesion	WD-O	75,0 %	<b>99,2</b> %	0,60	0,871
The findings align with the meta-analysis by Pu						
and Schwendicke (2024), which reports an						
overall sensitivity of 94 % and a specificity of 9						
%. WD-S maximises sensitivity and is therefore						
recommended for screening, whereas WD-O						
prioritises specificity, making it suitable for						

diagnostic confirmation.

Limitations include the monocentric and

2nd Screening WD-O High specificity st's I on

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