



Artificial Intelligence-Driven CBCT Analysis for Segmentation of Inferior Alveolar Nerve for Standardized Interpretation and Enhanced Patient Safety

Negin Soghli¹, Aminollah Khormali², Patricia A. Miguez³

¹ Department of Biomedical Sciences, Adams School of Dentistry, The University of North Carolina at Chapel Hill

² Adams School of Dentistry, The University of North Carolina at Chapel Hill

³ Department of Periodontology, Endodontics and Dental Hygiene, Adams School of Dentistry, The University of North Carolina at Chapel Hill

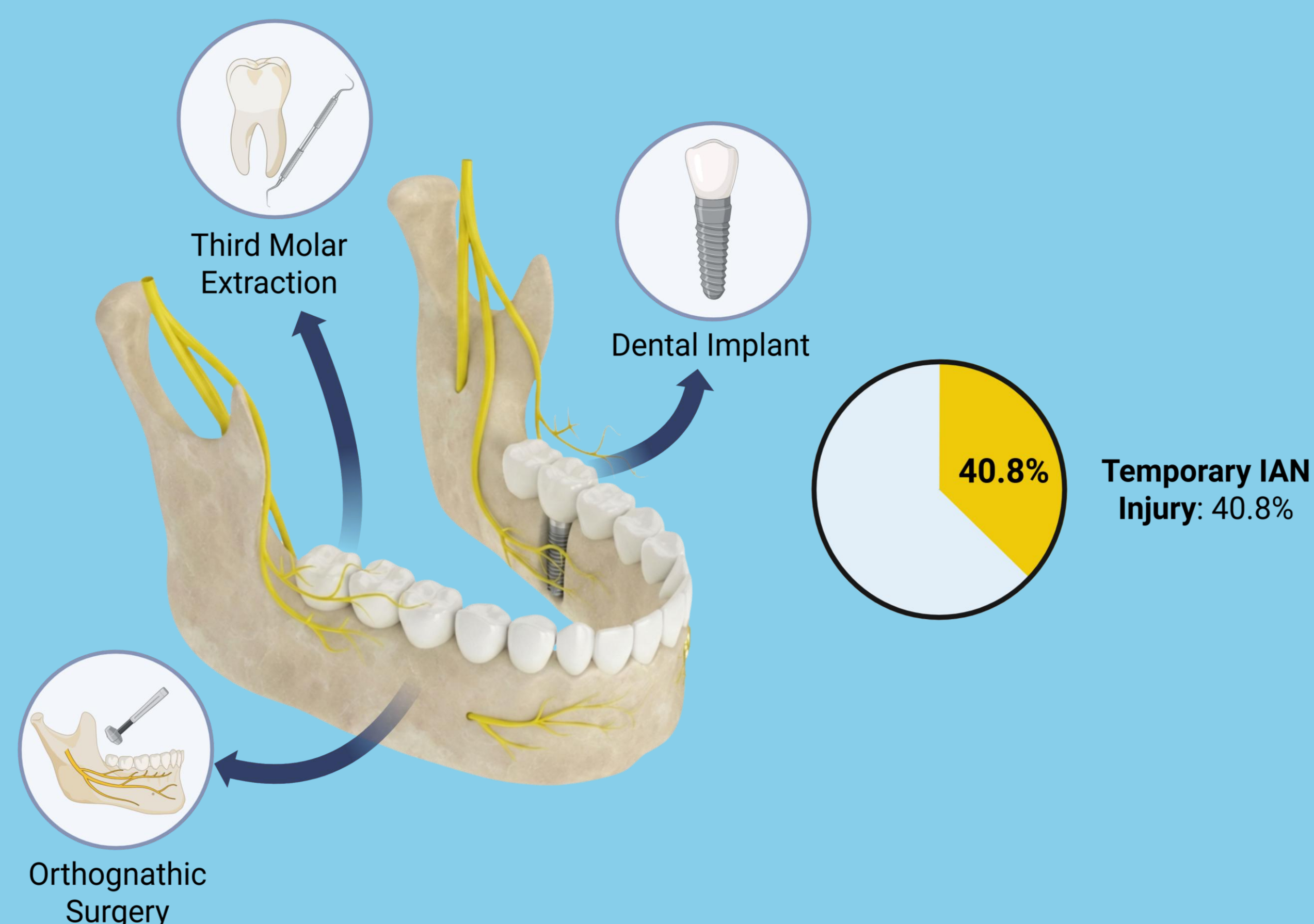


Backgrounds & Clinical Needs:



Clinical Problem:

- **Inferior alveolar nerve (IAN) injury** can occur during:
 - Third molar extraction
 - Mandibular implant placement
 - Orthognathic surgery
- **Consequences:**
 - Permanent paresthesia or dysesthesia
 - Chronic pain
 - Significant medico-legal risk



Motivation:

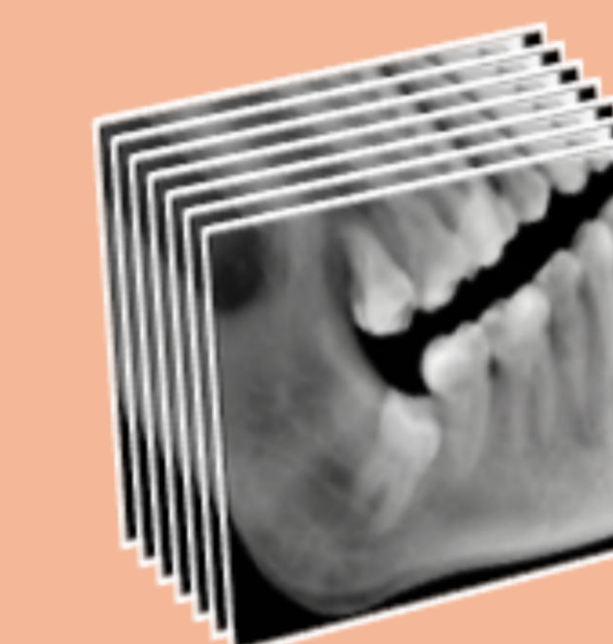
- **Manual tracing:**
 - Technically difficult
 - Time Consuming
 - Highly operator-dependent
- Low contrast between IAC and trabecular bone
- Metal artifact obscure canal borders



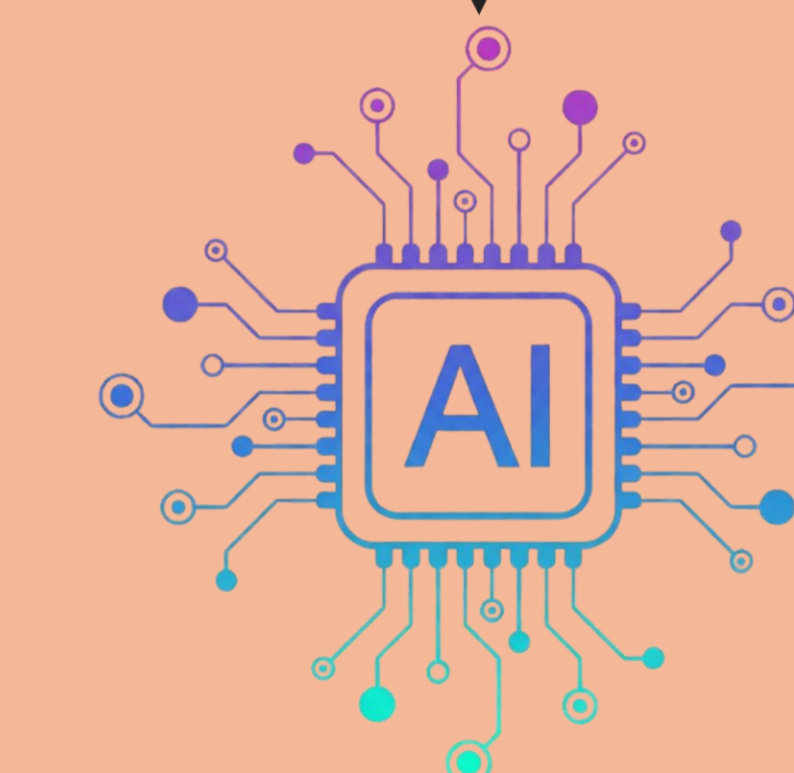
Study Aim:

- To develop and validate an advanced deep learning model for **automated IAN segmentation**
- Evaluate its **clinical impact** on inter-operator agreement and planning time across different expertise levels

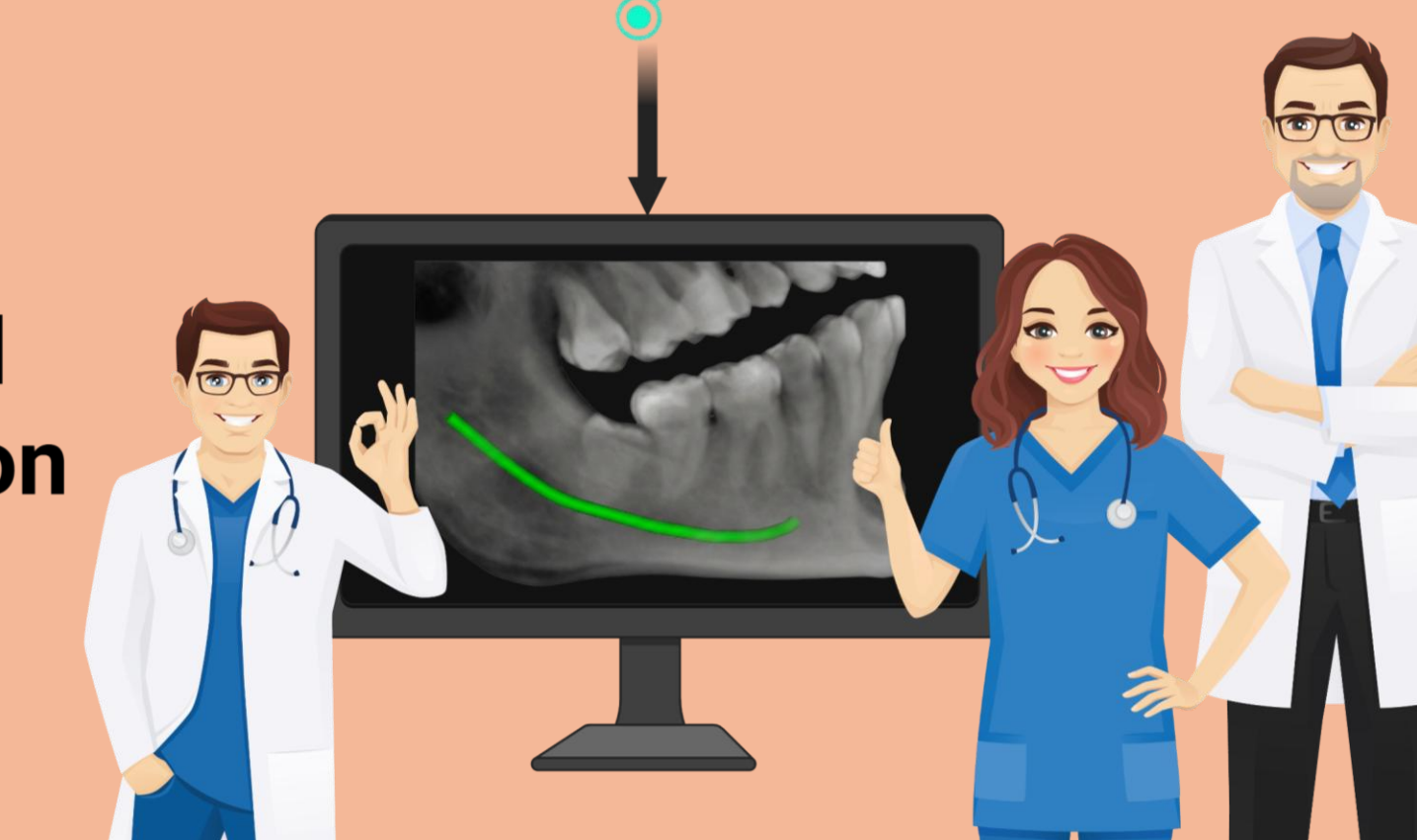
Cone Beam CT Image



Artificial Neural Network



Automated Segmentation





Materials and Methods:

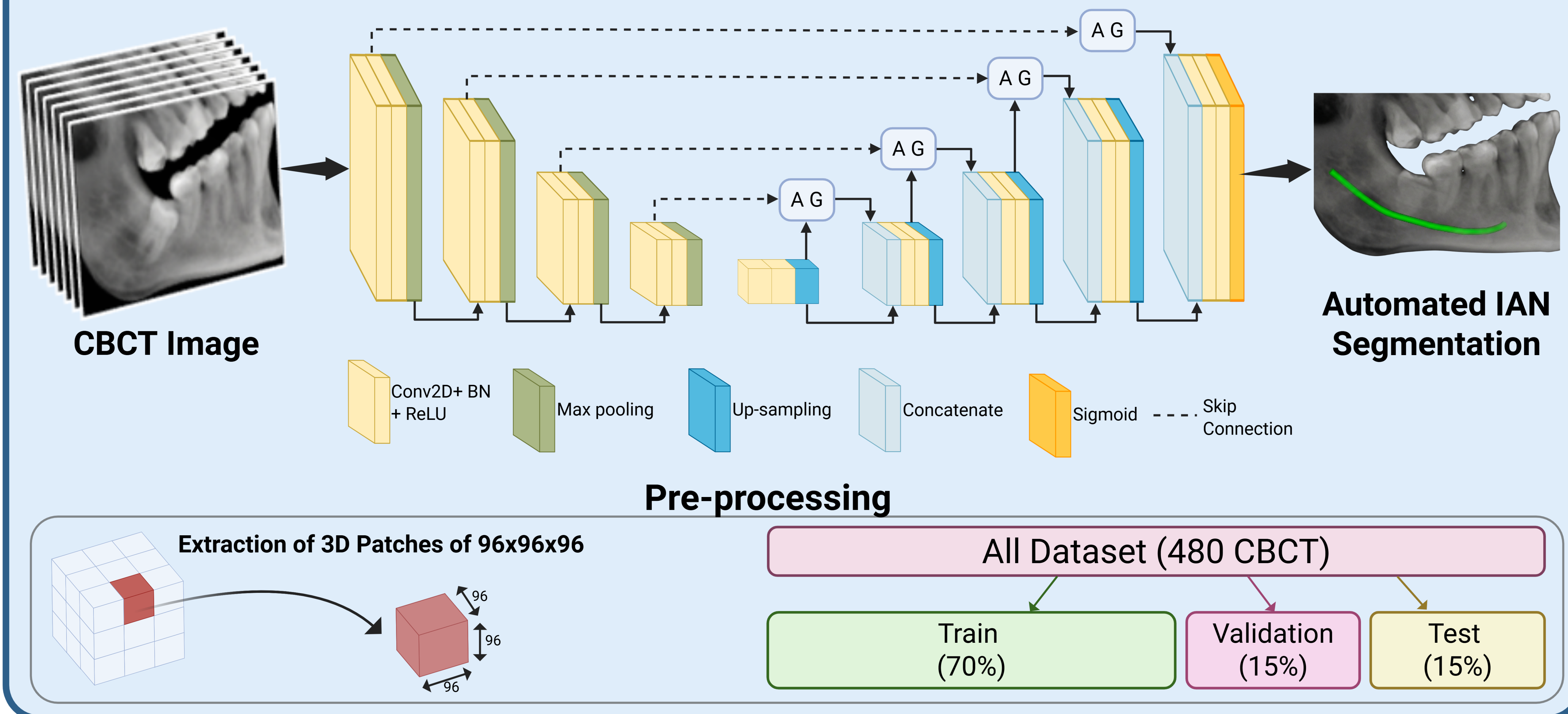
Dataset:

- Tooth Fairy 2 dataset (480 expertly annotated CBCT volumes)
- Diverse anatomy, image quality, and metal artifacts
- Standardized voxel spacing (0.117 mm³)

Model Architecture:

- Gated-Attention U-Net architecture
- Multi-scale feature fusion
- Metrics: Dice similarity coefficient (DSC), IoU

Automated Gated-Attention U-Net Architecture

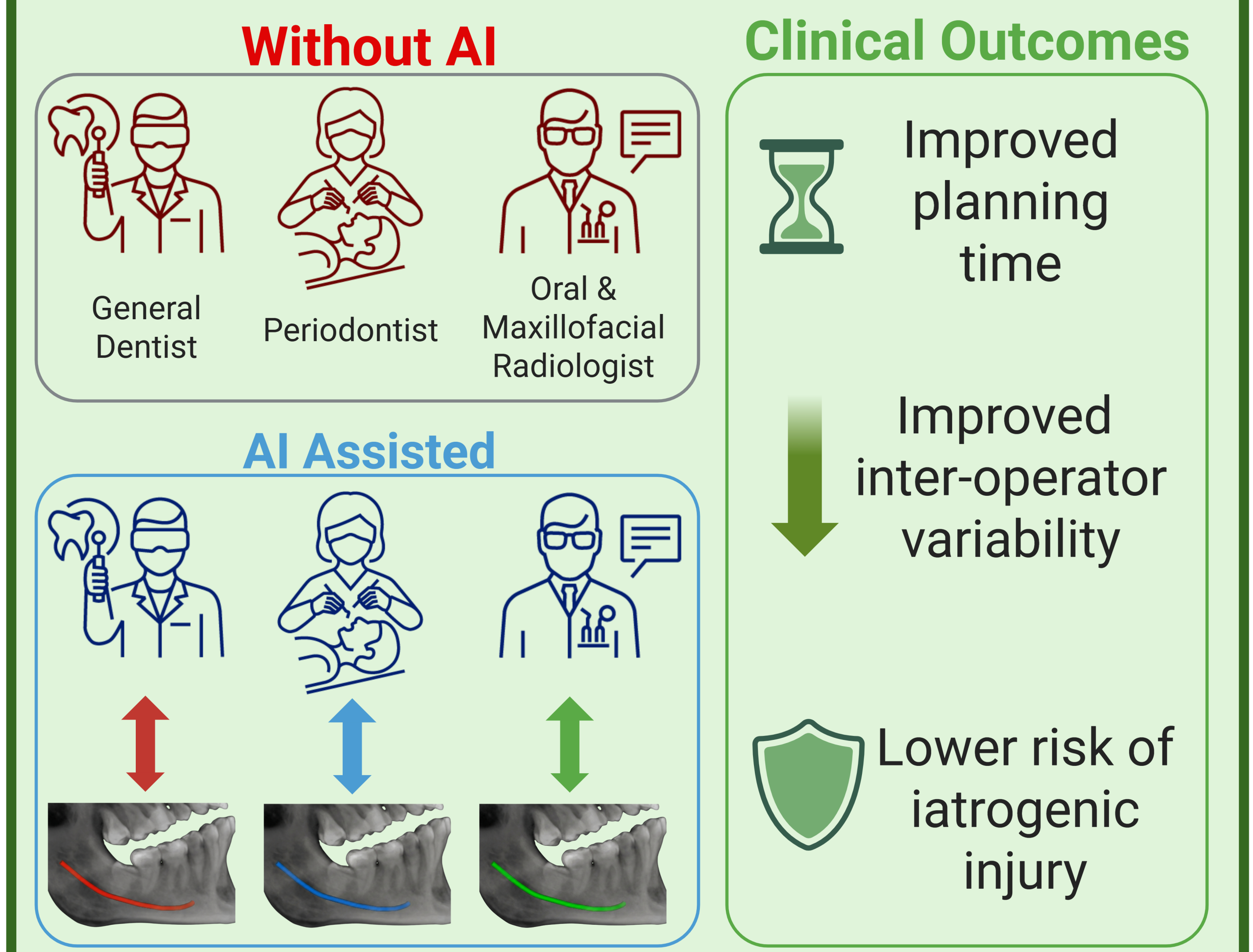


Clinical Validation (Ongoing):

Prospective multi-reader study

- 9 clinicians:
 - 3 General Dentists
 - 3 Periodontology Residents
 - 3 Radiology Residents

Bridging the Expertise Gap



Results and Comparisons:



Comparison with state of the art:

- High-Precision nerve segmentation with clinical relevance
- Outperformed standard U-Net and attention variants
- Maintained continuity near mental foramen

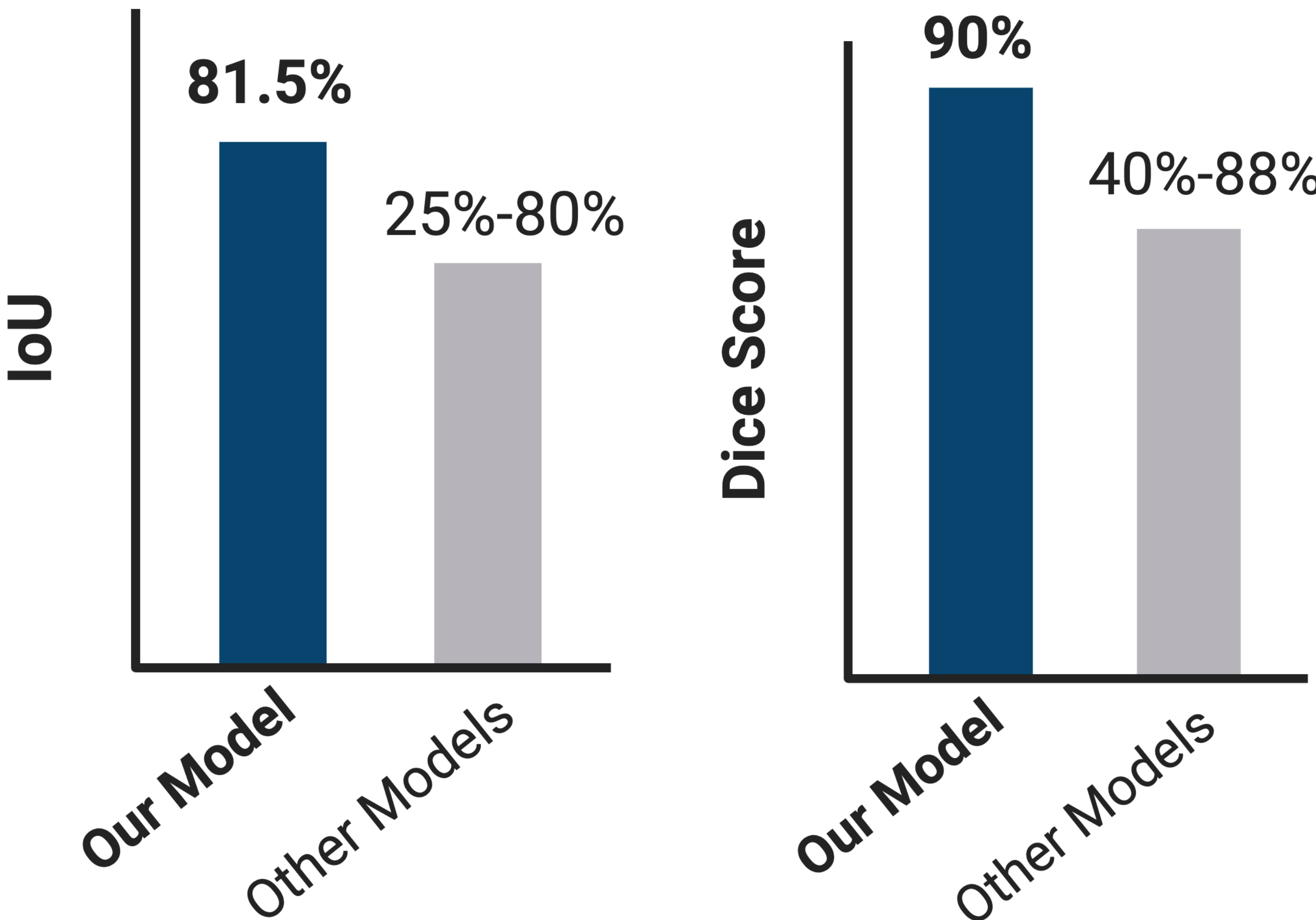
Clinical Performance:

- Reduced planning time
- Maintained high precision in low-contrast regions
- Robust to metal artifacts and thin cortical boundaries

Study	Model/ Method	Dice (%)	IoU (%)
Zhou et al., 2018	U-Net++	88.70 ± 6.16	80.13 ± 8.29
Oktay et al., 2018	Attention U-Net	85.06 ± 8.35	74.72 ± 10.10
Kwak et al., 2020	U-Net	87.98 ± 6.37	79.01 ± 8.45
Jaskari et al., 2020	Jaskari’s model	87.85 ± 6.35	78.79 ± 8.34
Hatamizadeh et al., 2021	SwinUNETR	40.47 ± 11.36	25.98 ± 8.83
Hatamizadeh et al., 2022	UNETR	73.84 ± 11.21	59.62 ± 12.69
Liu et al., 2024	Liu’s model	88.37 ± 4.90	79.45 ± 6.60
Our Model	Gated Attention U-Net	89.72 ± 4.28	81.47 ± 5.34

Comparative Performance:

- Superior to:
- U-Net
 - U-Net++
 - Attention U-Net
 - Transformer Models (UNETR, SwinUNETR):





Discussion and Conclusion:



Toward safer and more standardized mandibular surgeries:

Gated attention deep learning enables:

- Accurate IAN segmentation
- Reliable nerve continuity
- Explainable uncertainty visualization

AI assistance:

- Reduces planning time (bridging the expertise gap)
- Improves agreement across various experience levels

Clinical Significance:

- **Supports safer:**
 - Implant placement
 - Third molar extraction
 - Mandibular surgeries
- Provides **explainable AI** features through probability maps

Future directions:

- Integration into implant planning software
- Expansion to diverse CBCT platforms

AI –assisted IAN visualization leads to standardized care, reduced variability, and lower risk of avoidable nerve injury!

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Negin Soghli
DDS, PhD Candidate



Aminollah Khormali
PhD, DDS Candidate



Patricia A. Miguez
DDS, MS, PhD