

Medication-Safe Dual-Encoder Retrieval for Structured Dental Report Prediction from CBCT

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Seven fields.
One traceable source per field.

Retrieve observed clinical text from just 50 labeled reports, then constrain branch selection by medication mentions.



CODE · MIT LICENSE

50 labeled
CBCT-report pairs

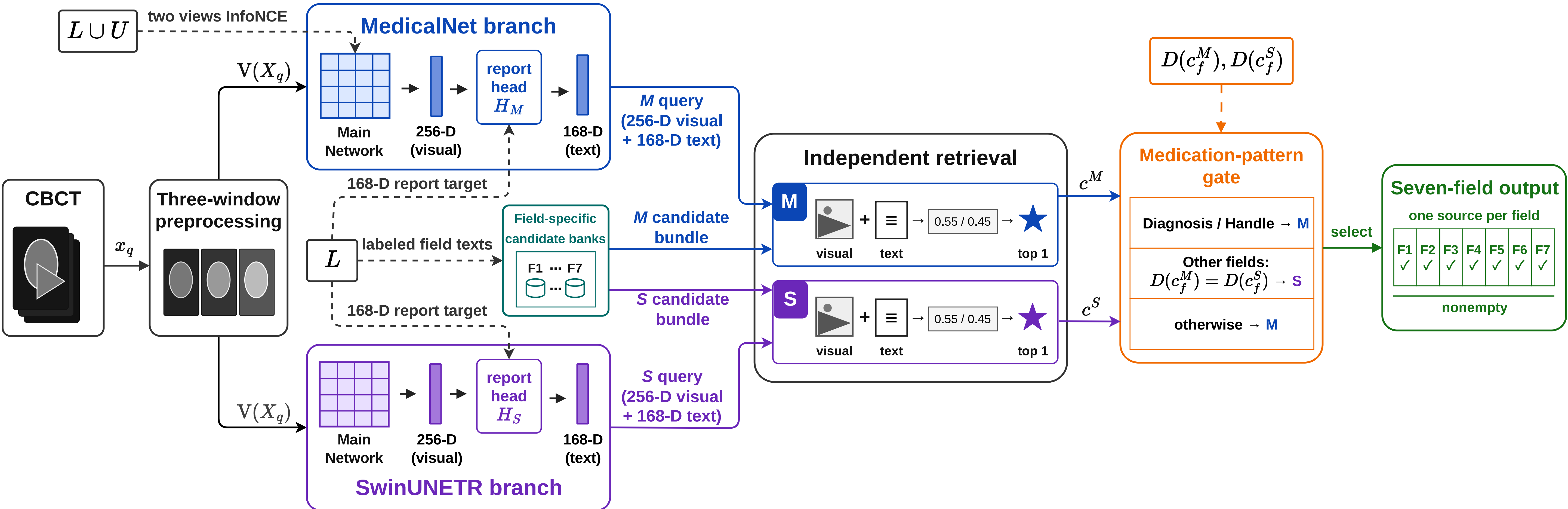
200 unlabeled
training CBCTs

50 validation
cases; reports hidden

The challenge: predict a structured clinical record from a CBCT query alone. [1]

Seven output fields: Main appeal · Present medical history · Oral Check · Diagnosis · Treatment plan · Handle · Doctor advices

01 Two independent retrievers. One constrained decision.



Method overview. L: labeled cases; U: unlabeled scans. Each branch independently scores visual and predicted-report similarity. Diagnosis and Handle use MedicalNet; other fields use Swin only if the recognized medication-pattern sets match.

Adapt the image encoder
Two-view InfoNCE adapts **MedicalNet only** on all 250 training CBCTs. The Swin encoder starts from public self-supervised weights. [2, 3]

Learn seven report spaces
The 50 reports define field-specific TF-IDF/SVD targets. Two **independent heads** predict 168-D report vectors using field-wise cosine loss.

Retrieve, then apply the gate
Banks store source IDs, field texts, text targets, and branch-specific visual vectors. Each top-1 uses **0.55 visual + 0.45 report** similarity.

02 Official results

Codabench evaluation · 50 cases [4]

0.235259 **100%**
Weighted Score Field completion

0 missing predictions · 350 nonempty fields

Diagnosis-code F1	0.161
Treatment-action F1	0.347
Medication F1	0.470
Treatment-plan recall	0.579
Tooth-notation F1	0.109
ROUGE-L	0.147

Tooth precision / recall: **0.066 / 0.453**.
Completion does not imply correctness; these results do not establish superiority over a matched baseline.

1.55 s / case
77.49 s for 50 cases; 7,273 MiB sampled peak.
Local RTX 4090 container run, not organizer latency.

Fine-tuning epochs: MedicalNet 300 / Swin 60.

03 Why the gate keeps MedicalNet

A recorded example · Query 552 · Oral Check

Query
552

MedicalNet source
391

Swin source
262

Central axial views of the existing preprocessed volumes; no lesion annotations.

MedicalNet candidate excerpt
 $D(c_M) = \{\text{iodophor}\}$
“26 27 **iodophor** gauze and sutures are stored, and the wound heals well”

Swin candidate excerpt
 $D(c_S) = \emptyset$
“17 The temporary inlay fell off.”

Unequal medication-pattern sets → retain MedicalNet.
The selected field keeps the observed “iodophor” mention.

$D(c)$ is the set of recognized medication patterns in the full candidate. Excerpts are shortened for display. This audit verifies execution and provenance, **not clinical correctness**.

All 350 field decisions, reconstructed from frozen models		
231 Swin equal sets	19 MedicalNet unequal sets	100 MedicalNet Diagnosis / Handle

Traceable text is not a clinical safety guarantee.

“Medication-safe” names a **lexical gate**, not a clinical claim. It checks recognized patterns—not indication, dose, route, negation, or correctness. Selection precedes deterministic shortening.

A bank of 50 reports limits coverage and may copy mismatched details. Fields can come from different patients. **Next:** controlled comparisons, calibrated abstention, and tooth-aware reranking.