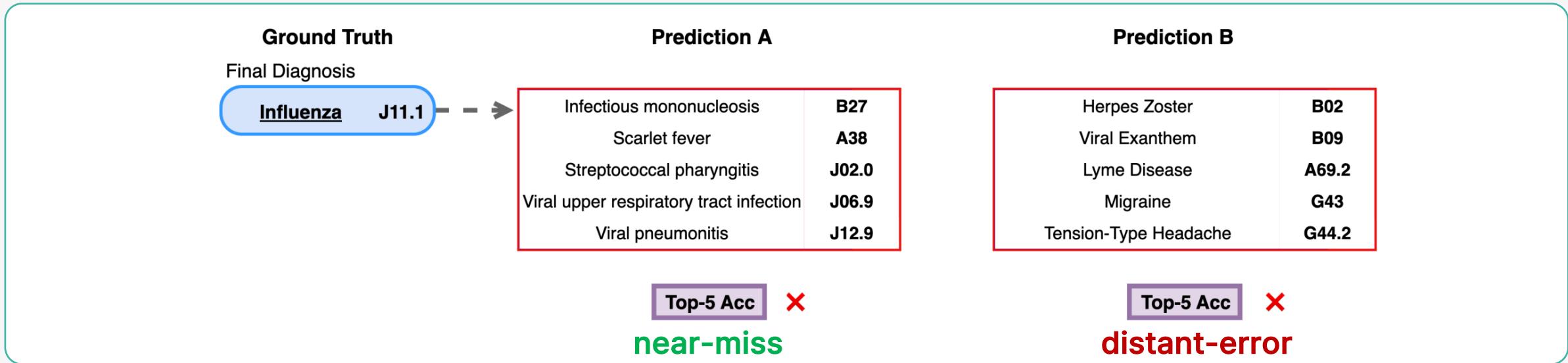

H-DDx: A Hierarchical Evaluation Framework for Differential Diagnosis

Seungseop Lim*, Gibaeg Kim*, Hyunkyung Lee, Wooseok Han, Jean Seo, Jaehyo Yoo, Eunho Yang



Are All Diagnostic Errors Equal?



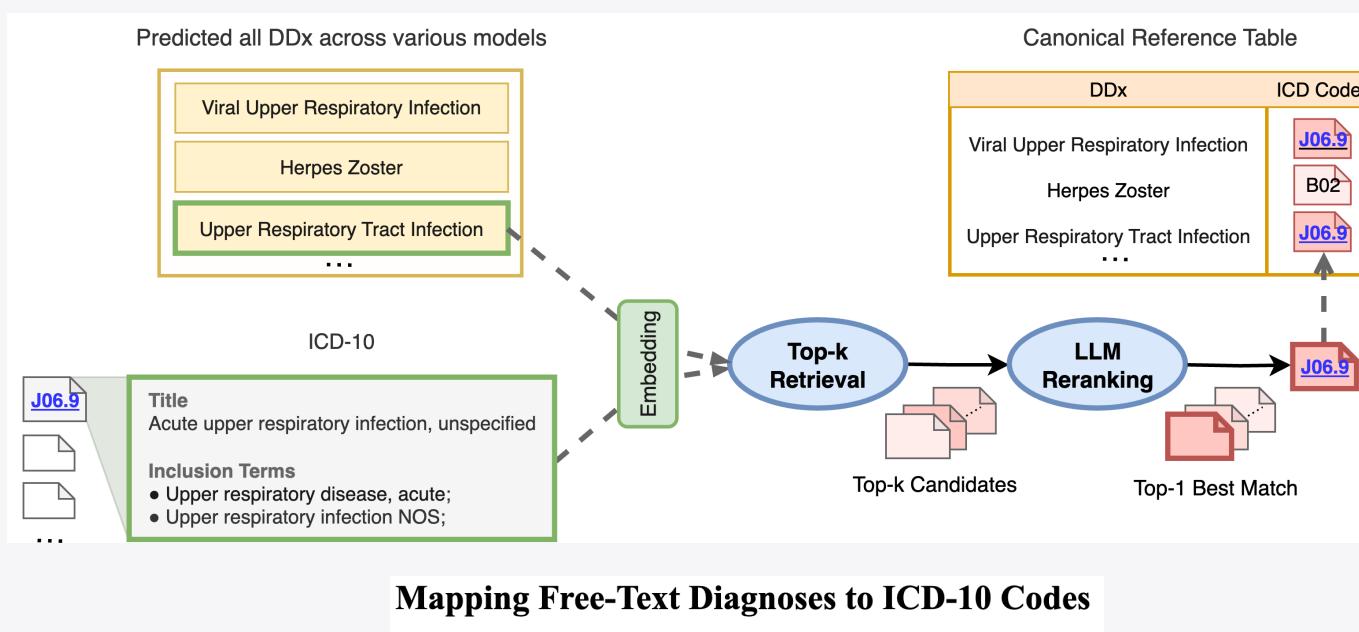
Limitations of Existing Evaluation Methods

- Dependence on flat metrics (e.g., Top-k Accuracy).
- Limited to a binary determination of diagnosis inclusion.
- Verification relies heavily on LLM-based evaluators.

The H-DDx Framework

Our framework consists of two steps:

(i) a mapping pipeline that combines embedding-based retrieval with LLM reranking for high-accuracy conversion from free-text diagnoses to the ICD-10 codes.



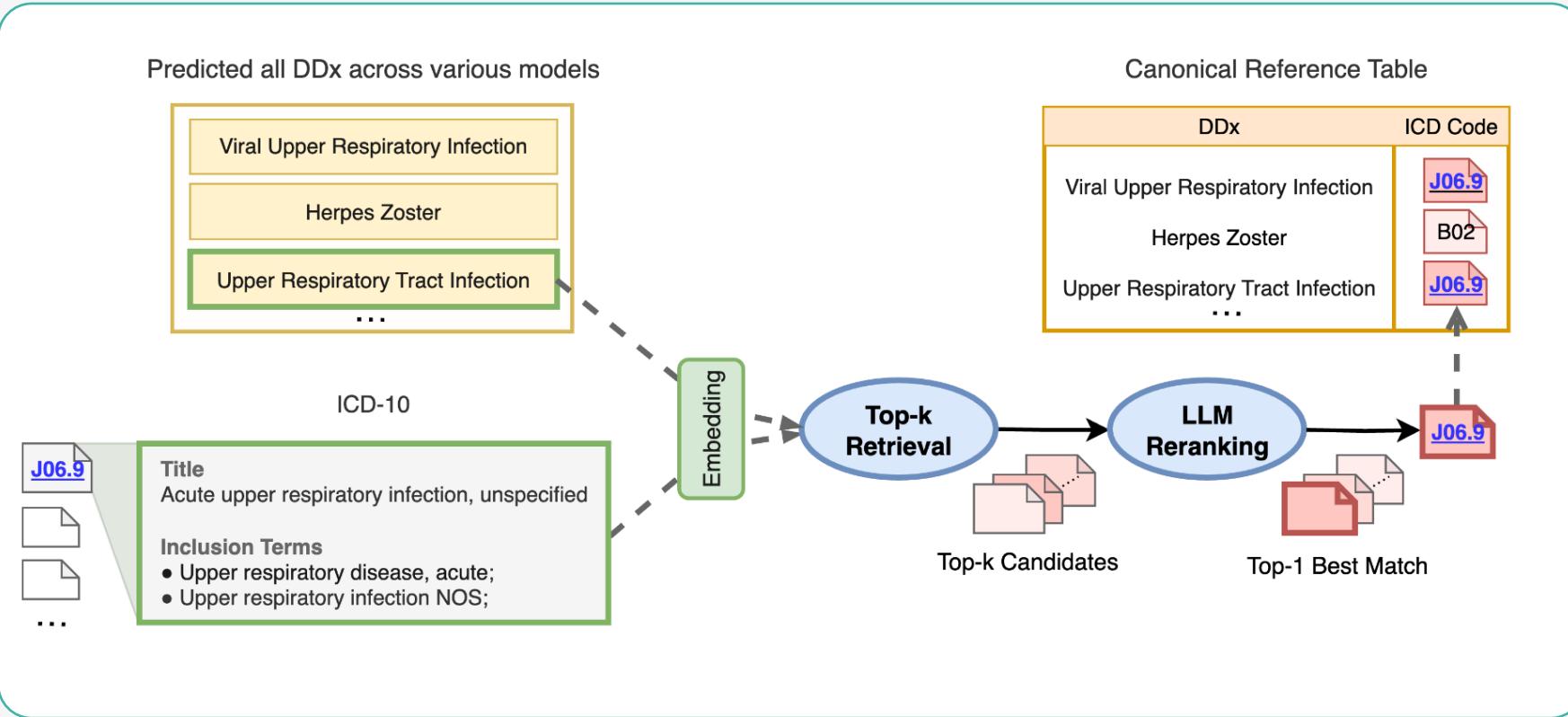
(ii) the Hierarchical DDX F1 (HDF1) metric that performs set-based comparisons using ancestral code expansion within the ICD-10 taxonomy.

$$\text{HDP} = \frac{1}{N} \sum_{i=1}^N \frac{|C_i \cap \hat{C}_i|}{|\hat{C}_i|}$$

$$\text{HDR} = \frac{1}{N} \sum_{i=1}^N \frac{|C_i \cap \hat{C}_i|}{|C_i|}$$

$$\text{HDF1} = \frac{2 \times \text{HDP} \times \text{HDR}}{\text{HDP} + \text{HDR}}$$

Mapping Pipeline

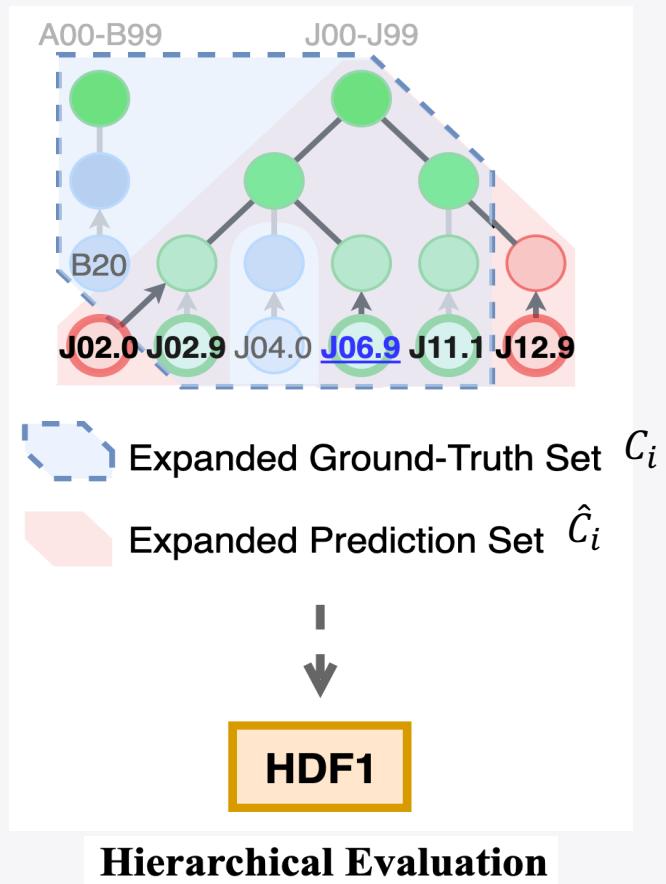


LLM-generated diagnosis	ICD-10 Code
Viral Upper Respiratory Infection	J06.9
Upper Respiratory Tract Infection	J06.9

Model	Top-1
Top-k Retrieval	biobert-v1.1
	BioSimCSE-BioLinkBERT-BASE
	Qwen3-Embedding-0.6B
	pubmedbert-base-embeddings
	text-embedding-3-large
LLM Reranking (on best retriever)	+ gemini-2.5-flash-lite
	+ gpt-4o-mini
	+ gemini-2.5-flash
	+ gpt-4o

Hierarchical Metric (HDF1)

The Hierarchical DDX F1 (HDF1) metric that performs set-based comparisons using ancestral code expansion within the ICD-10 taxonomy.



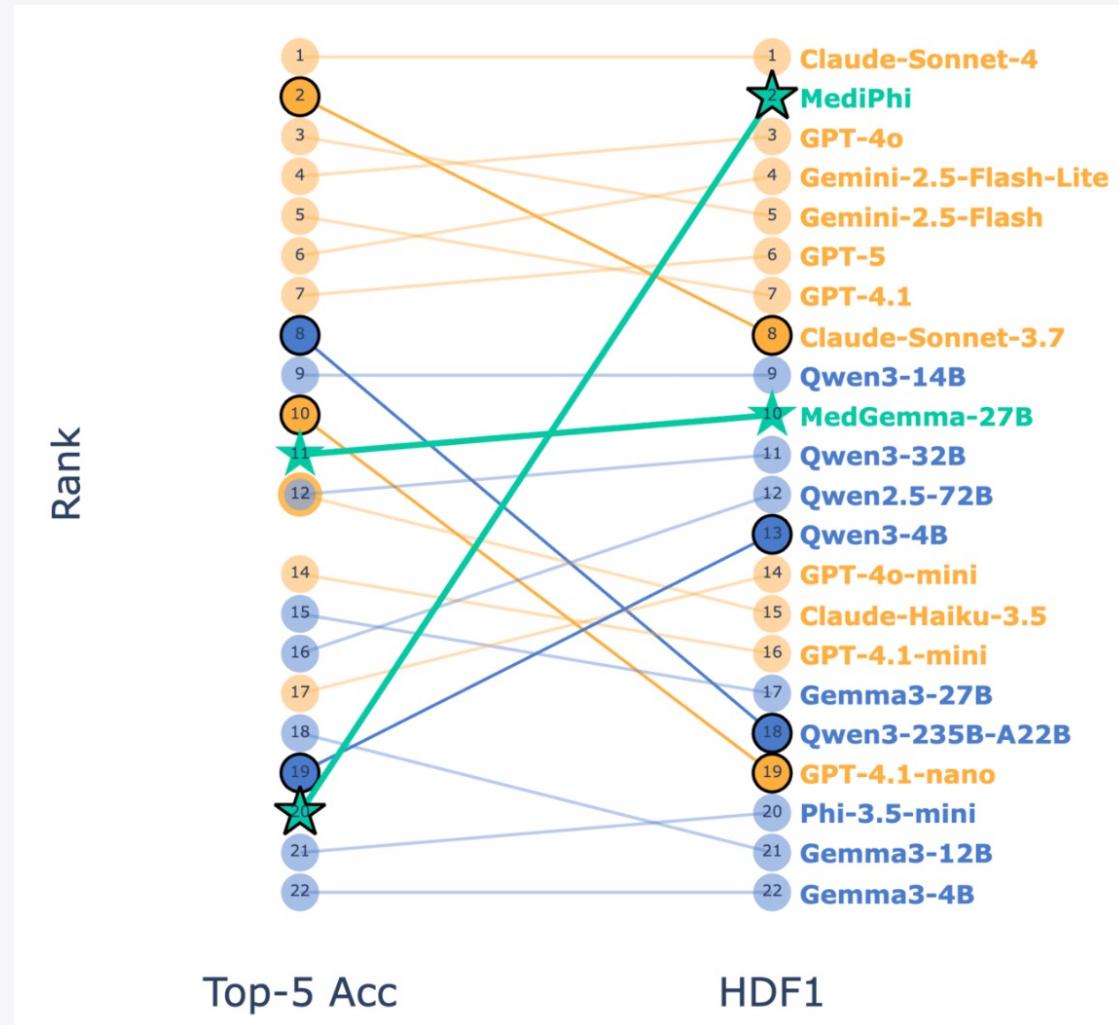
$$\text{HDP} = \frac{1}{N} \sum_{i=1}^N \frac{|C_i \cap \hat{C}_i|}{|\hat{C}_i|}$$

$$\text{HDR} = \frac{1}{N} \sum_{i=1}^N \frac{|C_i \cap \hat{C}_i|}{|C_i|}$$

$$\text{HDF1} = \frac{2 \times \text{HDP} \times \text{HDR}}{\text{HDP} + \text{HDR}}$$

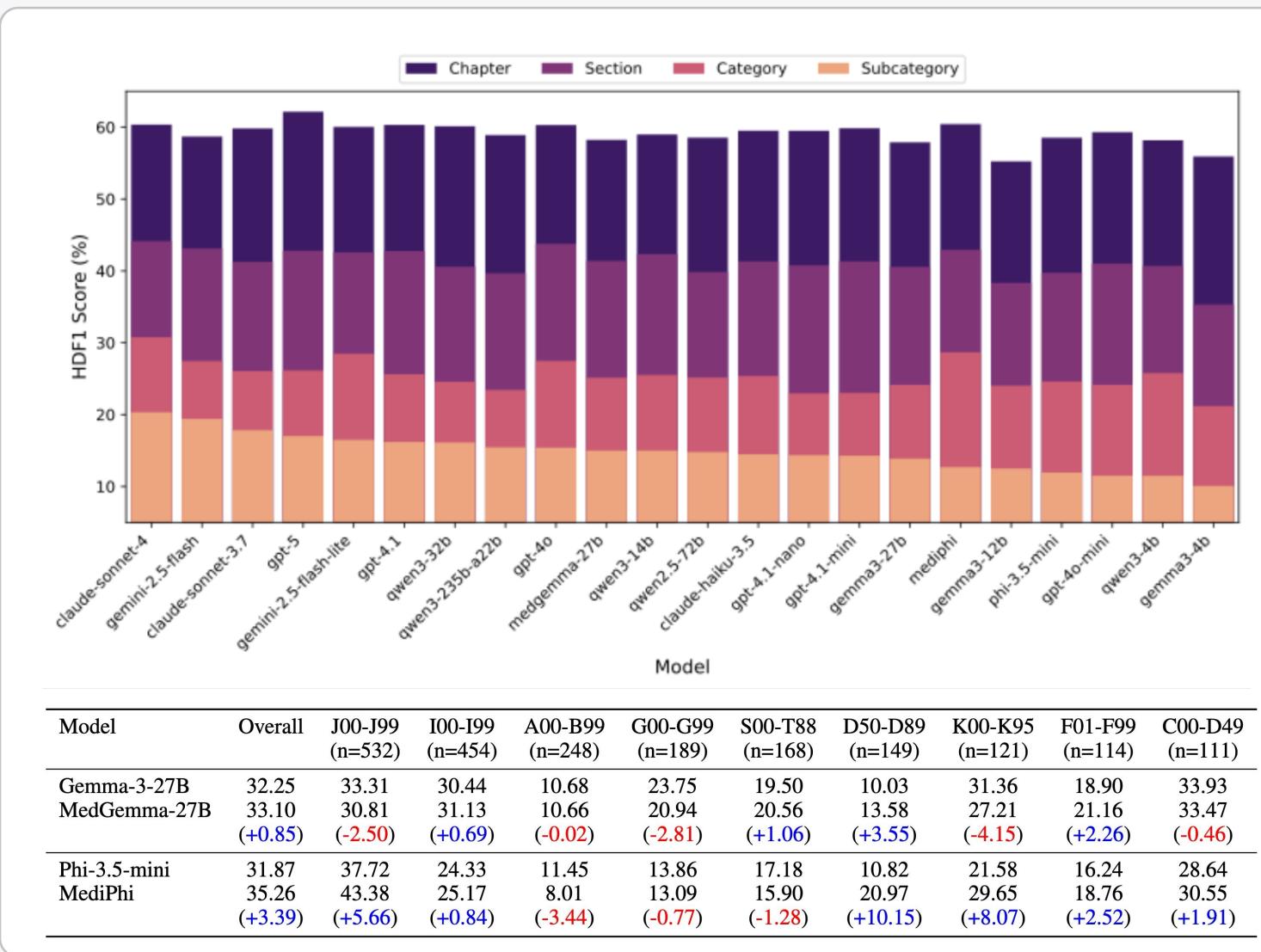
Ranking Shift

Model	Top-5	HDF1	ΔRank
Proprietary Models			
Claude-Haiku-3.5	0.7630	0.3237	↓ 3
Claude-Sonnet-3.7	<u>0.8360</u>	0.3380	↓ 6
Claude-Sonnet-4	0.8390	0.3673	—
Gemini-2.5-Flash-Lite	0.7890	0.3496	↑ 2
Gemini-2.5-Flash	0.8320	0.3483	↓ 2
GPT-4o-mini	0.7240	0.3276	↑ 3
GPT-4o	0.8040	0.3499	↑ 1
GPT-4.1-nano	0.7660	0.3213	↓ 9
GPT-4.1-mini	0.7590	0.3232	↓ 2
GPT-4.1	0.8010	0.3387	↓ 2
GPT-5	0.7830	0.3448	↑ 1
Open-source Models			
Phi-3.5-mini	0.6550	0.3187	↑ 1
Gemma3-4B	0.6080	0.2891	—
Gemma3-12B	0.7180	0.3075	↓ 3
Gemma3-27B	0.7460	0.3225	↓ 2
Qwen2.5-72B	0.7420	0.3299	↑ 4
Qwen3-4B	0.6720	0.3291	↑ 6
Qwen3-14B	0.7750	0.3367	—
Qwen3-32B	0.7630	0.3300	↑ 2
Qwen3-235B-A22B	0.7770	0.3215	↓ 10
Medical Fine-tuned Models			
MedGemma-27B	0.7650	0.3310	↑ 1
MediPhi	0.6660	<u>0.3526</u>	↑ 18



Hierarchical Error Analysis

Performance Analysis Leveraging ICD-10 Hierarchy



Conclusion & Future Works

Conclusion

- Moves beyond binary accuracy by capturing clinically meaningful, hierarchical correctness.
- Highlights near-miss diagnoses that traditional metrics overlook.

Future Works

- Validation on real-world EHRs
- Integration of richer clinical ontologies (e.g., SNOMED-CT)
- Extension to multi-turn diagnostic agents
- Correlation with physician evaluations

Thank you