

MEDTO: Medical Data to Ontology Matching Using Hybrid Graph Neural Networks

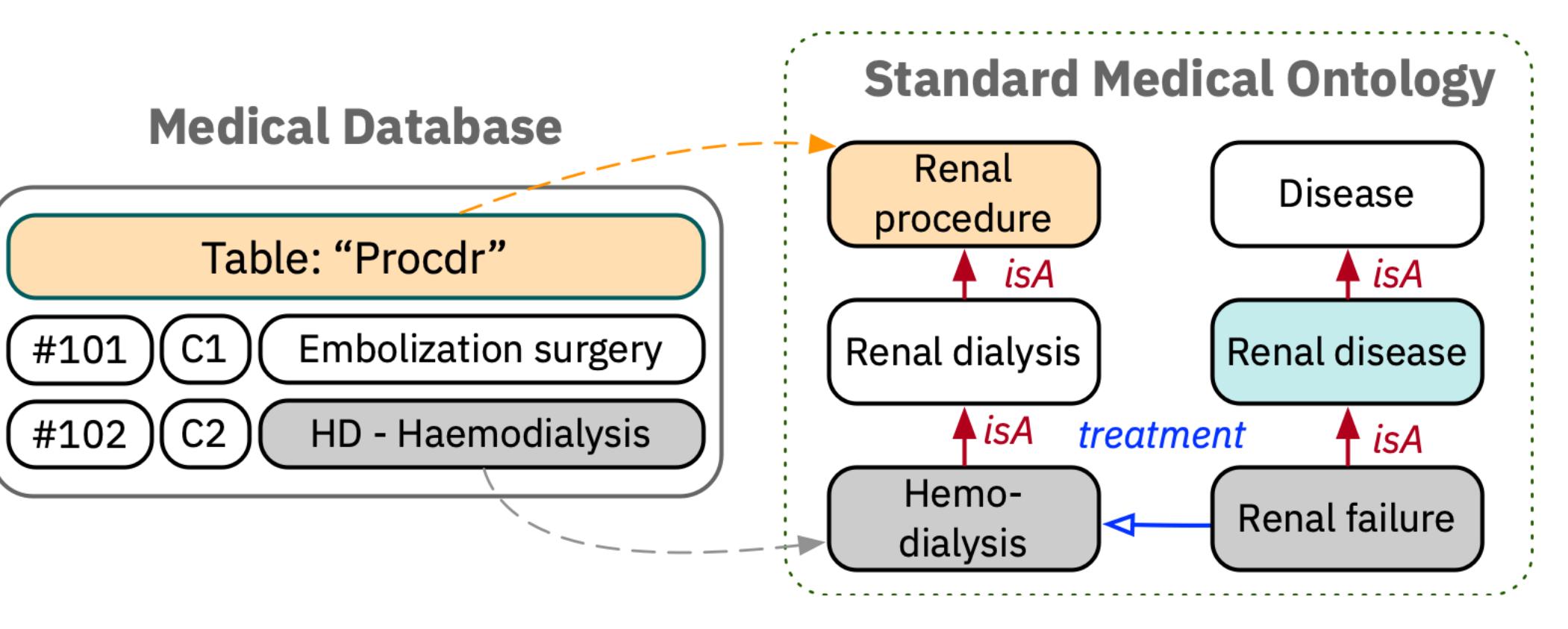
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DATA-TO-ONTOLOGY MATCHING IN MEDICAL DOMAIN

- Increasingly large-scale medical databases, in need of automatic AI-assisted analysis.
- Core task: Mapping database schema/tables to standard ontologies (for standardization)
- Existing methods focus on ontology matching, assuming ontologies are available for matching
- Effective data-to-ontology matching techniques



TRAINING

Matching: MLP, input as pairs of concept embeddings from O_1 and O_2

Training Loss: Matching loss + Graph decoders
 $\mathcal{L} = \mathcal{L}^M + \alpha_1 \cdot (\mathcal{L}_{O_1}^{HYP} + \mathcal{L}_{O_2}^{HYP}) + \alpha_2 \cdot (\mathcal{L}_{O_1}^{HET} + \mathcal{L}_{O_2}^{HET})$

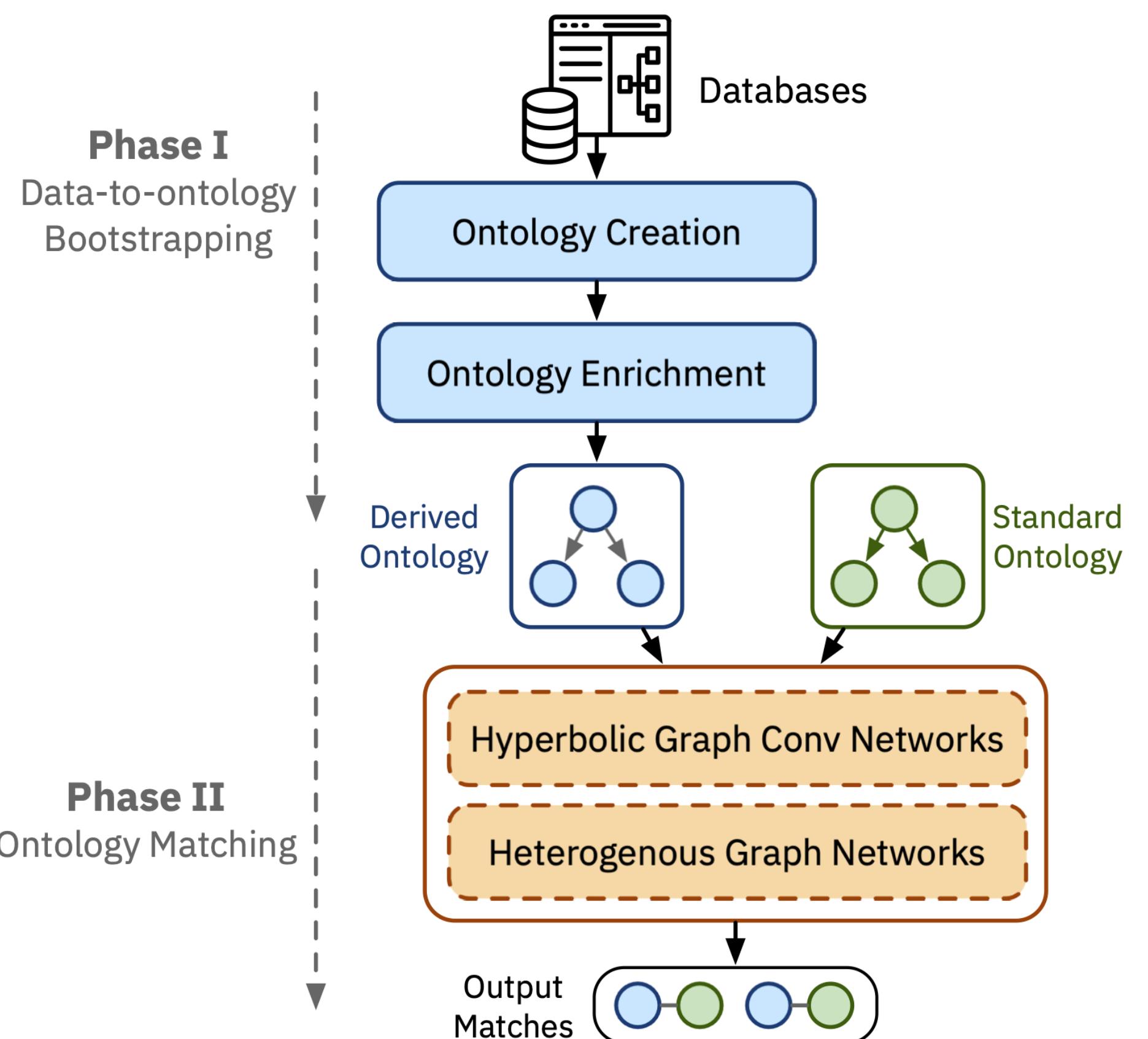
DATASET & BASELINES

Medical Databases: MIMIC-III, IBM Micromedex (MDX)

Medical Ontologies: FMA, NCI, SNOMED-CT

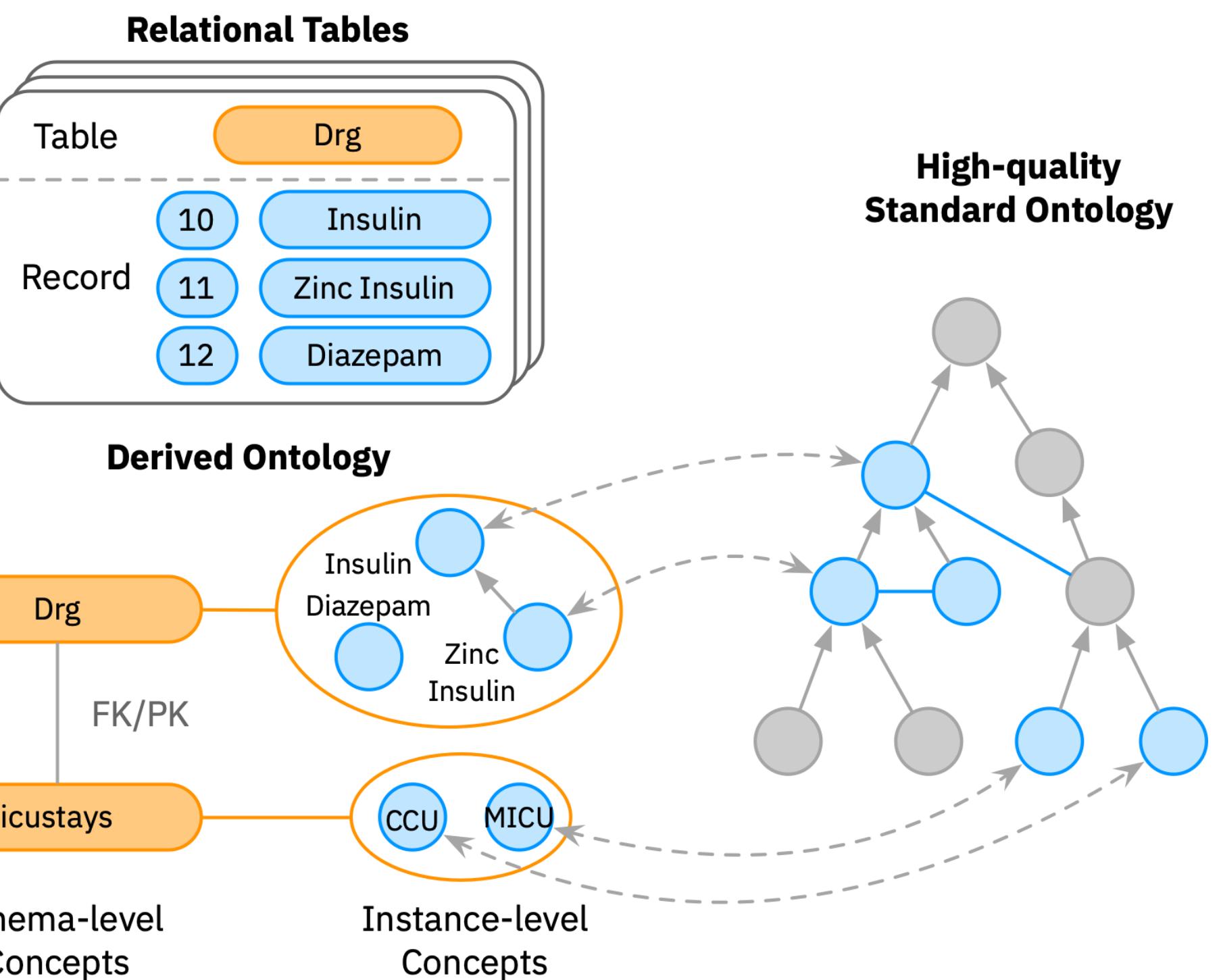
Baselines: AML, LogMap, RDGCN, etc.

MEDTO: SYSTEM ARCHITECTURE

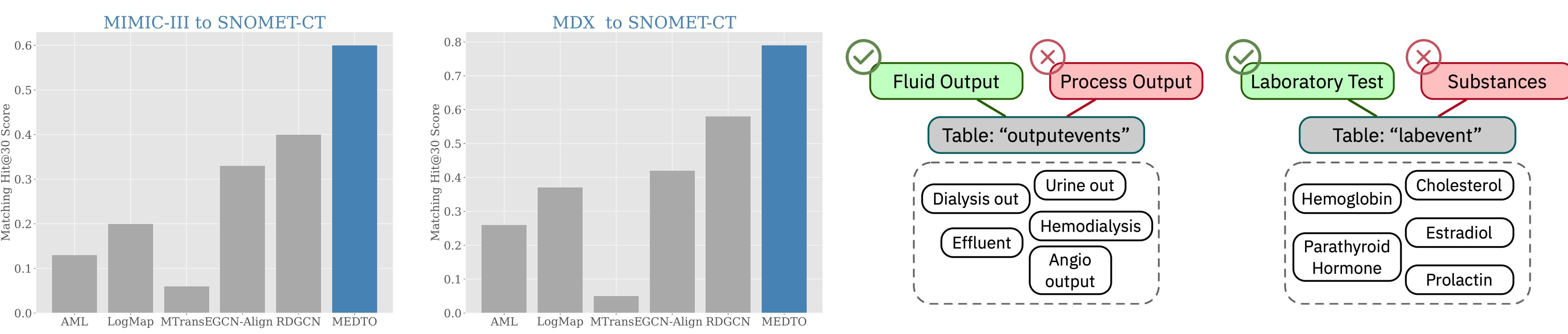


PHASE 1: BOOTSTRAPPING

A two-step process from database tables: (1) Ontology creation; (2) Ontology Enrichment.



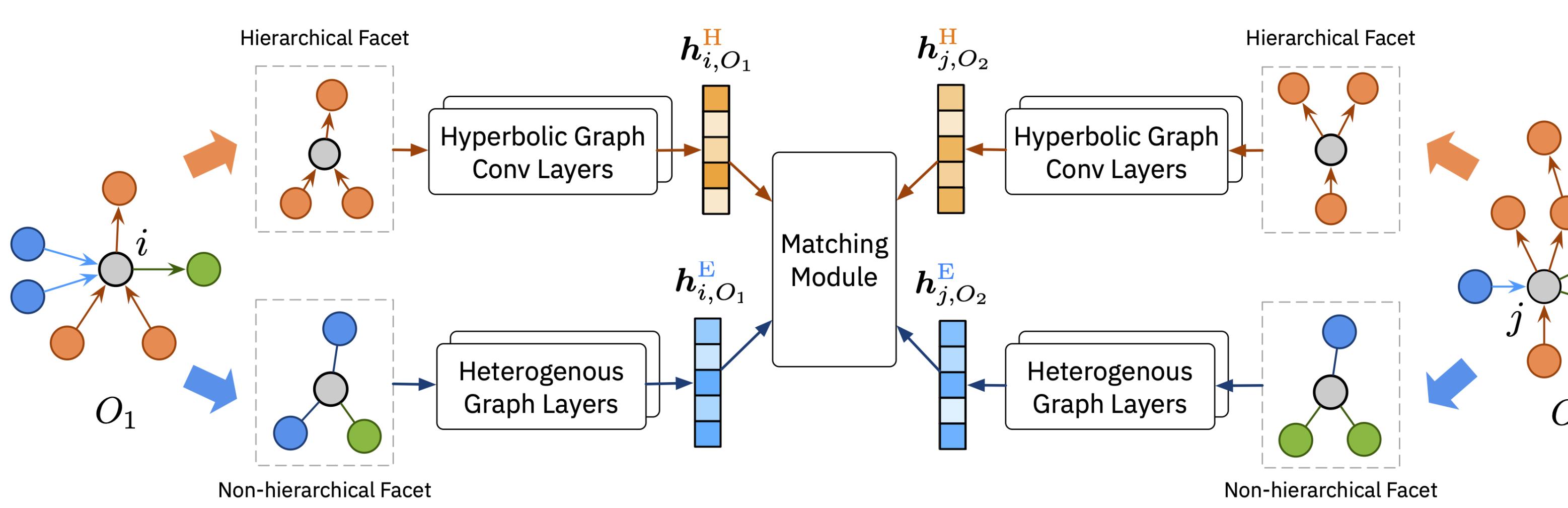
EXPERIMENTS AND CASE STUDY ON MEDICAL DATABASE



EXPERIMENTS ON ONTOLOGY MATCHING

Model Groups	Datasets Metrics	FMA-NCI		FMA-SNOMED		NCI-SNOMED	
		F1	MRR	F1	MRR	F1	MRR
Rule-Based	AML	0.920	-	0.806	-	0.810	-
	LogMap	0.905	-	0.819	-	0.805	-
GNN-based Entity Alignment	MTransE	0.633	0.416	0.490	0.372	0.304	0.349
	GCN-Align	0.798	0.561	0.746	0.526	0.760	0.467
	RDGCN	0.849	0.761	0.786	0.683	0.816	0.679
Ours	MEDTO	0.908	0.783	0.813	0.690	0.849	0.704

PHASE 2: ONTOLOGY MATCHING (GRAPH ENCODERS)



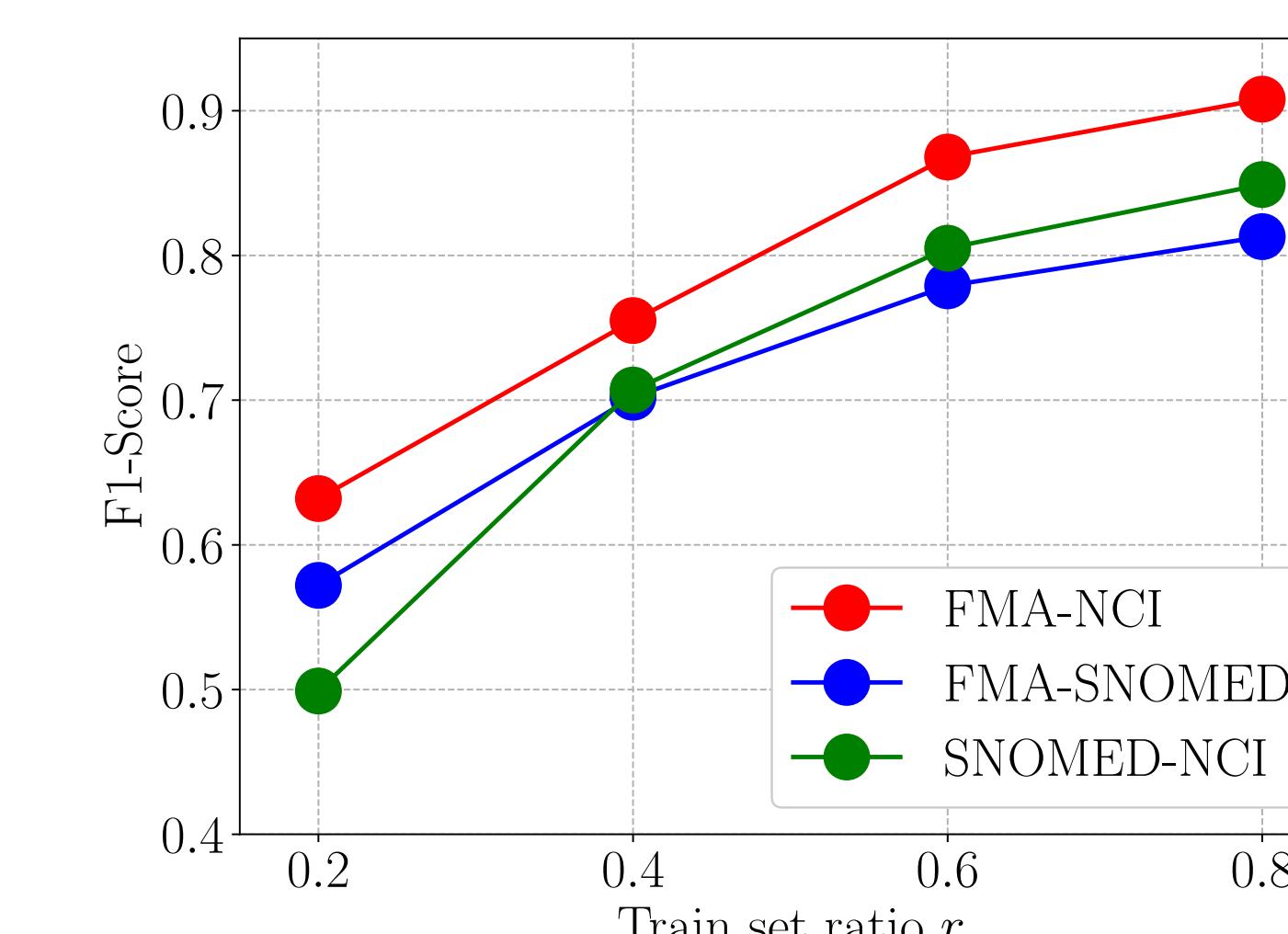
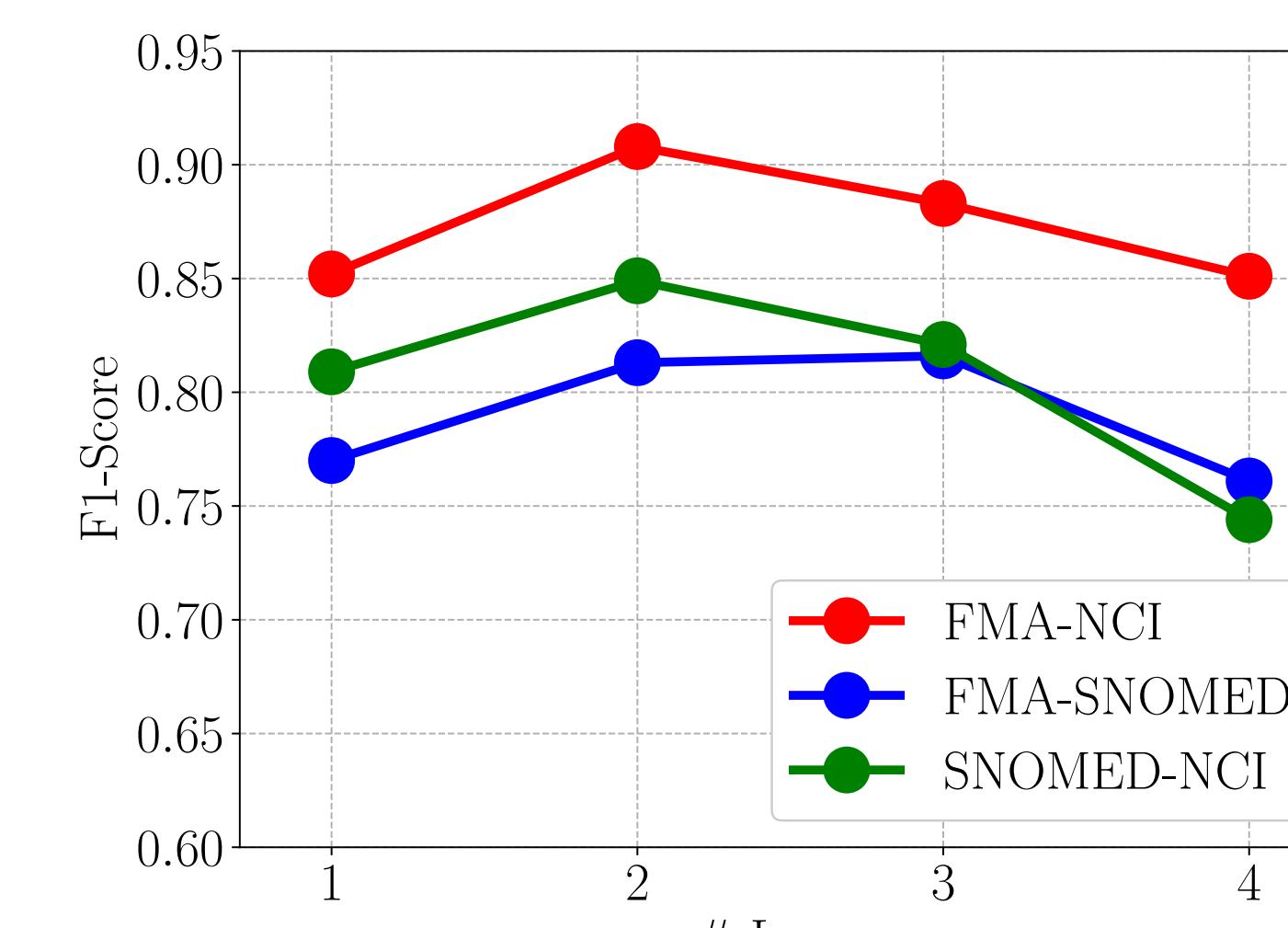
Hyperbolic Graph Module (HYP): Better capture concept hierarchies or taxonomies in the medical ontologies in hyperbolic space (HGCN).

$$\mathbf{h}_i^{l,H} = (\mathbf{W}^l \otimes^{K_{l-1}} \mathbf{h}_i^{l-1,H}) \oplus^{K_{l-1}} \mathbf{b}^l, \quad \mathbf{h}_i^{l,H} = \sigma^{\oplus^{K_{l-1}, K_l}} (\text{AGG}^{K_{l-1}} (\mathbf{h}_i^{l,H}))$$

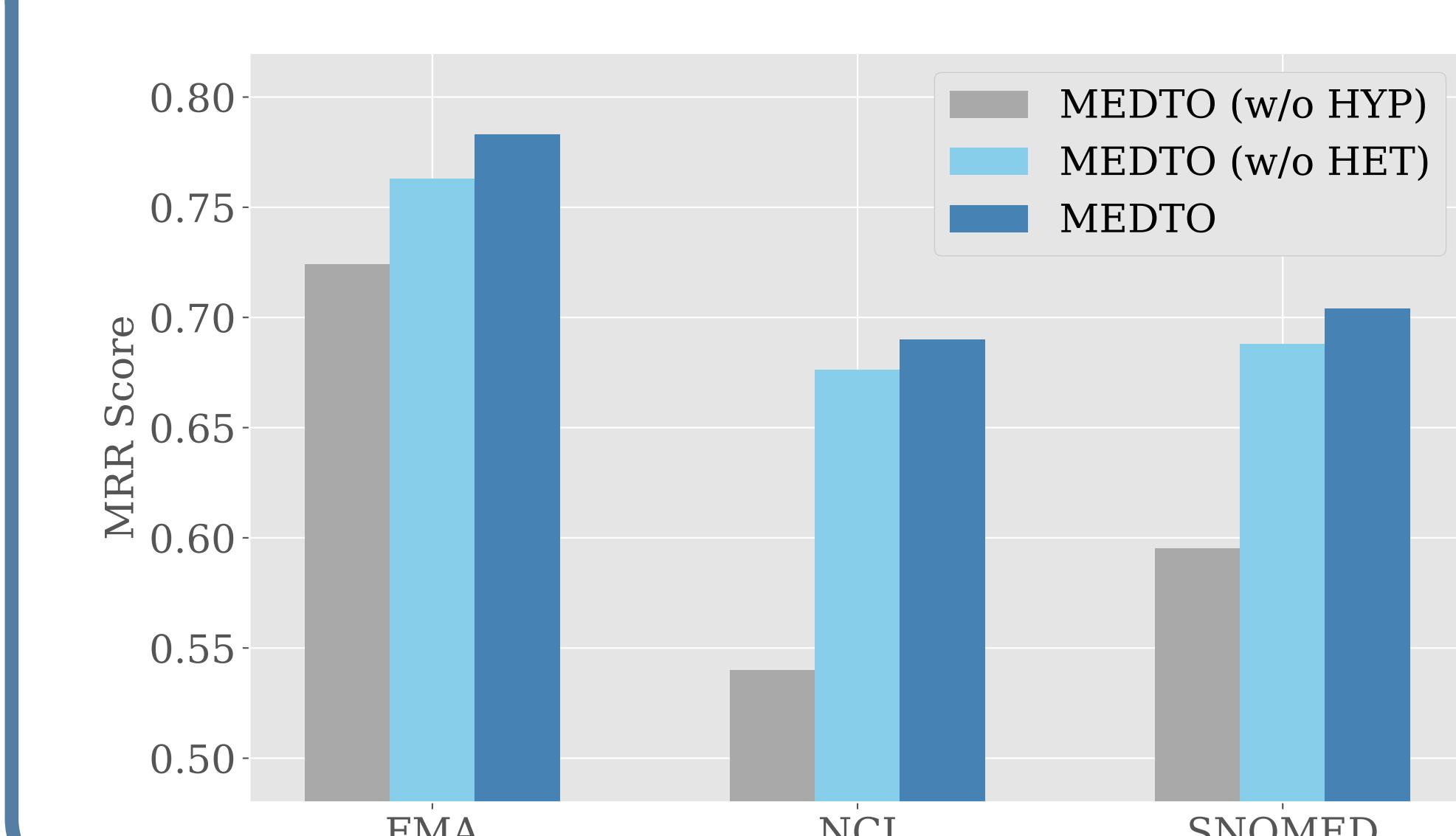
Heterogeneous Graph Module (HET): Model non-hierarchical relational facts with other concepts of multiple types in the ontologies

$$\mathbf{h}_i^{l,E} = \sigma(\mathbf{W}_0^l \cdot [\mathbf{h}_i^{l-1,E} || \mathbf{g}_i^{l-1,E}] + \sum_{r \in \mathcal{R}} \sum_{j \in \mathcal{N}_i} \frac{1}{c_{i,r}} \mathbf{W}_r^l \cdot [\mathbf{h}_j^{l-1,E} || \mathbf{g}_j^{l-1,E}])$$

HYPERPARAMETERS



MEDTO VARIANTS



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