

KARMA: A Knowledge Graph Framework for Explainable and Personalized AI in Healthcare

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Introduction

Data Complexity in Healthcare: Clinical, genomic, imaging, and real-world data in healthcare are often siloed, heterogeneous, and lack semantic interoperability.

Challenges in Data Utilization: Fragmented data formats and inconsistent terminologies hinder integration, analysis, and reuse across systems.

AI Advancements: Deep learning and NLP models have improved diagnostics, treatment recommendations, and predictive analytics in oncology and beyond.

Limitations of Current AI:

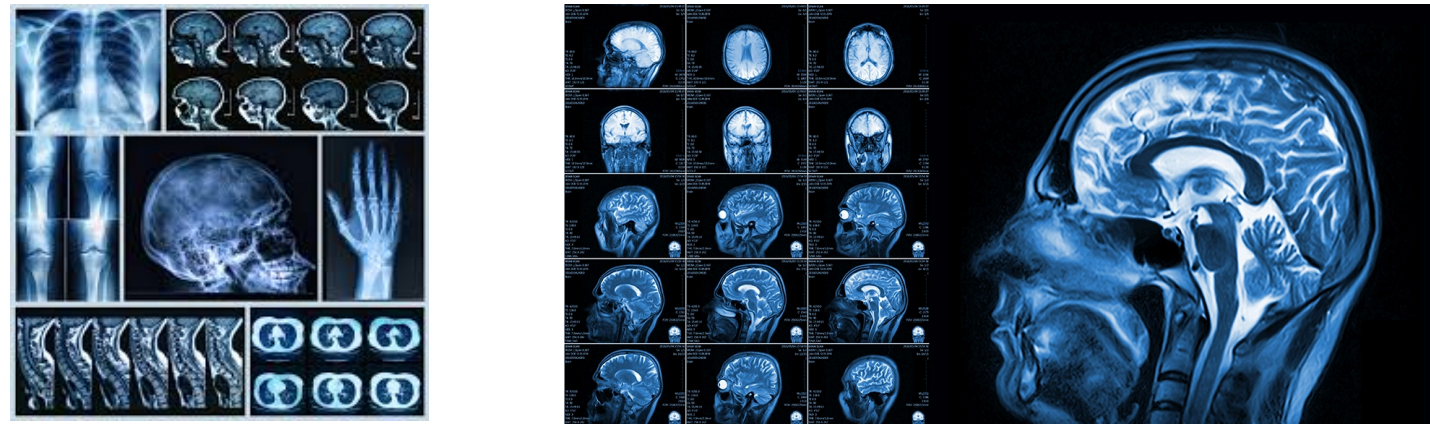
- Often function as "black boxes" with limited explainability.
- CORAL [1] demonstrates the value of structured data extraction in oncology. However, it lacks **robust ontology integration, validations** and does not support community-level patient insights.
- Struggle with incorporating domain knowledge and context.

Potential of Knowledge Graphs (KGs):

- Enable **semantic integration** of diverse data using standardized ontologies[2].
- Support **explainable** and **context-aware** reasoning.
- Facilitate personalized, evidence-driven clinical decision-making.

Domain

Struggles in Healthcare: Healthcare grapples with integrating complex data (EHR, MRI scans, X-rays) for decision-making. Without AI, manual processes are slow and error-prone, often leading to generic treatments. With AI, lack of explainability hinders trust, and semantic data relationships are often missed, limiting personalization [3][4]



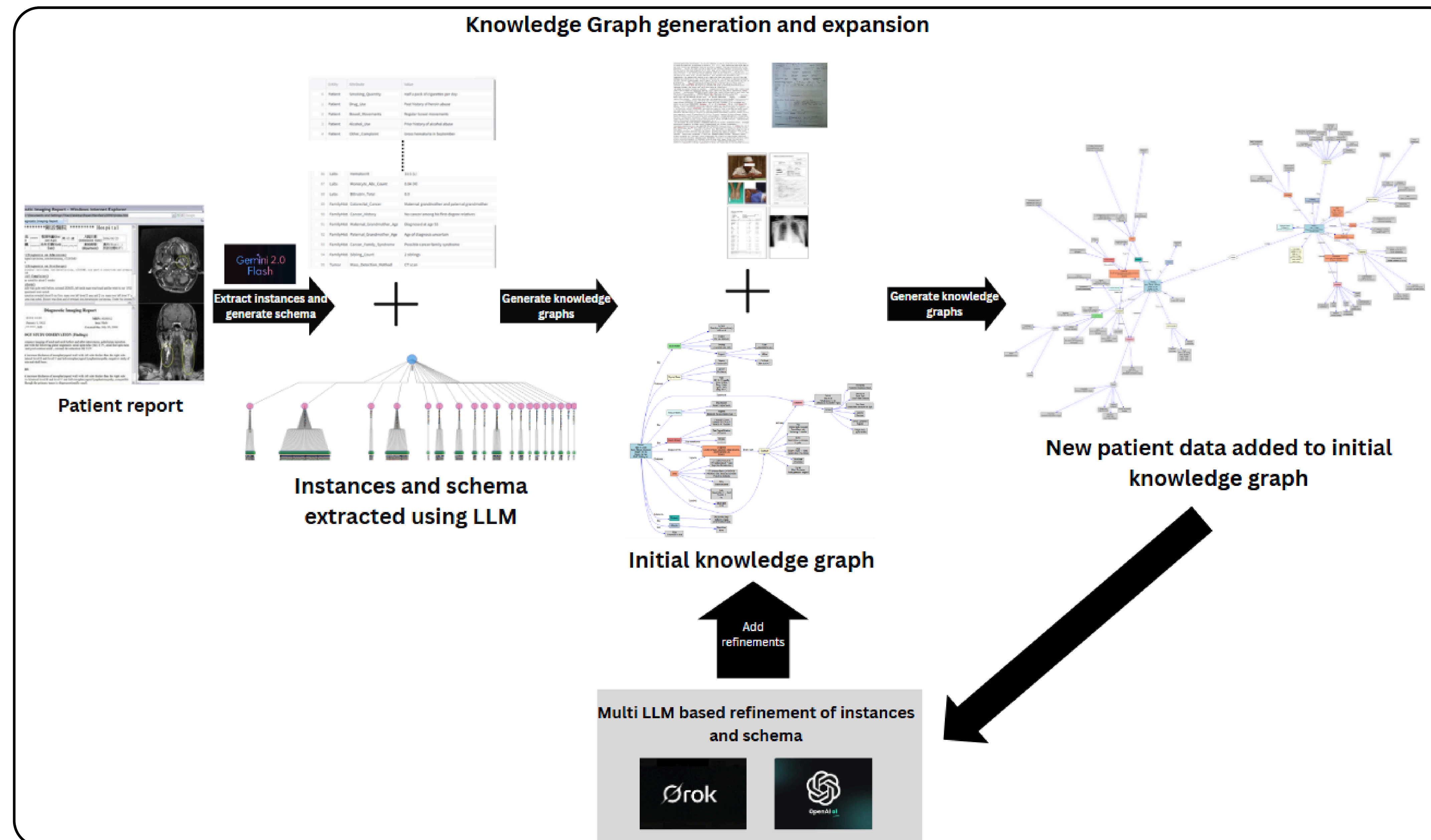
Relevance of KARMA: KARMA introduces a knowledge graph framework to unify diverse healthcare data, making AI recommendations both explainable and personalized. We tested KARMA using the CORAL dataset (Expert-Curated Oncology Reports), which includes real-world oncology reports—detailed medical summaries of cancer patients, such as their diagnoses, treatments, and genetic information. For example, CORAL data might describe a pancreatic cancer patient's tumor mutations and past chemotherapy responses. KARMA uses this data to connect a patient's genetic profile to specific drug mechanisms, helping doctors understand why a particular treatment is recommended (e.g., suggesting a chemotherapy drug tailored to a patient's unique cancer markers). By making these connections clear and actionable, KARMA bridges the gap between complex data and personalized, trustworthy care, addressing both manual and AI-related challenges.

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Methodology

- Data Normalization:** Extract and standardize clinical data with metadata annotation for consistency.
- Entity Extraction:** Use **Gemini Flash 2.0** to convert clinical entities into RDF instances.
- Relationship & Schema Creation:** Define semantic relationships and construct initial ontology-aligned schemas.
- Knowledge Graph Construction:** Build ontology-enriched patient KGs using structured RDF and standard ontologies.
- Graph Expansion:** Incrementally update schemas and RDF with new patient data for KG growth.
- Community Detection & Reasoning:** Apply Louvain clustering and semantic rules (SPARQL, SWRL) for personalization.
- Dynamic Query Generation:** Convert clinician keywords into precise SPARQL queries via backend AI (Gemini Flash 2.0).
- Variable Discovery Refinement:** Use Grok and GPT-o1 for second-pass entity-attribute refinement via model agreement.



Knowledge graph applications

Compare Vital Signs Across Patients

Find Treatments and Their Details for Patient_1

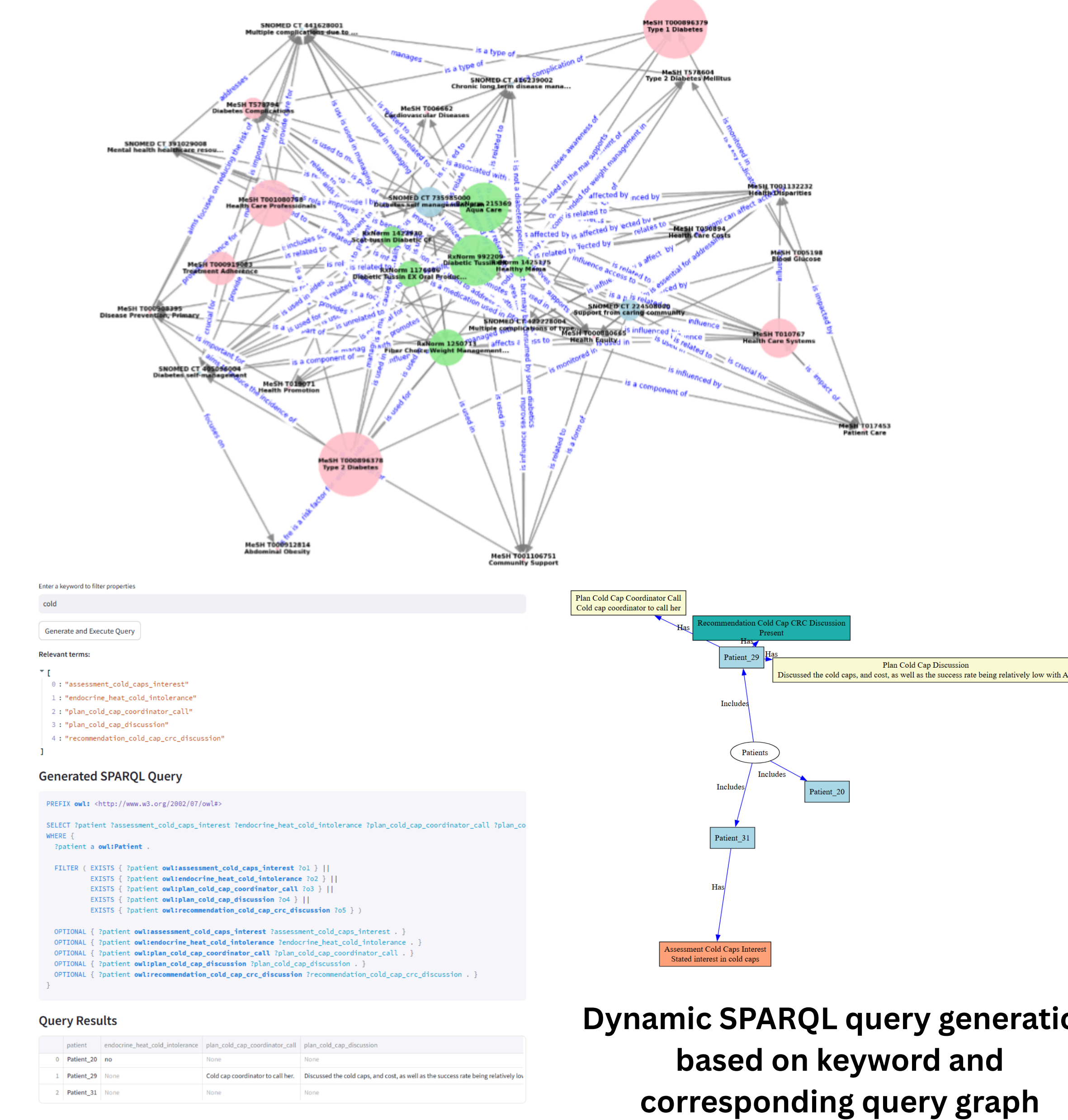
Retrieve Basic Information for Both Patients

Dynamic SPARQL query generation and execution with the help of LLM

Results

A case study with two patient groups—**Breast Cancer (BRCA)** and **Pancreatic Cancer (PDAC)**, with 20 patients in each group—demonstrated KARMA Web's effectiveness of generating a knowledge graph for patient 0 and extending it by adding data from other patients subsequently.

Knowledge graph generated in application



Future works and conclusion

KARMA enables scalable, ontology-enriched clinical knowledge graphs using LLMs.

Future work includes refining variable extraction with **GPT-o1** and **Grok**, evaluated via **ontology coverage, semantic consistency, hallucination rate, and information gain**—enhancing reliability and reducing model hallucinations.

References

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