







# HypER: Literature-grounded Hypothesis Generation

and Distillation with Provenance

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• LLMs can act as AI scientists, generating hypotheses

But scientific hypotheses must be logically grounded,

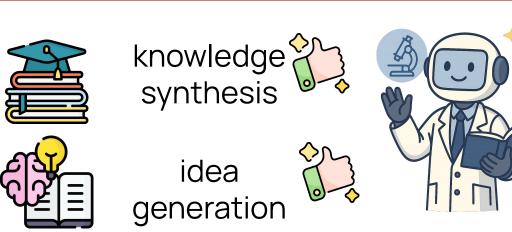
interpretable, and based on existing knowledge.



HypER Code

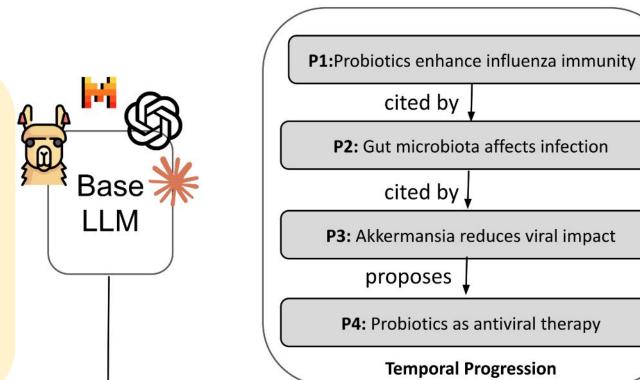


### Motivation

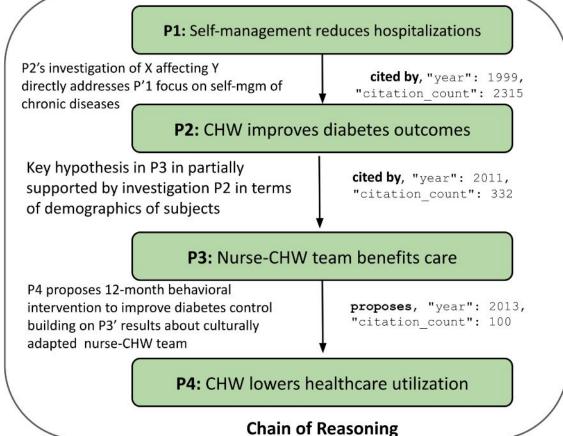


experiment 2

and designing experiments.



(1) Generated hypothesis (2) Generated hypothesis (no grounding (no coherent argument for entire chain) information)



(3) Generated well-grounded hypothesis (with literary provenance and coherent reasoning)

#### Existing work:

- Reliance on co-occurrence patterns or surface-level similarity
- Poor logical progression and no provenance between ideas

In biomedical research, hypotheses must be grounded in evidence, logically interpretable, and scientifically valid → not just creative.

## HypER: Our Approach

HypER is fine-tuned jointly on three complementary tasks to capture reasoning from local dependencies to multi-hop chains

#### One-hop Relevance Classification (1-hop)

**Input:** source paper + target paper

**Output:** fine-grained relevance score  $\rightarrow$  {0: irrelevant, 1: inspired, 2:

dependent}

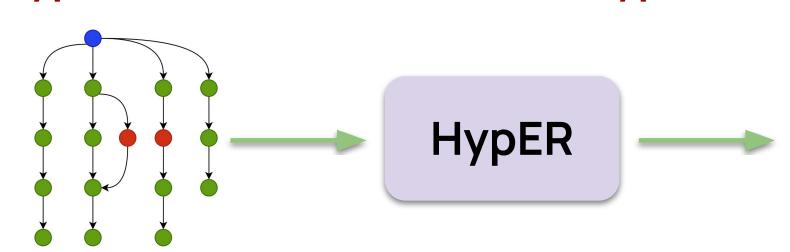
#### Multi-hop Agnostic Chain Validation (multi-hop-A)

Input: temporally ordered paper chain Output: valid / invalid + breakpoints if invalid

#### Multi-hop Contextual Chain Validation (multi-hop-C)

Input: paper chain + target hypothesis Output: valid / invalid + breakpoints

#### Hypothesis Generation with HypER



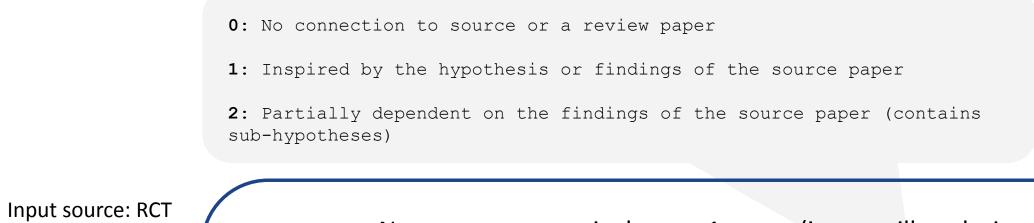
Analysis: . . . Rationale: . . .

Research Idea: . . .

Hypothesis: In patients with critical limb ischemia due to infrapopliteal artery disease, drug-eluting BVS will result in higher primary patency rates, lower rates of major adverse limb events, and improved limb salvage rates compared to angioplasty at 1 year.,....

### **Building Temporal Reasoning Chains**

Curated temporal reasoning chains form the backbone of HypER



next two years (Pi)

How can we train an LLM to navigate the noisy literature and generate novel and

impactful ideas that are grounded in a solid understanding of existing work?

Review paper source paper, year Citing papers from and associated

Next **source paper** is the top-1 paper (iterate till exploring 2024 papers) Relevancy Top relevant papers Scoring Model Top-1 Paper Selection Pi -> {title, explanation, (Relevance + Impact)

relevancy score}

Valid chains:

Summarization data

pubmed

papers

→ directly obtained from the iterative construction process (379 chains)

### **X** Negative chains:

- → Easy: swap with irrelevant paper (1184)
- → Hard: break or insert incoherent links (455)

### **TakeAways**

- +22% absolute F1 over base models in distinguishing valid vs. invalid reasoning chains.
- Generates more evidence-grounded hypotheses (0.327 vs. 0.305) baseline).
- Human experts (clinicians & biomedical researchers) rated outputs > 3.5 / 5 for feasibility and impact, in some cases anticipatory of recent studies.

Example of a hypothesis generated by HypER with expert review

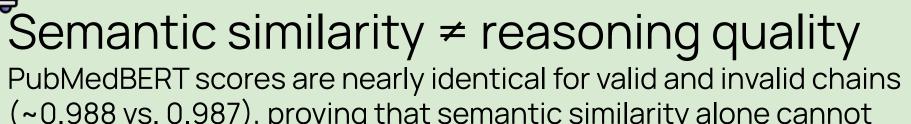
#### **Research Questions**

RQ1. Can HypER differentiate between valid and invalid reasoning chains?

RQ2. Does reasoning chain validation improve the quality of generated hypotheses?

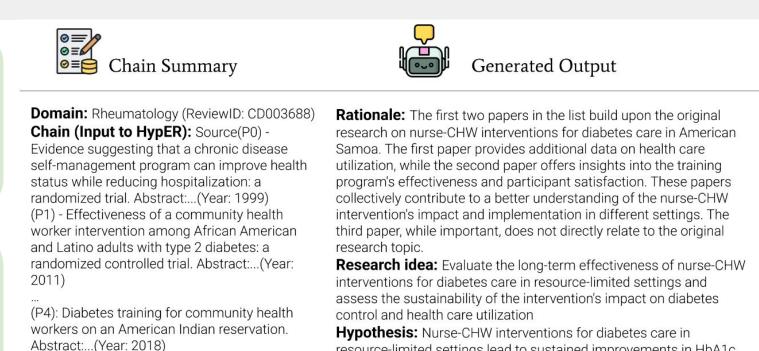
#### **Dataset and Models**

- Models: Phi-3-3.8B, LLaMA-3.2-3B, MistralLite-7B-32K
- Chosen for similar size, instruction-tuning, and long-context handling
- Baselines: no reasoning-chain supervision
- HypER refers to the fine-tuned version of the Phi-3-mini-128k-instruct-3.8B model
- Chosen as main model performed best among all tested SLMs



(~0.988 vs. 0.987), proving that semantic similarity alone cannot ensure logical coherence.

- HypER aligns with valid parts of partially invalid chains, showing deeper reasoning awareness.
- Captures scientific dependencies and differentiates valid vs. invalid reasoning chains beyond surface similarity.
- Generates grounded hypotheses and enables evidence-driven, structured research ideation.



(P5): What Is the Impact of a Context-Specific

Training Program for Home-Based Carers? An

Evaluation Study Abstract:...(Year: 2020)

Chain Label: Valid

on improving health status of chronic patients using community based intervention programs **Feasibility (5)**: "Somewhat confident that hypothesis is testable" Originality (4): "Actual paper talks about chronic disease self management and its impact leading to reduced hospitalization, while resource-limited settings lead to sustained improvements in HbA1c the hypothesis focuses more on Nurse CHW intervention and its levels and reductions in emergency department visits for impact on Diabetes care. Chronic illnesses may also include asthma, arthritis, etc. not limited to diabetes" diabetes-related complications. Impact (4): LLM-as-judge "https://www.sciencedirect.com/science/article/pii/S221133552400 0603. This was present in recent papers" Clarity: (3) Feasibility (3) Originality (3) Impact (3.6)

Rationale followed from the analysis?

"Yes. it follows with the rationale as it focuses