HypER : Literature-grounded Hypothesis Generation and Distillation with Provenance

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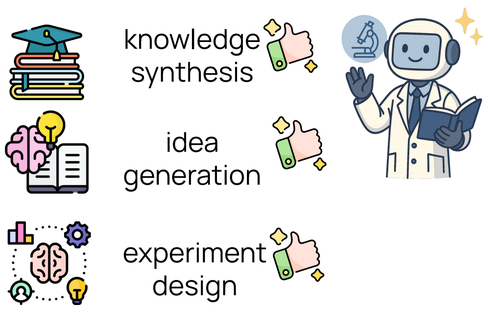
HypER Code

HypER PDF

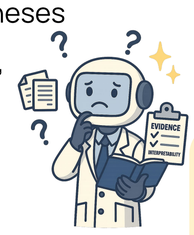
Motivation



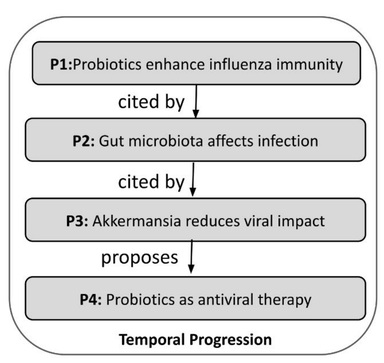
- LLMs can act as AI scientists, generating hypotheses and designing experiments.
- But scientific hypotheses must be logically grounded, interpretable, and based on existing knowledge.

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

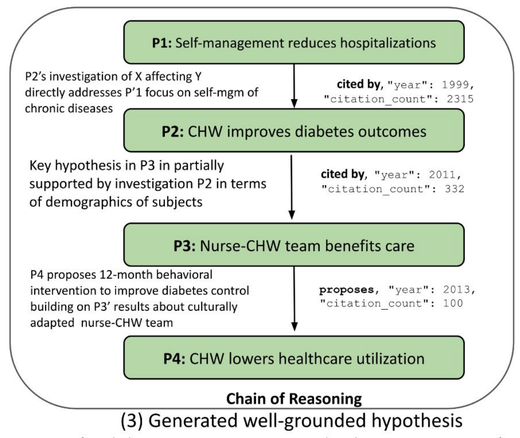
In biomedical research, hypotheses must be grounded in evidence, logically interpretable, and scientifically valid → not just creative.



(1) Generated hypothesis (no grounding information)



(2) Generated hypothesis (no coherent argument for entire chain)



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

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?

HypER: Our Approach

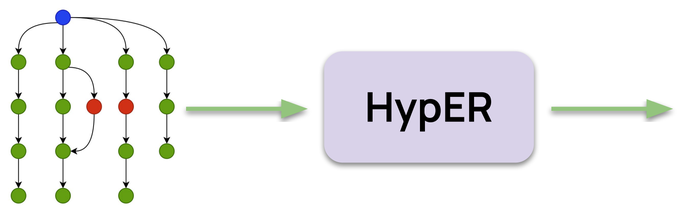
HypER is fine-tuned jointly on three complementary tasks to capture reasoning from local dependencies to multi-hop chains

One-hop Relevance Classification (1-hop)
Input: source paper + target paper
Output: fine-grained relevance score → {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

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0: No connection to source or a review paper

1: Inspired by the hypothesis or findings of the source paper

2: Partially dependent on the findings of the source paper (contains sub-hypotheses)
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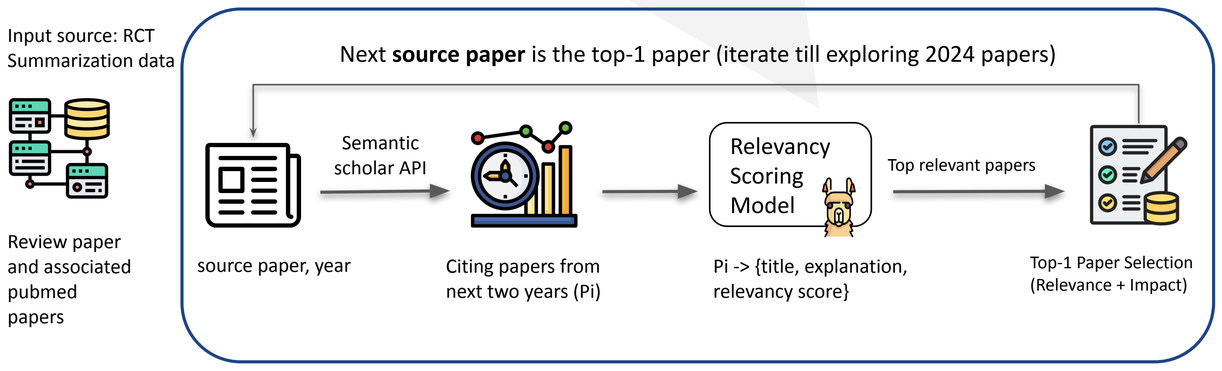
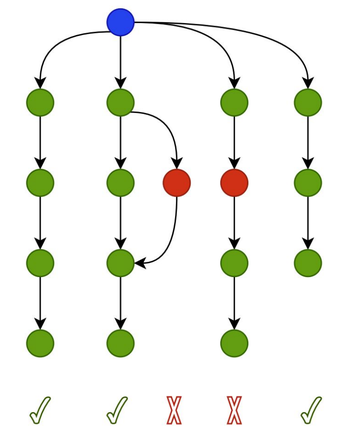


✓ **Valid chains:**

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

✗ **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

Semantic similarity ≠ reasoning quality

PubMedBERT scores are nearly identical for valid and invalid chains (~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.