

Short Paper Track

CardiO: Predicting Cardinality from Online Sources

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| Motivation | | | |
|---|--|--|--|
| Counts from top SE snippets returned for count questions | | Incompleteness and bias in online world | |
| how many AI startups in Europe? (well-defined, but not very popular) | how many programmers in Europe? (vaguely-defined, quite popular) | Different sources focus on different angles. | |
| Few relevant counts Europe has 1,157 AI startups Counts from popular subsets | Frequent relevant estimates estimated 6.1 million developers approx. 6.1 million software developers Noise 500.000 developers short in job market Time-variance July 2023 approx. 5.7 million software developers in Europe | Topic popularity and crispness affect count availability. Distribution of counts on the Web Many relevant contexts with close counts; noise. Explanatory evidence over direct answers | |

| | how many beaches are Blue Flag certified? |
|----------------------|---|
| More count questions | how many films produced by Warner Bros? |
| • | how many lakes are there in the world? |

- > CardiO is a lightweight and modular framework.
- > Provides traceability with tangible evidence.
- > Makes distribution-aware prediction by using supporting evidence

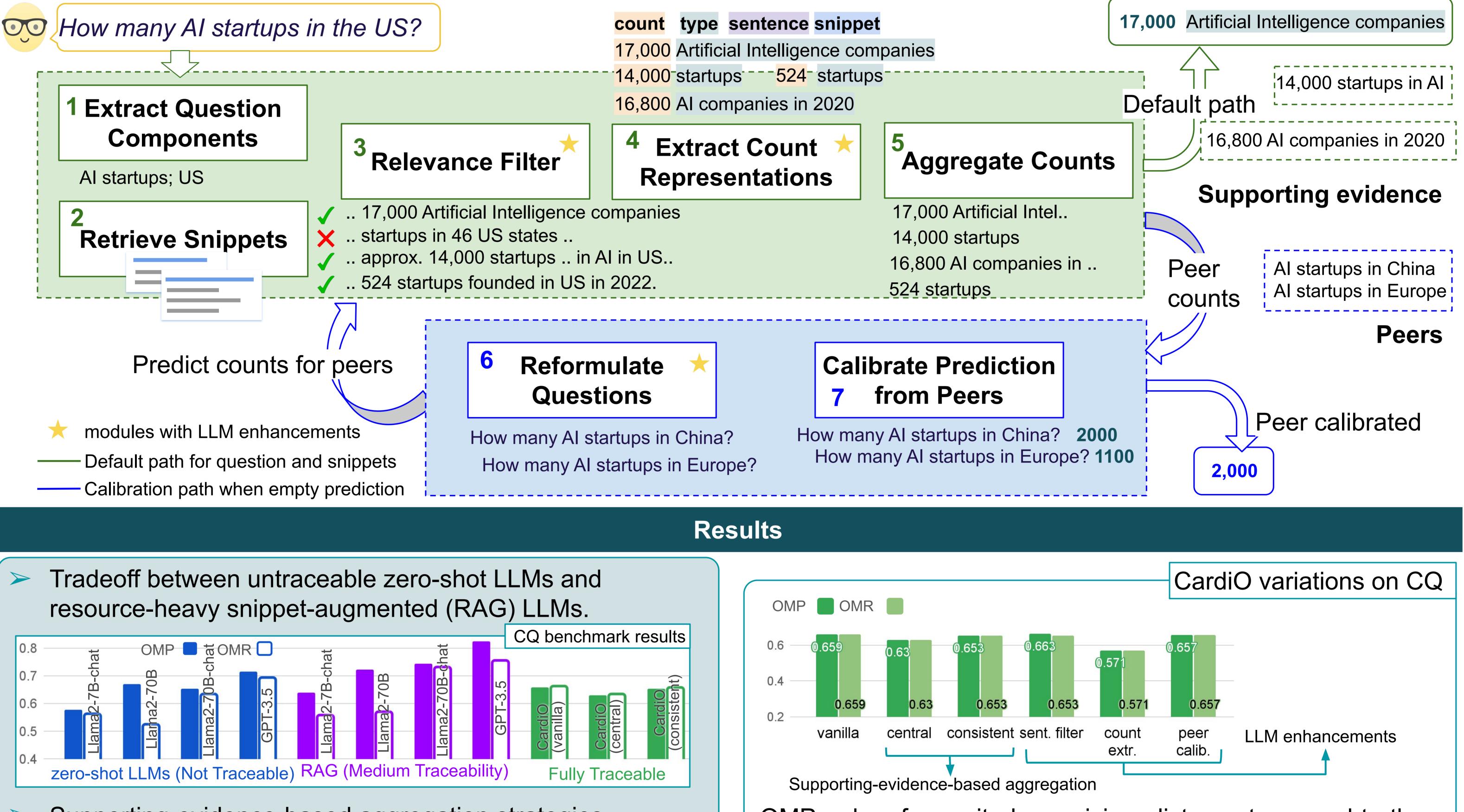
Context, instances, and snippets increase user comprehension.

LLMs have high precision but lack traceability

LLM-only approach has no traceability and relies on (outdated) parametric knowledge.

Retrieval-augmentation does not guarantee provenance.

CardiO Framework



Supporting-evidence-based aggregation strategies.

OMP order of magnitude precision: distance to ground-truths. OMR order of magnitude recall: penalize empty predictions when ground-truth present.

- Controlled LLM enhancements in three modules:
 - Sentence filter based on relevance
 - Count representation extraction
 - Peer calibration from entities with close cardinalities



Cardinality Benchmarks

 CQ (ours): 500 questions; annotated question and ground-truth properties.
 NQ: 84 count questions from Natural Questions dataset.
 CoQuAD: 312 count questions.

[1] Ghosh et al. 2023. Class Cardinality Comparison as a Fermi Problem. In WWW [2] Ghosh et al. 2022. Answering Count Queries with Explanatory Evidence. In SIGIR Our research page on Count Knowledge: https://www.mpi-inf.mpg.de/count-knowledge Dataset and Results: https://github.com/ghoshs/CardiO

