Scaling up Logical Query Embeddings on Knowledge Graphs



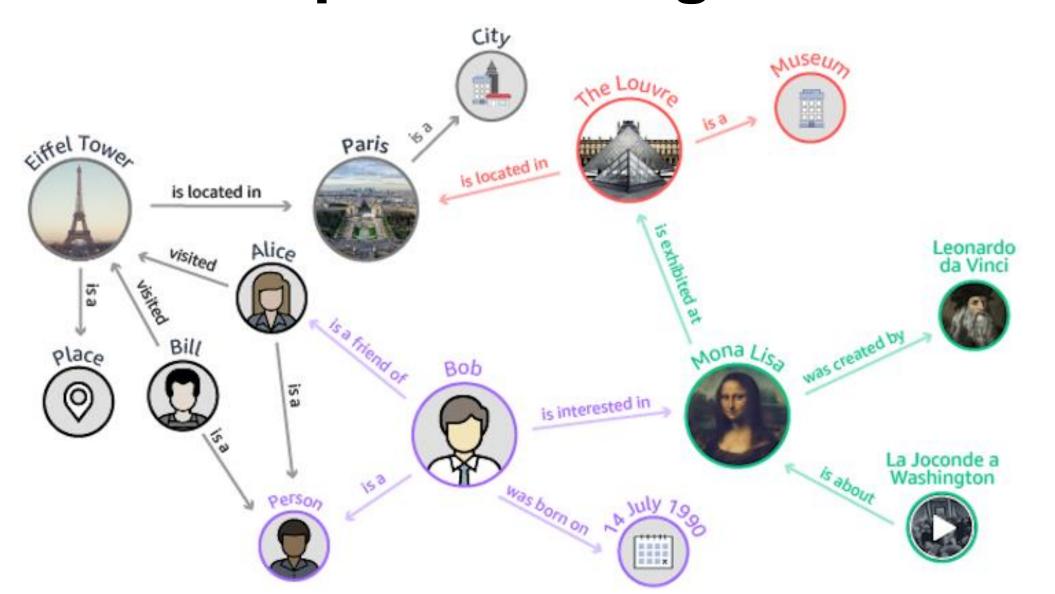


Google Research

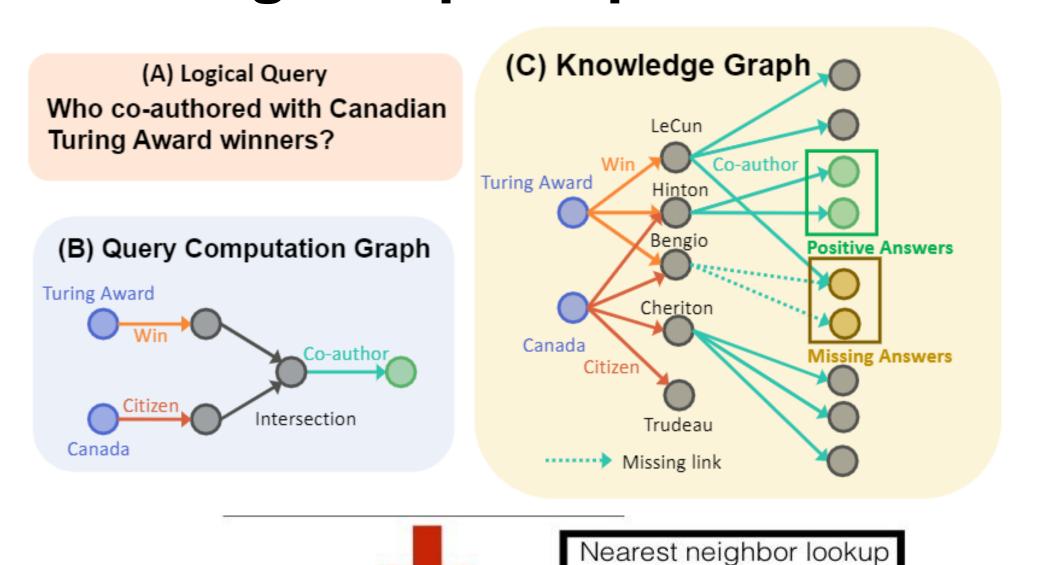
Hongyu Ren*, Hanjun Dai*, Bo Dai, Xinyun Chen, Denny Zhou, Jure Leskovec, Dale Schuurmans

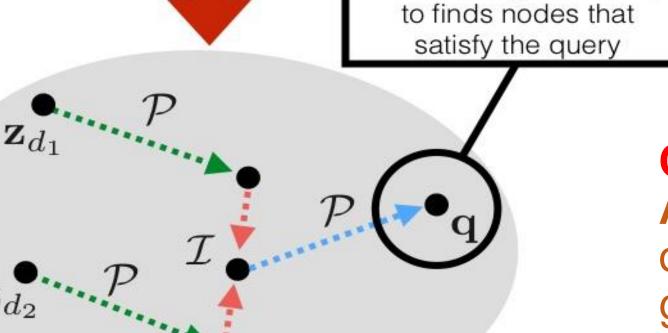
Stanford University, Google Brain, UC Berkeley

Multi-hop Reasoning on KGs



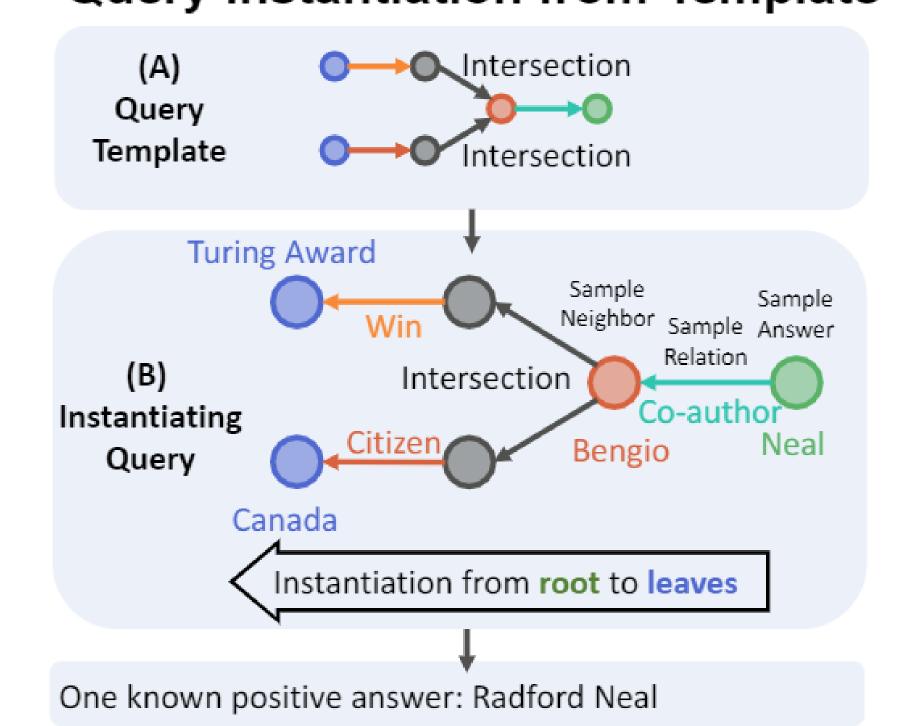
Answering complex queries on KGs

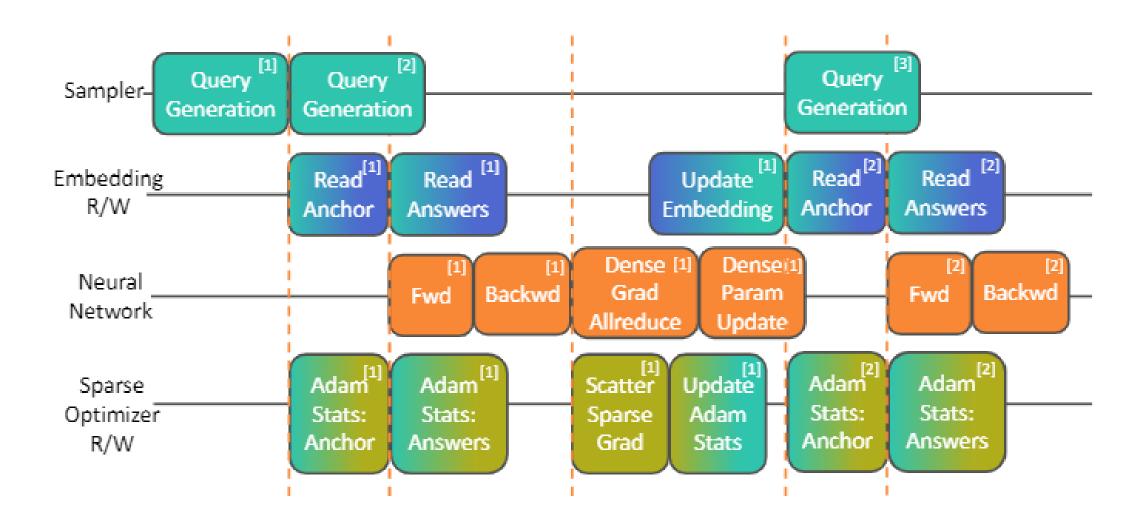




Operations in an embedding space

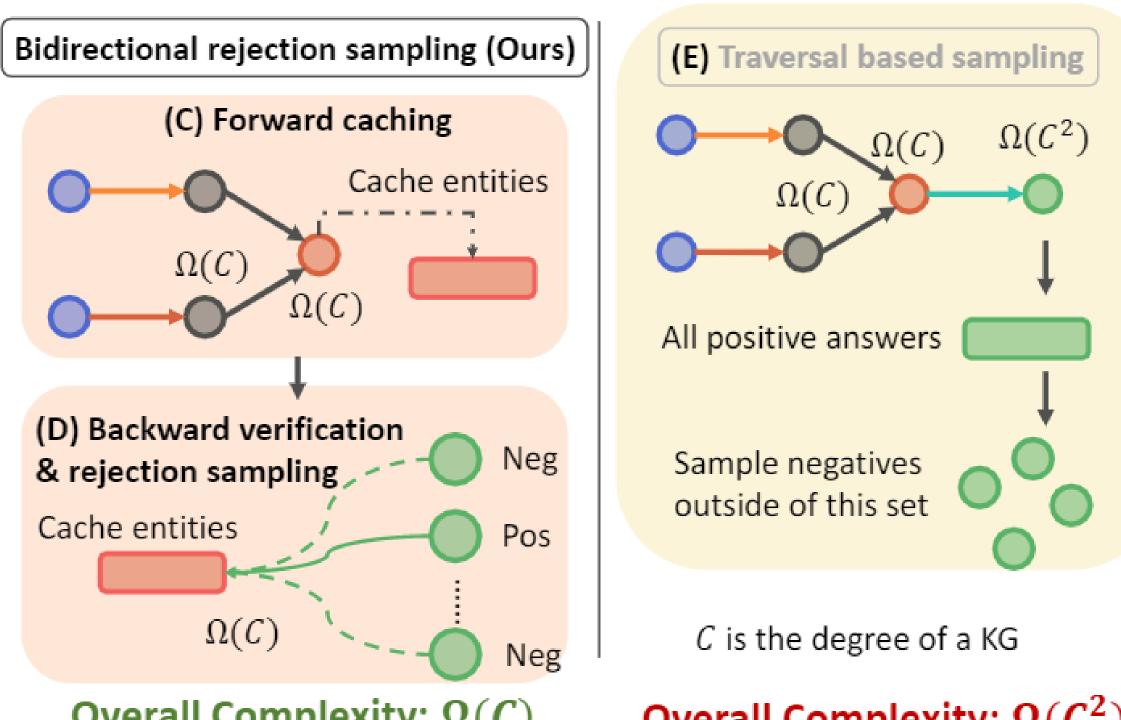
Query Instantiation from Template



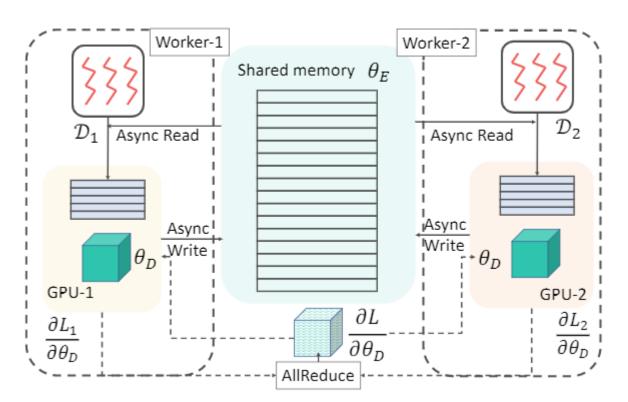


SrKG: Scalable reasoning Knowledge Graph Embeddings

Negative Entities Sampling



Overall Complexity: $\Omega(C)$ Overall Complexity: $\Omega(C^2)$



SrKG performance:

Batch index

Read:

- (1) For the first time, scale reasoning algorithms to KGs with more than 86m nodes
- (2) On small graphs, +99.5% speed, -33.6% GPU memory

Concurrent

(3) Achieve (almost) graph-size agnostic speed and GPU usage

Challenges to scale to large KGs:

Algorithmic: needs an efficient training data (query, one positive answer, several negative answers) generation algorithm

System: (1) sparse + dense parameters (2) scheduling of training data generation, embedding r/w, neural network feed forward, optimization