

Which Programming Language and Model Work Best With LLM-as-a-Judge For Code Retrieval?

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Code Search and Code Generation-R is for Retrieval

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Why this matters?

Developer onboarding, productivity, and maintenance.

Using a Retrieval Augmented code Generation (RAG)?

Even if you aren't searching code directly, your RAG system is. R (in RAG) is for retrieval.

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Relevance and Benchmarks

- Tedious, time consuming, costly, difficult to find annotators, difficult to build data, and difficult to setup good dataset.

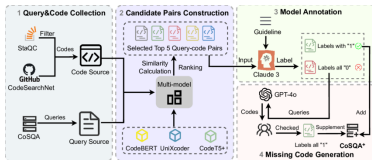


Figure: Construction of CosQA benchmark.

Why this matters?

A search system is a complex system of interlocking technologies.

Benchmarks allow us to track changes in performance across system changes.

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LLM-as-a-judge

- ▶ Given query text and contents of each search result, concatenate with prompt template.
- ▶ Compare these relevance labels with those from human judgment.

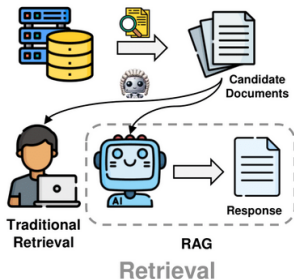


Figure: A standard LLM-as-a-judge workflow.

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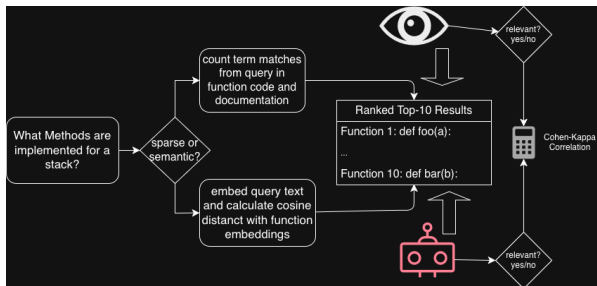
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Experiment Workflow



LLM-as-a-judge annotation experiment process workflow.

- ▶ Sparse search uses BM25 for scoring/ranking.
- ▶ Semantic retrievers are transformers models trained on text & code. CodeBERT is encoder, CodeT5+ is encoder-decoder.
- ▶ Human (as-a-judge) label each query-result pair as relevant or not.
- ▶ LLM-as-a-judge: Nova-lite, gemini-2.0, gpt-4o-mini, Llama-4.

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Research Questions

- ▶ Are there affinities between Programming Languages (PL) and LLMs when evaluating relevance?
- ▶ Does the representation (sparse vs semantic) matter when evaluating relevance?
- ▶ If there are affinities with PLs how to scale benchmarks to other PLs? (most existing benchmarks are in a single programming language)

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- ▶ We select git repositories with open source licenses across 5 popular programming languages.
- ▶ Java, Javascript, Go, Python, C.
- ▶ Each repository contains common data structures, Stacks, Priority Queues, Maps, etc.
- ▶ Parse the functions for each repo using a (programming) language parser, tree-sitter.



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Corpus Statistics

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| | C | Js | Python | Go | Java |
|---------------|--------|--------|--------|---------|--------|
| Stack | ✓ | ✓ | ✓ | ✓ | ✓ |
| List | ✓ | ✓ | ✓ | ✓ | ✓ |
| Set | ✓ | ✓ | ✗ | ✓ | ✓ |
| Map | ✓ | ✓ | ✗ | ✓ | ✓ |
| Ordered Set | ✓ | ✓ | ✗ | ✓ | ✓ |
| Tree | ✗ | ✗ | ✓ | ✓ | ✓ |
| Queue | ✓ | ✓ | ✓ | ✓ | ✓ |
| Heap | ✓ | ✓ | ✓ | ✓ | ✓ |
| Trie | ✗ | ✗ | ✓ | ✗ | ✗ |
| % Docs Absent | 25.17 | 93.87 | 56.94 | 54.01 | 19.91 |
| # Functions | 576 | 163 | 144 | 1,409 | 844 |
| Lines of Code | 7,285 | 1803 | 978 | 16,567 | 6,515 |
| # Doc Tokens | 32,762 | 677 | 502 | 17,344 | 27,057 |
| # Code Tokens | 43,257 | 12,326 | 6,724 | 132,402 | 42,482 |

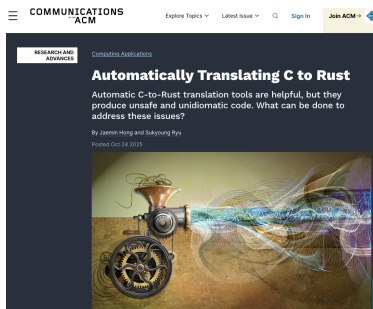
Search Queries

- ▶ We construct a list of queries, to cover the 3 main types of queries listed by Broder's taxonomy of web search.
- ▶ Some queries are constructed specifically for a certain data structure whilst others are parametrized by data structure name.
- ▶ Compared 3 retrievers, sparse: BM25, semantic: CodeBERT, and CodeT5+.
- ▶ Relevance labels from humans:

$$\underbrace{2}_{\text{humans}} \times \underbrace{5}_{\text{PLs}} \times \underbrace{3}_{\text{Retriever}} \times \underbrace{33}_{\text{query}} \times \underbrace{10}_{\text{Results size}} = 9,900.$$

Transpiler Experiment

- ▶ Translate from Python to C.
- ▶ We use the Cos-QA dataset which contains 19,000 query, code, and relevance markings.
- ▶ Send the translated code to the LLM-as-a-judge.



Transpilers-Noteworthy Tools, Comm. ACM Oct 2025 Cover.

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Programming Language Affinities

Cohen- κ correlation between human and LLM relevance determinations

| | Code T5+ | | BM25 | |
|--------|-----------|-------------|-----------|-------------|
| | Nova-lite | GPT 4o-mini | Nova-lite | GPT 4o-mini |
| Python | 0.23112 | -0.05479 | 0.25286 | 0.02932 |
| C | 0.00075 | 0.22783 | 0.07680 | 0.36064 |
| Go | 0.13167 | 0.00389 | 0.28438 | 0.10107 |
| Js | 0.03704 | -0.00464 | 0.14663 | 0.06904 |
| Java | 0.08710 | 0.19906 | 0.26513 | 0.01243 |

Many clear preferences-regardless of representation.
Many more metrics and model comparisons in the paper.

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Representations

Improvement or Reduction in Human-AI Relevance Annotation Alignment (Cohen- κ correlation), Best Sparse and Best Semantic.

| Language | BM25 | CodeT5+ | % change |
|----------|----------|---------|----------|
| Python | 0.255286 | 0.23112 | - 9.5% |
| C | 0.36064 | 0.22783 | -36.8 % |
| Go | 0.10107 | 0.13167 | 30.3 % |
| Js | 0.14663 | 0.03704 | -74.7 % |
| Java | 0.26513 | 0.19906 | -24.9 % |

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Semantic Search Finds Data Structure Synonyms

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Search the repo: Codes

Two alternately implemented and
named data structures, heap and
priority queue are both surfaced
at the top with the heap query.

Size distance 437:

Size distance 4581

Semantic Search finds heap/priority queue synonym.
Whereas sparse methods like BM25 require thesaurus/query
expansion/wordnet etc.

CosQA Transpiler (Py-2c) Results

LLM-as-a-judge on transpiled CosQA data, Python transpiled to C.

| <i>nova-lite</i> | not-relevant | relevant | |
|--------------------|--------------|----------|------|
| not-relevant | 3845 | 987 | 4832 |
| relevant | 3171 | 1018 | 4189 |
| 53.91% | 7016 | 2005 | 9021 |
| <i>gpt-4o-mini</i> | not-relevant | relevant | |
| not-relevant | 2657 | 2175 | 4832 |
| relevant | 2060 | 2129 | 4189 |
| 53.05% | 4717 | 4304 | 9021 |

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Findings Summary

- ▶ There are affinities with LLM-as-a-judge and PLs.
- ▶ Representation matters, perhaps (un)surprisingly, BM25 is a strong candidate for retrieval mechanism for code. Many design consequences from this choice.
- ▶ To scale benchmarks to other PLs consider using a transpiler.

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