MiBi at BioASQ 2024: Retrieval-Augmented Generation for Answering Biomedical Questions

September 9, 2024

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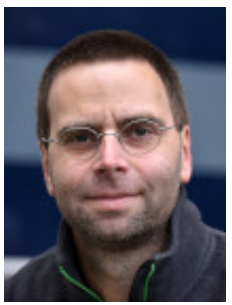
Alexander Bondarenko



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Medical Q&A

Example: What are active ingredients of Trikafta?

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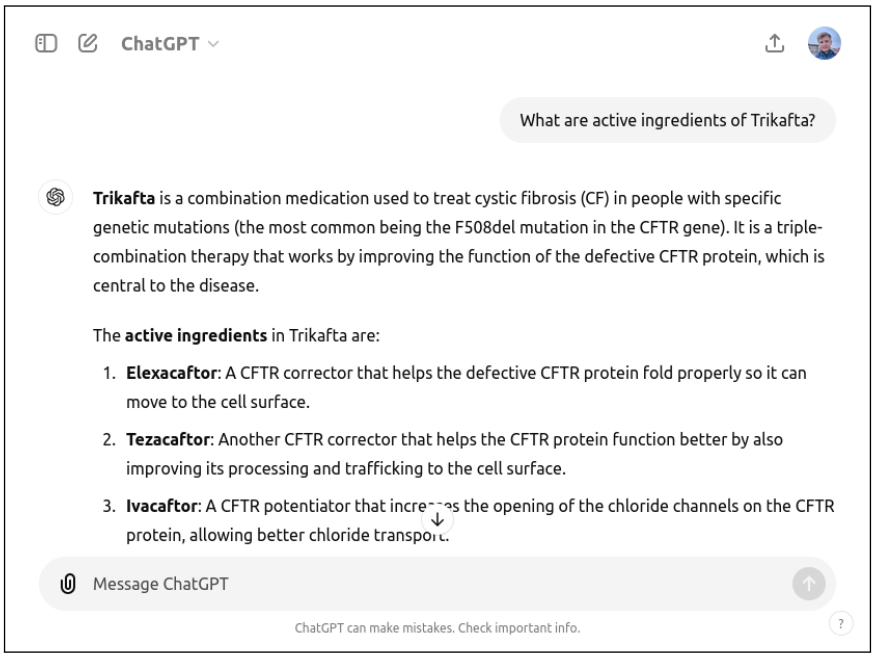
Baselines

Example: What are active ingredients of Trikafta?



Why not just use GPT . . . ?

(correct ingredients, no dosage, no sources)

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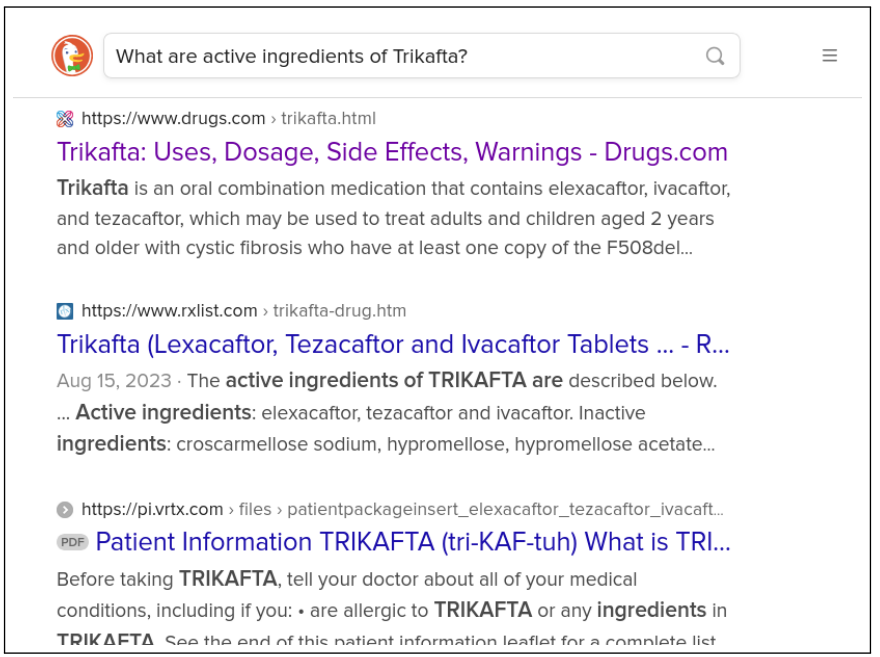
Baselines

Example: What are active ingredients of Trikafta?



. . . or a quick web search . . . ?

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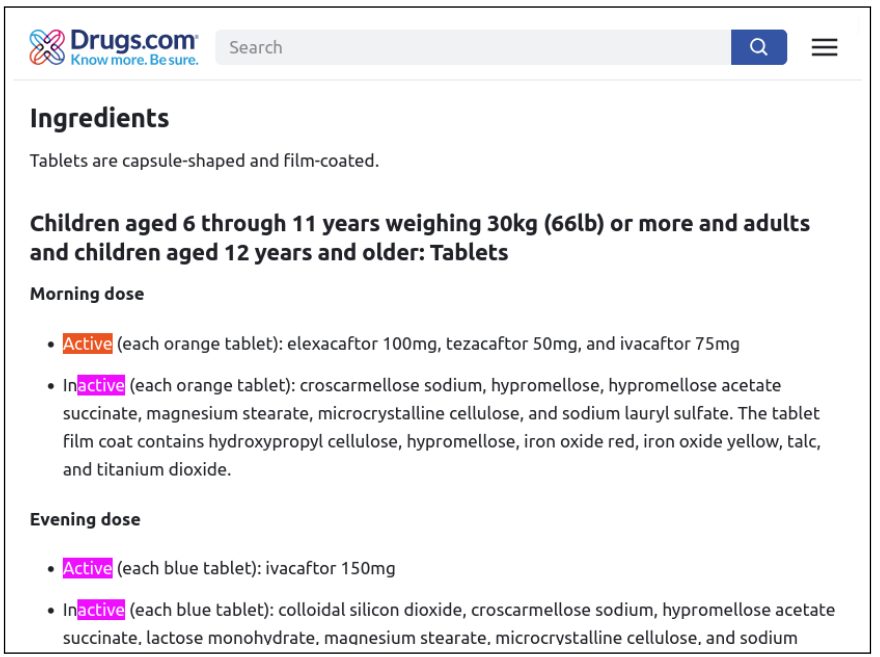
Baselines

Example: What are active ingredients of Trikafta?



🌐 . . . with Ctrl+F on the first result?

(correct ingredients and dosage, good source, but takes longer)

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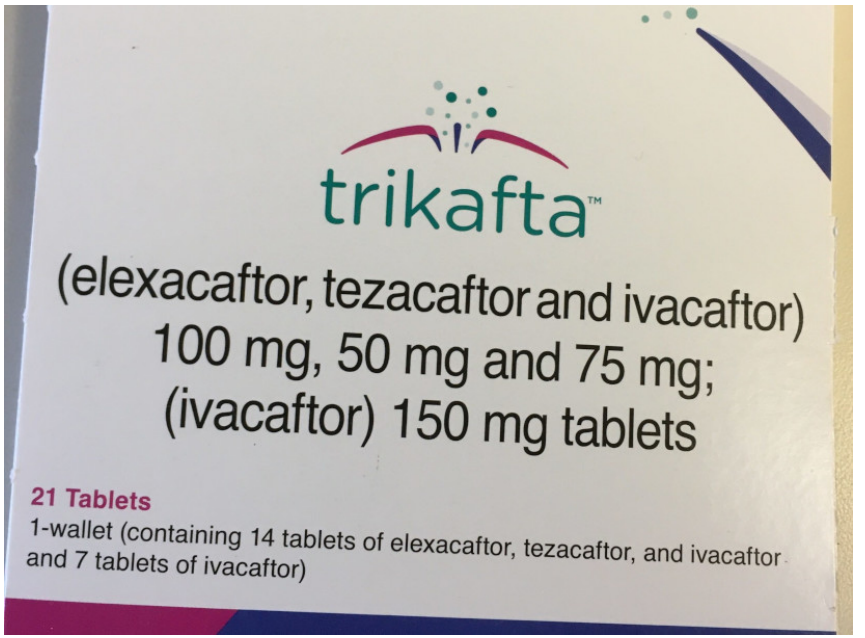
Baselines

Example: What are active ingredients of Trikafta?



😳 . . . and what about this?

(correct ingredients and dosage, good source, fastest?!)

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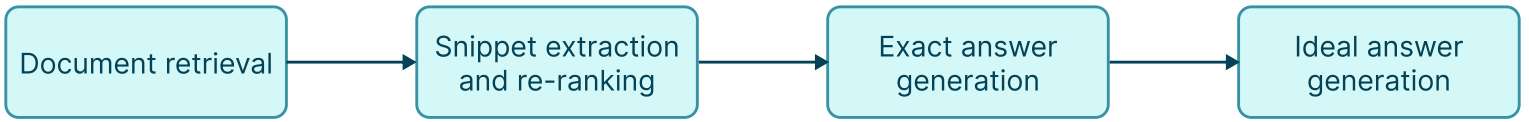
Goal: RAG for Medical Questions



Stages

- Document retrieval → Find relevant medical articles (from PubMed).
- Snippet extraction and re-ranking → Extract snippets and rank by relevance.
- Answer generation → Generate exact answer and “ideal” summary answer.
- RAG → Combine retrieval- and generation-focused components.

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Approaches: Document Retrieval

Goal: Find relevant medical articles (from PubMed).

- PubMed search API
  - Re-rank with BM25, MiniLM, and MPNet
- Custom BM25 index with metadata (Elasticsearch)
  - Match abstract, title, and MeSH terms
  - Disallow non-peer-reviewed publication types

→ Do we need to wory about indexing?

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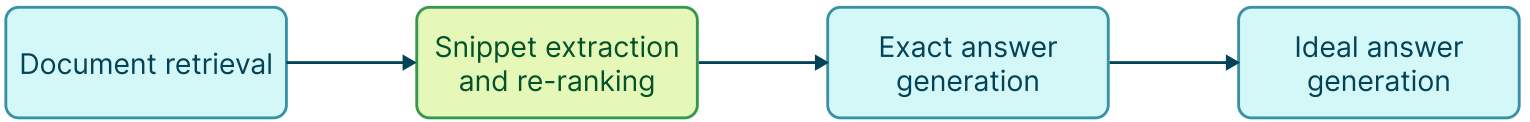
Approaches: Snippet Extraction and Re-Ranking



Goal: Extract concise snippets from the article's abstract (or title). Rank extracted snippets by relevance to the question.

- Using LLMs
  - Chain-of-thought 3-shot prompt (GPT-3.5-turbo)
  - No re-ranking
- Rule-based
  - Split abstract in sentences
  - Candidates: full title + sentence n-grams (up to 3 sent.) from abstract
  - Re-rank with TAS-B and duoT5

→ Are LLMs better at snippet extraction?

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Approaches: Answer Generation with LLMs

Goal: Generate exact (e.g., yes-no) answer and “ideal” summary answer.

- Few-shot prompting with function calling
  - Manual prompts per question/answer type (e.g., yes-no / exact)
  - Context: top-3 abstracts or all (top-10) snippets
  - GPT-3.5-turbo and GPT-4
- Modular “programming” with DSPy
  - Automatic prompts via DSPy (signature of in-/outputs are Python classes)
  - Context: abstracts, snippets, previous answer
  - Mixtral-7B

→ Do we need manual prompts?

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Approaches: RAG paradigms

Goal: Combine retrieval- and generation-focused components.

- Retrieve-then-generate (exact → ideal → documents → snippets)
- Generate-then-retrieve (documents → snippets → exact → ideal)
- GtRtG / RtGtR (e.g., exact → ideal → documents → snippets → exact → ideal)
- Let the LLM decide (DSPy, Mixtral-8x7B)

→ Which paradigm to use when?

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Results

- 42 submitted runs, different systems per phase and batch
- Retrieval:
  - PubMed search API struggles with question-like queries
  - Enhancing index with metadata pays off for domain-specific retrieval
- Snippet extraction and re-ranking:
  - Neither GPT- nor rule-based snippet extraction competitive
- Answer generation:
  - Snippets instead of abstracts as context less “confusing” for LLMs
  - Models: GPT-4 $>>$ GPT-3.5 $>$ Mixtral-7B
  - No difference by prompting strategy (manual vs. DSPy)
- RAG paradigm:
  - With ground truth: RtG, GtRtG
  - Without ground truth: GtR, GtRtG

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Summary

- Mixed results, but some recommendations:
  - Put work into first-stage retrieval
  - Show snippets to LLMs, not long texts
  - Use the latest LLMs (e.g., GPT-4)
  - GtRtG seems to work well with/without ground-truth evidence
- Limitation: comparability across test batches
- Future work: systematic evaluation of grounded RAG paradigms

Code and Data

`github.com/webis-de/CLEF-24`

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Summary

- Mixed results, but some recommendations:
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Thank you & merci!