

Retrieval Augmented Generation (RAG) for Biomedicine

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Large Language Models Hallucinate



Please list the most common symptoms associated with the long-term sequelae of COVID-19. Each symptom should list the source article title and PMID.

Large Language Models Hallucinate



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Sure, here are some of the most common symptoms associated with long-term sequelae of COVID-19, along with their corresponding source article title and PMID:

- Fatigue and Muscle Weakness: "Post-acute COVID-19 syndrome: a case series and multidisciplinary approach to treatment" (PMID: 33571309)
- Shortness of Breath: "Symptomatic Outcomes in Young Adults With COVID-19" (PMID: 32936950)
- 3. Joint Pain: "Musculoskeletal symptoms in COVID-19 patients" (PMID: 32767232)
- Chest Pain: "Chest CT Findings in Long-COVID Patients with Dyspnea and Chest Pain" (PMID: 33828099)



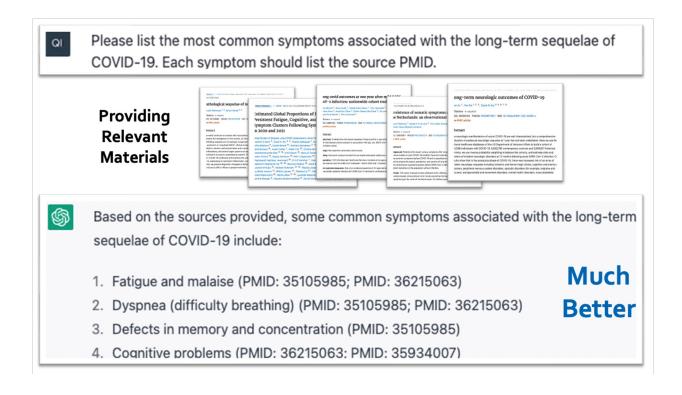
Looking Plausible

Large Language Models Hallucinate

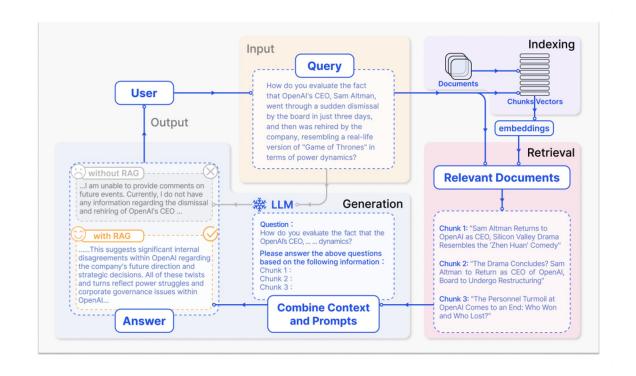
Please list the most common symptoms associated with the long-term sequelae of QI COVID-19. Each symptom should list the source article title and PMID. fabricated Sure, here are some of the most common symptoms associated with long-term sequelae of COVID-19, along with their corresponding source article title and PMID: 1. Fatigue and Muscle Weakness: "Post-acute COVID-19 syndrome: a case series and multidisciplinary approach to treatment" (PMID: 33571309) 2. Shortness of Breath: "Symptomatic Outcomes in Young Adults With COVID-19" (PMID: 32936950) 3. Joint Pain: "Musculoskeletal symptoms in COVID-19 patients" (PMID: 32767232) 4. Chest Pain: "Chest CT Findings in Long-COVID Patients with Dyspnea and Chest Pain" (PMID: 33828099)



Retrieval as a Solution to Hallucination



Retrieval-augmented Generation (RAG) Architecture



Medical Information Retrieval-Augmented Generation Evaluation

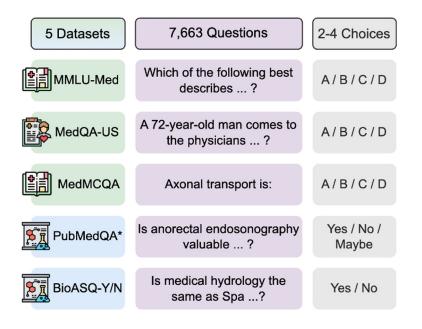


Figure 1: Composition of the MIRAGE benchmark.

MedRAG Toolkit

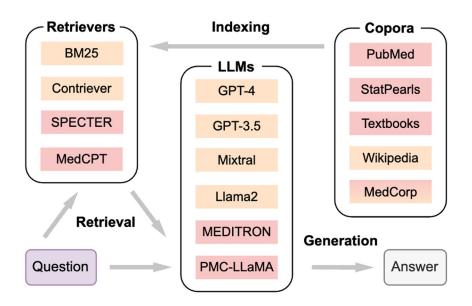
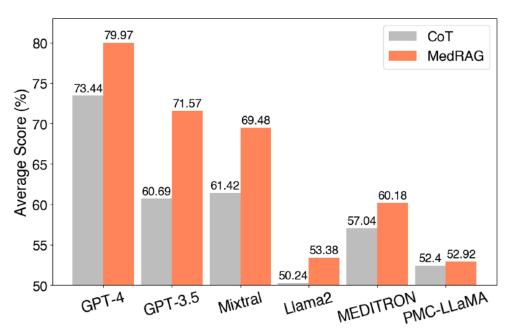


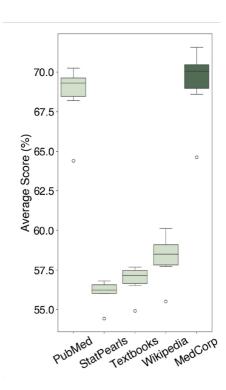
Figure 2: Component overview of the MEDRAG toolkit.

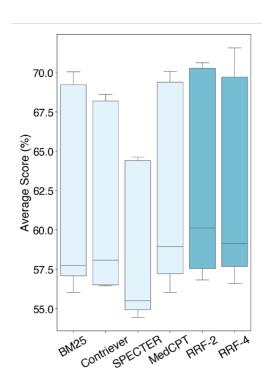
MedRAG Improves Medical QA Performance

On MIRAGE, MedRAG improves the accuracy of different LLMs by up to 18%



Effects of Corpora and Retrievers in MedRAG





(left) Corpora have more influence on performance

(right) Different retrievers get less different results

(overall) The combination of various medical corpora and retrievers achieves the best performance

MedRAG: Scaling Properties

Log-linear improvement with # retrieved snippets

Favoring long-context LLMs

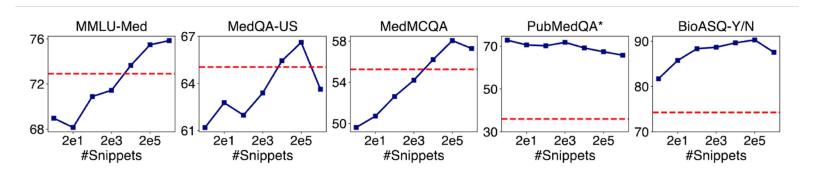
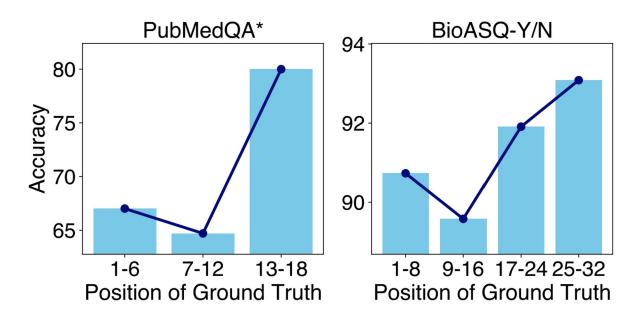


Figure 3: MEDRAG accuracy with different numbers of retrieved snippets. Red dotted lines denote CoT performance.

"Lost-in-the-middle"

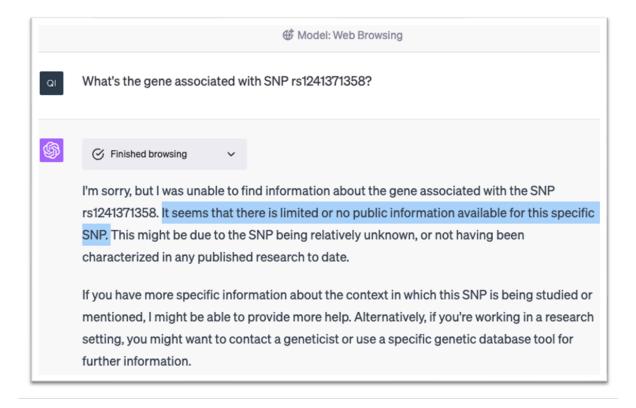
MedRAG performance is lowest when relevant article is in the middle of prompt



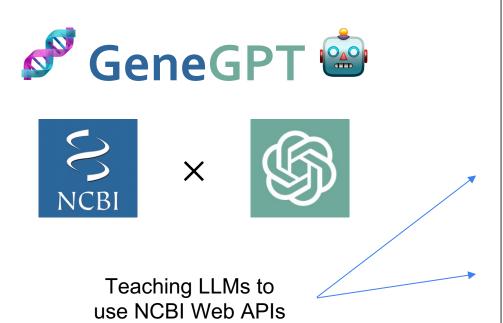
Coding & QA

https://colab.research.google.com/drive/1HoYWH0yJBQliWC_XRKL-COwfdbXTCDUj?usp=sharing

RAG is **NOT** Enough



GeneGPT: First Tool-augmented LLM in Biomedicine



GeneHop question (SNP gene function):

What is the function of the gene associated with SNP rs1241371358? Let's decompose the question to sub-questions and solve them step by step.

GeneGPT output:

Sub-question 1: Which gene is SNP rs1241371358 associated with?

[https://eutils.ncbi.nlm.nih.gov/ entrez/eutils/esummary.fcgi?db=snp& retmax=10&retmode=json&id=1241371358]

->[API call results]
Answer: LRRC23

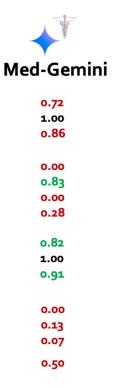
Sub-question 2: What is the function of LRRC23? [https://eutils.ncbi.nlm.nih.gov/entrez/eutils/esummary.fcgi?db=gene&retmax=10&retmode=json&id=10233]

->[API call results]

Answer: Predicted to be active in cytosol. (correct)

GeneGPT: First Tool-augmented LLM in Biomedicine

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0.00				ChatGPT	New Bing	-full	-slim
0.00							
	0.00	0.04	0.09	0.07	0.66	0.80*	0.84*
0.00	0.00	0.00	0.00	0.00	0.85	1.00	1.00
0.00	0.00	0.02	0.05	0.04	0.76	<u>0.90</u>	0.92
0.00	0.00	0.00	0.00	0.00	0.00	1.00*	1.00
0.01	0.04	0.12	0.09	0.09	0.61	0.62	0.66
0.03	0.05	0.01	0.02	0.05	0.01	$\overline{1.00}$	0.98
0.01	$\overline{0.03}$	0.04	0.04	0.05	0.21	<u>0.87</u>	0.88
0.00	0.02	0.16	0.34	0.31	0.84	0.76*	0.66
0.00	0.18	0.37	0.70	0.54	0.97	0.76	1.00
0.00	0.10	0.27	0.52	0.43	0.91	0.76	<u>0.84</u>
0.02	0.07	0.03	0.00	0.00	0.00	0.44*	0.44*
0.02	$\overline{0.00}$	0.00	0.20	0.00	0.00	0.86	0.88
0.02	0.04	0.02	0.10	0.00	0.00	0.65	0.66
0.00	0.04	0.08	0.16	0.12	0.44	0.80	0.83
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Thank you!

Questions?