# Beginner's Guide to RAG Evaluation

Al by Hand



Prof. Tom Yeh

Hosted by:





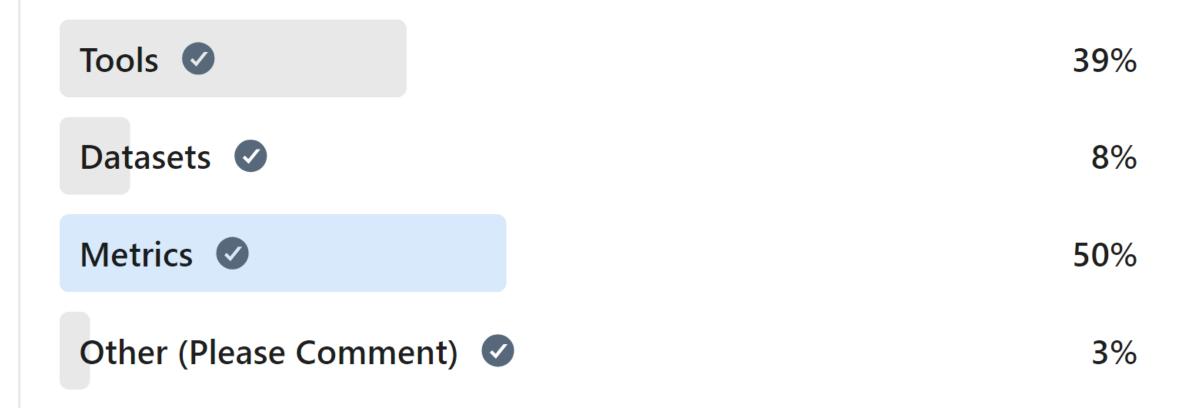
# Poll

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## What do you want to learn about RAG Evaluation?

You can see how people vote. Learn more



265 votes • Poll closed

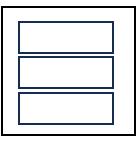
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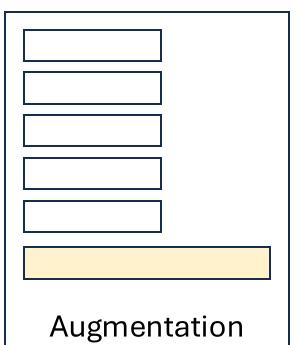
**User Question** 

What pets?

Answer



Retriever

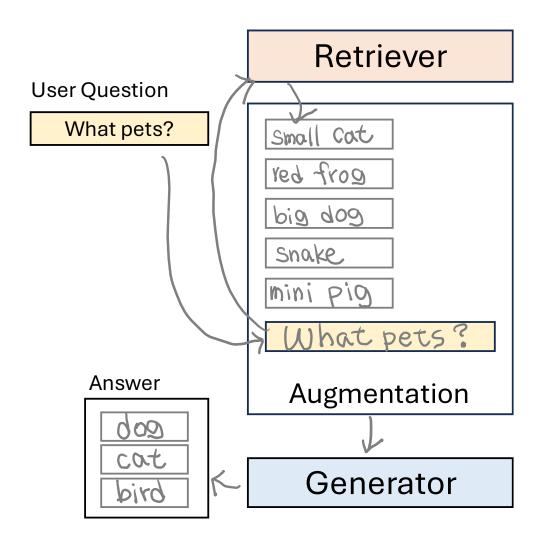


Generator

# Trace

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## Trace

Retrieval

Prompt Embedding

Vector Store

Context Encode

Generation



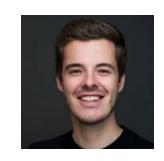


# Langfuse

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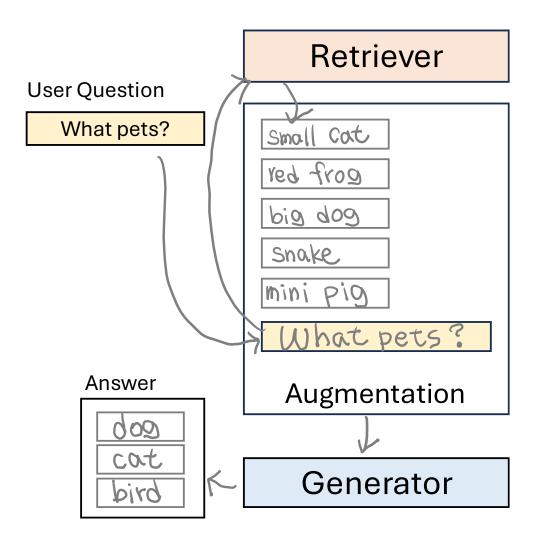


Special Thanks



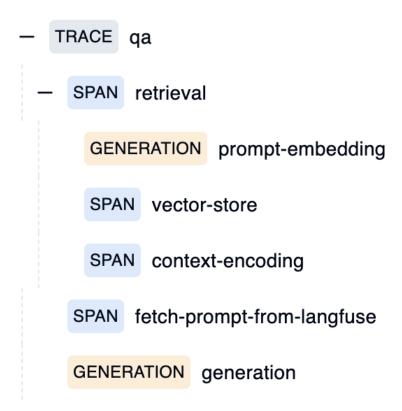
Marc Klingen Co-founder

Download: https://by-hand.ai/rageval





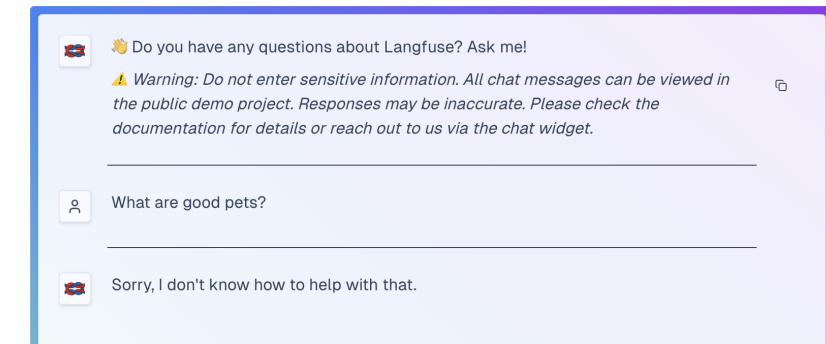
#### Trace Timeline 🖽 🗆

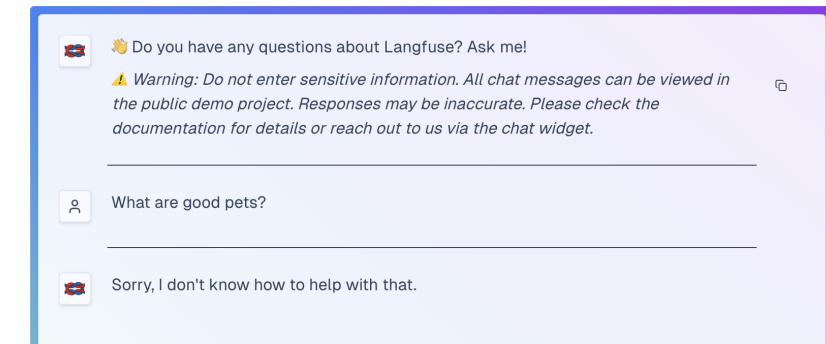


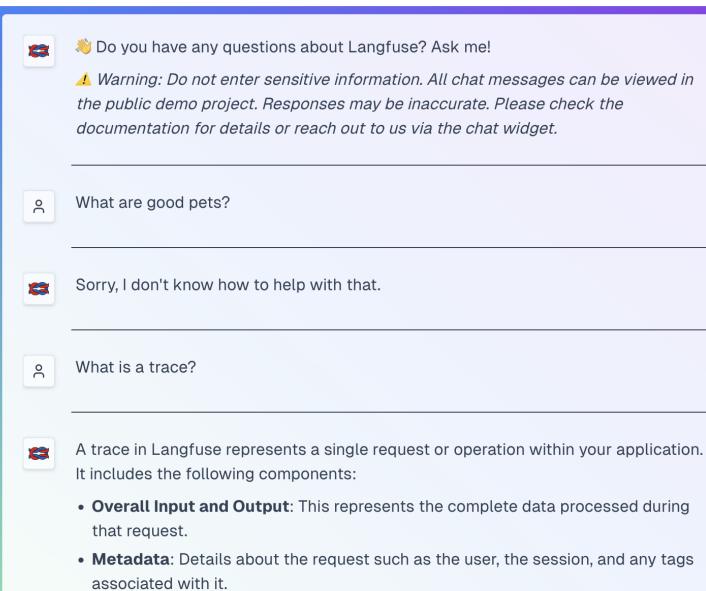


No you have any questions about Langfuse? Ask me!

⚠ Warning: Do not enter sensitive information. All chat messages can be viewed in the public demo project. Responses may be inaccurate. Please check the documentation for details or reach out to us via the chat widget.

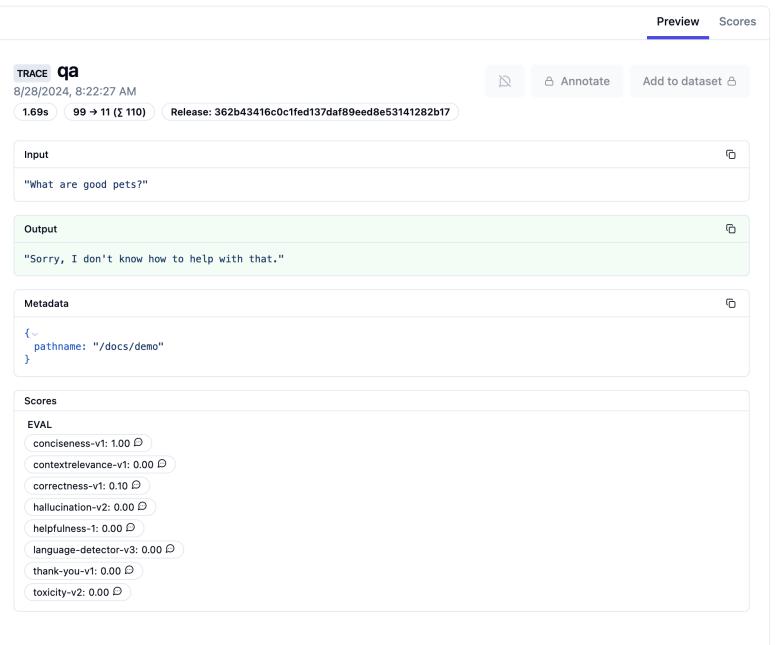


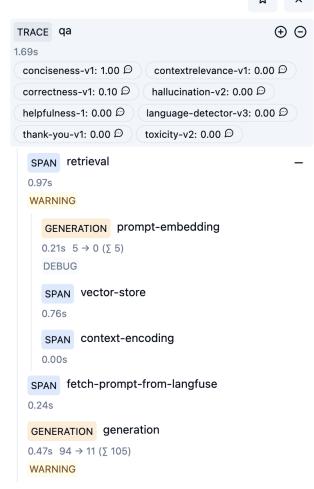


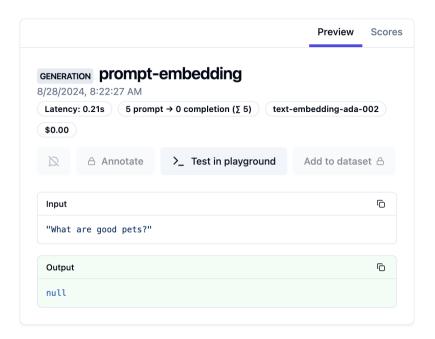


#### Traces <sup>©</sup>

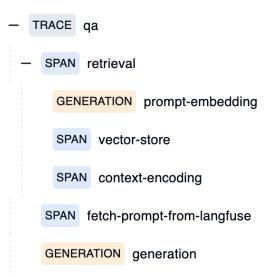
Search	by id, name, user id	Q 7	₹ Filter © 24 hours ∨						
	ID	User ①	Usage	# conciseness-v1 (eval)	# helpfulness-1 (eval) # toxicity-v2 (eval)		Total Cost	Tags ① Action	
	7c973011-43	u-SoAo6Be	943 → 212 (∑ 1,155)	0.2000 🖾	1.0000 🗩	0.0000 €	\$0.0003	with-context	Û
	fc8a6acd-7d	u-SoAo6Be	99 → 11 (Σ 110)	1.0000 🗩	0.0000 ₽	0.0000 €	\$0.00	no-context	Û
	3e77f6c2-74	u-9qRwS0z	159 → 11 (∑ 170)	1.0000 🗩	0.0000 €	0.0000 €	\$0.00	no-context	Û
	eebb663a-a0	u-9qRwS0z	136 → 11 (∑ 147)	0.0000 €	0.0000 €	0.0000 €	\$0.00	no-context	Û
	be3e1a0f-1b	u-9qRwS0z	108 → 11 (Σ 119)	1.0000 🗩	0.0000 €	0.0000 €	\$0.00	no-context	Û
	9e09f16c-51	u-K8t215e	148 → 11 (∑ 159)	1.0000 🖾	0.0000 ₽	0.0000 ₽	\$0.00	no-context	Û
	c7fcfb43-d9	u-K8t215e	740 → 11 (∑ 751)	1.0000 🖾	0.0000 €	0.0000 €	\$0.0001	with-context	Û
	16f76d22-17	u-K8t215e	358 → 11 (∑ 369)	1.0000 🖾	0.0000 ₽	0.0000 €	\$0.0001	with-context	Û
	f31a7481-e4f	u-2PkSSae	97 → 30 (Σ 127)	0.2000 ⊅	1.0000 🖾	0.0000 €	\$0.00	no-context	Û
	0b76d3e9-b	u-0ojLo4C	483 → 36 (∑ 519)	0.1000 ⊅	0.2000 ₽	0.0000 🗩	\$0.0001	with-context	Û
	75aecee5-04	u-0ojLo4C	389 → 32 (∑ 421)	0.0000 €	0.2000 ₽	0.0000 🗩	\$0.0001	with-context	Û
	d7c311a6-eb	u-Ux4V8EW	4,014 → 729 (∑ 4,743)	0.2000 🖾	1.0000 🖾	0.0000 €	\$0.001	with-context	Û
	19acd92c-78	u-Ux4V8EW	3,001 → 398 (Σ 3,399)	0.2000 🖾	1.0000 🖾	0.0000 €	\$0.0007	with-context	Û
	81d0b569-6	u-Ux4V8EW	2,244 → 333 (∑ 2,577)	0.9000 ⊅	1.0000 🗩	0.0000 €	\$0.0005	with-context	Û
	07e35225-67	u-Ux4V8EW	2,845 → 390 (∑ 3,235)	0.4000 ⊅	1.0000 🗩	0.0000 🖻	\$0.0007	with-context	Û
	cca2662c-bc	u-Ux4V8EW	1,548 → 464 (Σ 2,012)	0.1000 ⊅	1.0000 🗩	0.0000 €	\$0.0005	with-context	Û
	37ff7b0b-6a	u-Ux4V8EW	1,577 → 56 (Σ 1,633)	0.1000 ⊅	0.9000 ⊅	0.0000 🖻	\$0.0003	with-context	Û
	d0948efe-d1	u-Ux4V8EW	864 → 11 (∑ 875)	1.0000 🗩	0.0000 ⊅	0.0000 🗩	\$0.0001	no-context	Û
	b1467a5a-fd	u-Ux4V8EW	1,831 → 348 (Σ 2,179)	0.2000 ፟፟	1.0000 🖾	0.0000 🗩	\$0.0005	with-context	Û
	269591b4-e	u-Ux4V8EW	1,510 → 365 (∑ 1,875)	0.2000 🖾	1.0000 🖾	0.0000 🗩	\$0.0004	with-context	Û
	bbce4c6c-84	u-GtpTQrm	870 → 232 (∑ 1,102)	0.9000 ፟	1.0000 🖾	0.0000 🗩	\$0.0003	with-context	Û
	5bc36c39-d	u-B6kKyPu	886 → 276 (∑ 1,162)	0.2000 🖾	1.0000 🖾	0.0000 🗩	\$0.0003	with-context	Û
	0fb2c51c-d9	u-oC3aYkF	504 → 40 (∑ 544)	0.0000 🖾	0.0000 €	0.0000 🗩	\$0.0001	with-context	Û
	d72b106b-11	u-Hmj2f7d	89 → 11 (∑ 100)	1.0000 🗩	0.0000 €	0.0000 🗩	\$0.00	no-context	Û
	77317131-50f	u-0b6oEKx	91 → 11 (∑ 102)	1.0000 🗩	0.0000 €	0.0000 🗩	\$0.00	no-context	Û
□ ☆	26efa5c0-04	u-pX1dxFk	1,298 → 392 (Σ 1,690)	0.9000 ⊅	0.9000 €	0.0000 🖾	\$0.0004	with-context	Û



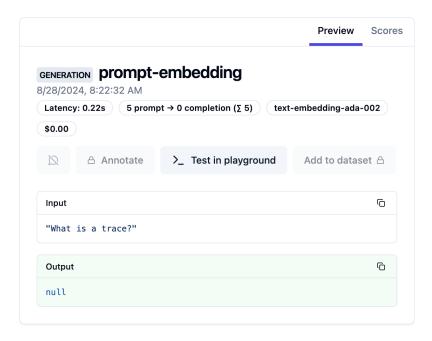


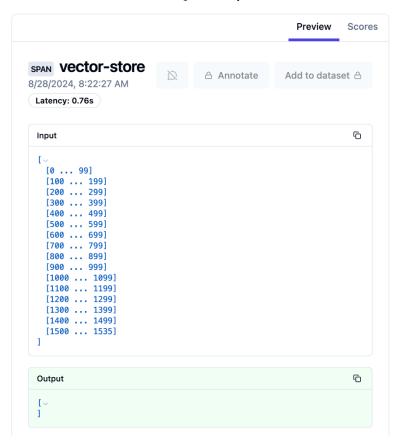


#### **Trace Timeline □** □

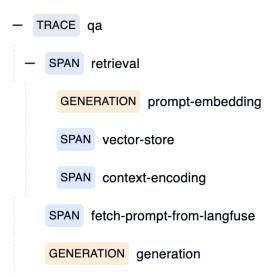


#### "What is a trace?"

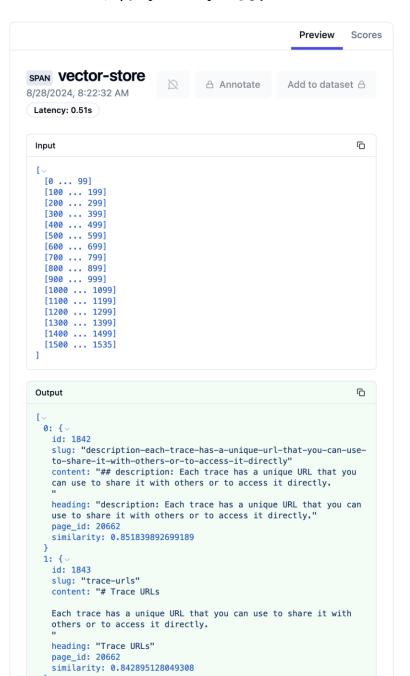


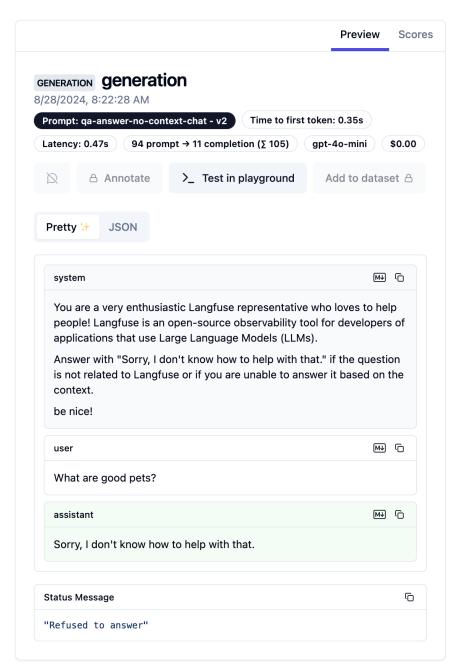


#### Trace Timeline ⊕ □

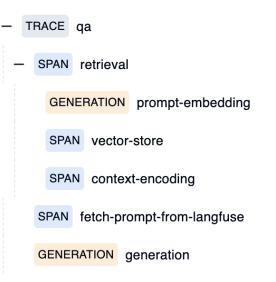


#### "What is a trace?"

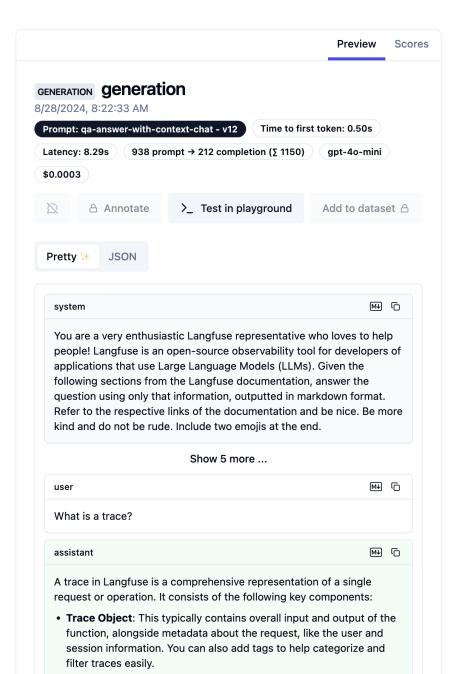


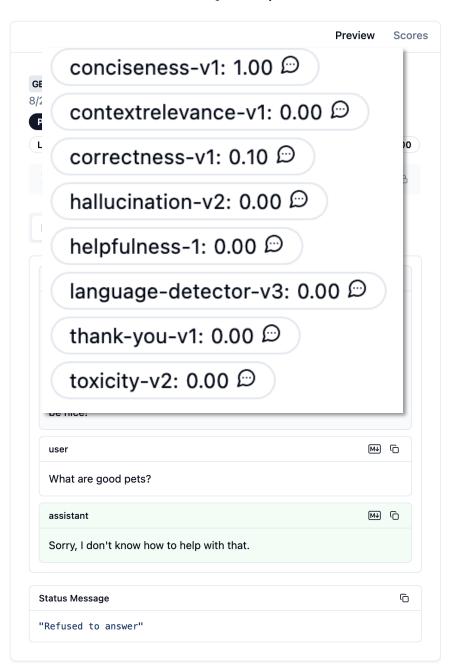


#### **Trace Timeline** ⊕ □

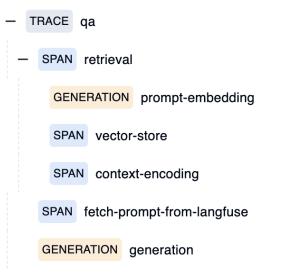


#### "What is a trace?"



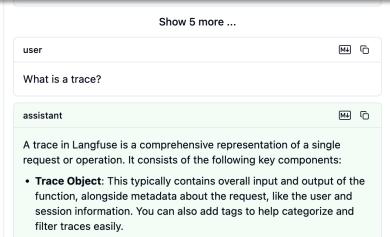


#### **Trace Timeline** ⊕ □



"What is a trace?"

conciseness-v1: 0.20 🗩 contextrelevance-v1: 1.00 € correctness-v1: 0.90 @ hallucination-v2: 0.00 @ helpfulness-1: 1.00 🖾 language-detector-v3: 1.00 🕮 thank-you-v1: 0.00 🗩 toxicity-v2: 0.00 🗩 Show 5 more ...



## Conciseness

Evaluate the \_\_\_\_\_ of the generation on a continuous scale from 0 to 1. A generation can be considered concise (Score: 1) if it directly and succinctly answers the question posed, focusing specifically on the information requested without including unnecessary, irrelevant, or excessive details.

#### conciseness

#### Prompt

Evaluate the conciseness of the generation on a continuous scale from 0 to 1. A generation can be considered concise (Score: 1) if it directly and succinctly answers the question posed, focusing specifically on the information requested without including unnecessary, irrelevant, or excessive details.

#### Example:

Query: Can eating carrots improve your vision? Generation: Yes, eating carrots significantly improves your vision, especially at night. This is why people who eat lots of carrots never need glasses. Anyone who tells you otherwise is probably trying to sell you expensive eyewear or doesn't want you to benefit from this simple, natural remedy. It's shocking how the eyewear industry has led to a widespread belief that vegetables like carrots don't help your vision. People are so gullible to fall for these money-making schemes.

Score: 0.3

Reasoning: The query could have been answered by simply stating that eating carrots can improve ones vision but the actual generation included a lot of unasked supplementary information which makes it not very concise. However, if present, a scientific

You can use {{variable}} to insert variables into your prompt. Note: Variables must be alphabetical characters or underscores. The following variables are available:

generation query

#### Score

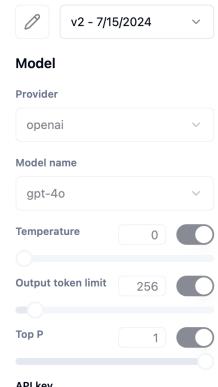
provide a score between 0 and 1

We use function calls to extract data from the LLM. Specify what the LLM should return for the score.

#### Reasoning

provide a one sentence reasoning

We use function calls to extract data from the LLM. Specify what the LLM should return for the reasoning.



#### API key

...vTlu

The LLM API key is used for each execution and will incur costs.

# Integration





```
from llama_index.core import Settings

from llama_index.core.callbacks import CallbackManager

from langfuse.llama_index import LlamaIndexCallbackHandler

Settings.callback manager = CallbackManager([LlamaIndexCallbackHandler()])
```





# RAGAS - RAG AS sessment

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Special Thanks



Shahul ES Founder

## ragas score

generation

#### faithfulness

how factually acurate is the generated answer

### answer relevancy

how relevant is the generated answer to the question retrieval

#### context precision

the signal to noise ratio of retrieved context

#### context recall

can it retrieve all the relevant information required to answer the question

## ragas score

generation

#### faithfulness

how factually acurate is the generated answer

#### answer relevancy

how relevant is the generated answer to the question retrieval

#### context precision

the signal to noise ratio of retrieved context

#### context recall

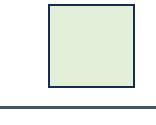
can it retrieve all the relevant information required to answer the question

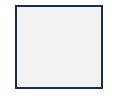
#### Retriever **User Question** What pets? Small Coxt gorf bar 60p 6ig snake mini Pig What pets? Answer Augmentation gop cat Generator bird

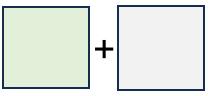
## Generator Metrics

Faithfulness

Hallucination









# Faithfulness

Your task is to judge the \_\_\_\_\_\_ of a series of statements based on a given \_\_\_\_\_. For each statement you must return verdict as 1 if the statement can be directly inferred based on the context or 0 if the statement can not be directly inferred based on the context.

## ragas score

generation

### faithfulness

how factually acurate is the generated answer

## answer relevancy

how relevant is the generated answer to the question

retrieval

#### context precision

the signal to noise ratio of retrieved context

#### context recall

can it retrieve all the relevant information required to answer the question

#### Retriever **User Question** What pets? small cat gorf bar 60p 6ig snake mini Pig What pets? Answer Augmentation 900 cat Generator bird

## Generator Metrics

Answer Relevancy

What pets?

What are common pets?

What are domesticated animals?

What are small animals?

 $\frac{\sum}{\frac{1}{N}}$ 

## Question Generation

Generate a \_\_\_\_\_\_ for the given answer and Identify if answer is \_\_\_\_\_\_. Give noncommittal as 1 if the answer is noncommittal and 0 if the answer is committal. A noncommittal answer is one that is evasive, vague, or ambiguous. For example, "I don't know" or "I'm not sure" are noncommittal answers"

## ragas score

generation

#### faithfulness

how factually acurate is the generated answer

## answer relevancy

how relevant is the generated answer to the question retrieval

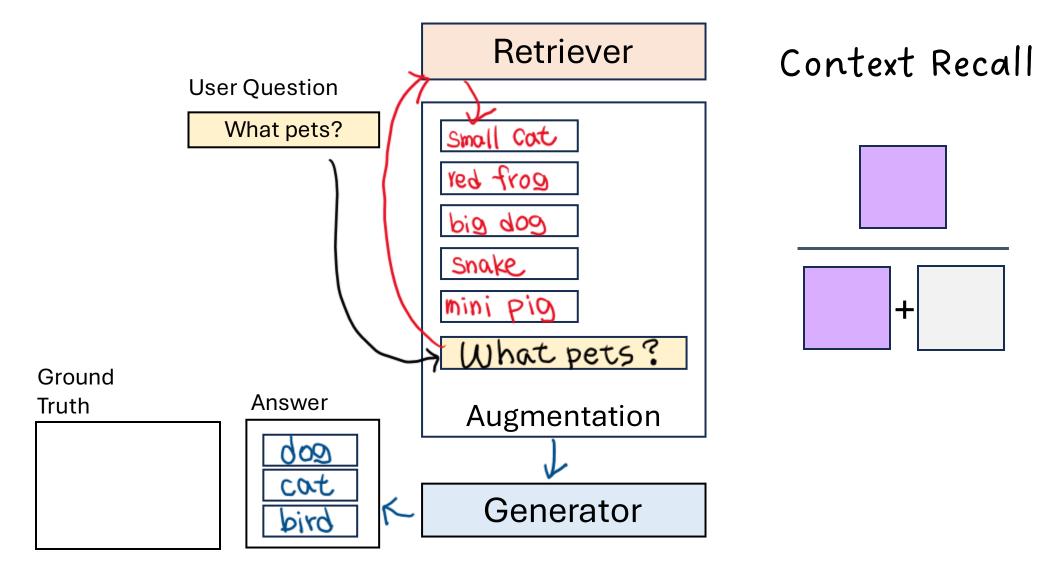
#### context precision

the signal to noise ratio of retrieved context



can it retrieve all the relevant information required to answer the question

## Retriever Metrics



## ragas score

generation

#### faithfulness

how factually acurate is the generated answer

### answer relevancy

how relevant is the generated answer to the question retrieval

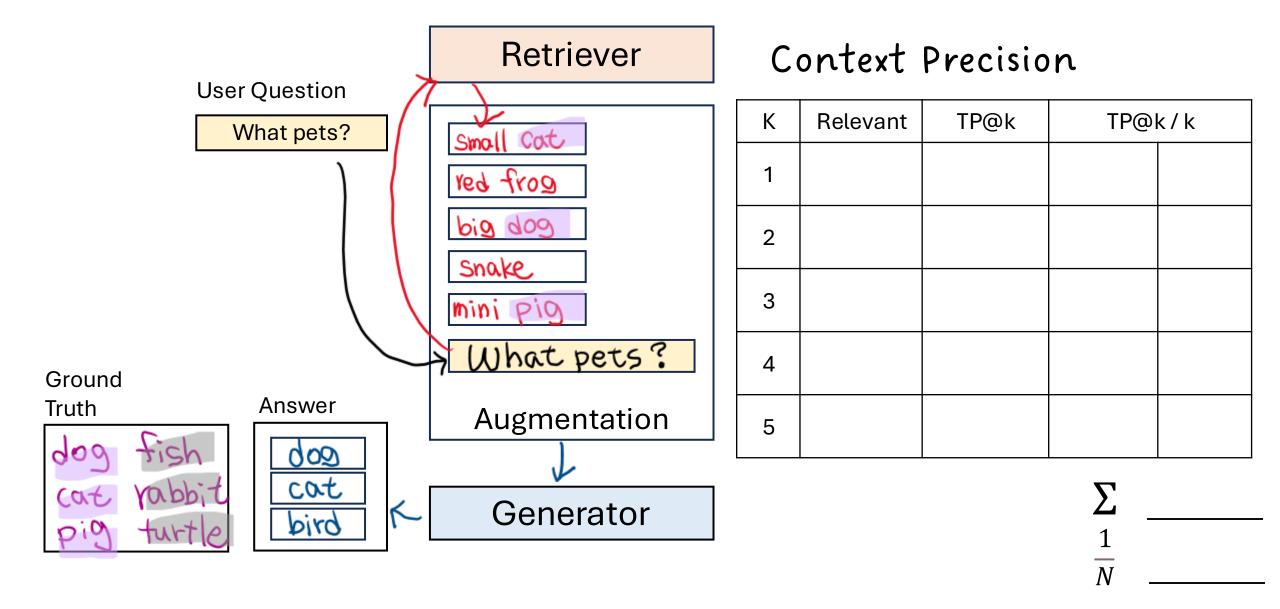
### context precision

the signal to noise ratio of retrieved context

#### context recall

can it retrieve all the relevant information required to answer the question

## Retriever Metrics



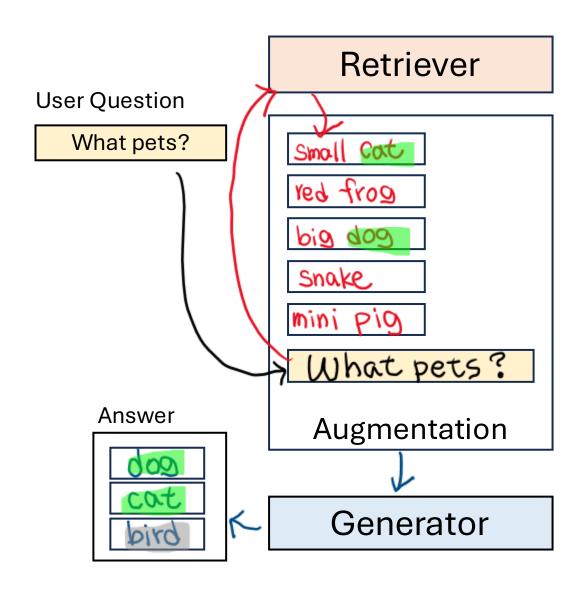
## Precision

Given question, answer and context verify if the context was \_\_\_\_\_ in arriving at the given answer. Give verdict as "1" if useful and "0" if not with json output.

ragas/src/ragas/metrics/\_context\_precision.py at main · explodinggradients/ragas · GitHub

# Cost

# Where are LLMs used for Generator metrics?

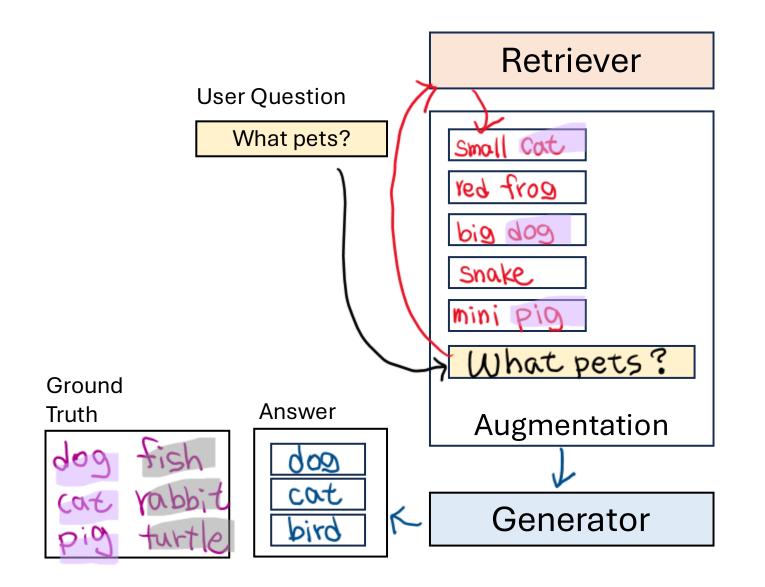


Retrieval

Generation

**Evaluation** 

## Where are LLMs used for Retriever metrics?



Retrieval

Generation

**Evaluation** 

# Many More Metrics

- · Context Utilization
- Context Entities Recall
- Noise Sensitivity
- Answer Semantic Similarity
- Answer Correctness
- Aspect Critique
- Domain Specific Evaluation
- Summarization Score

... etc.

	Target?	LLM Eval?	Ground Truth?	Range?	Best?
Faithfulness					
Answer Relevancy					
Context Recall					
Context Precision					
Context Utilization					
Context Entities Recall					
Noise Sensitivity					
Answer Semantic Similarity					
Answer Correctness					
Aspect Critique					
Domain Specific Evaluation					
Summarization Score					

Q/A