



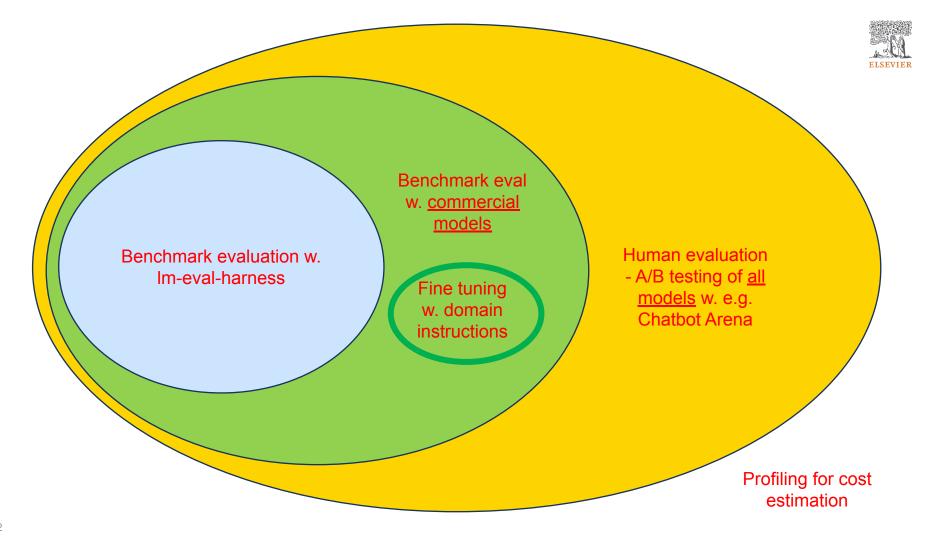
NLP4Chemistry

Exploring the Fringe

Camilo Thorne, Saber Akhondi

May 2024 - COLING/LREC '24







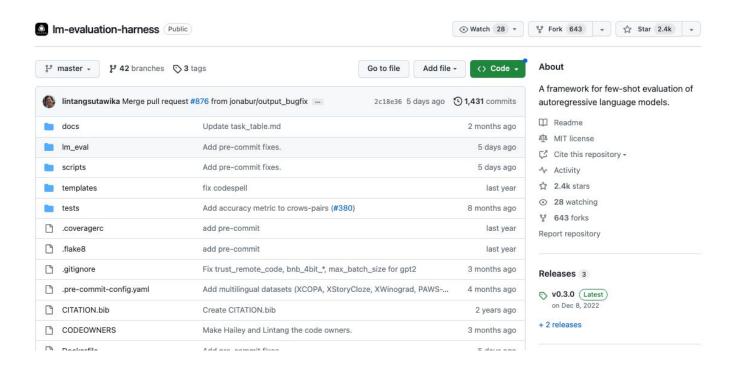
Benchmark Evaluation



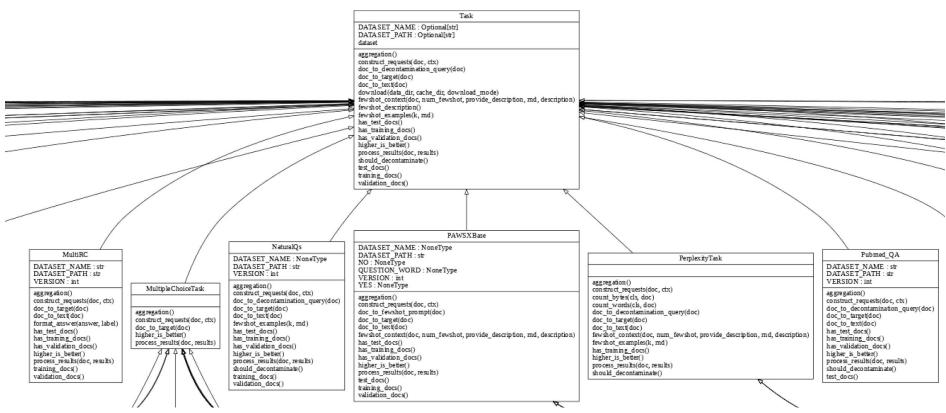
Benchmark evaluation w. lm-eval-harness [2022]



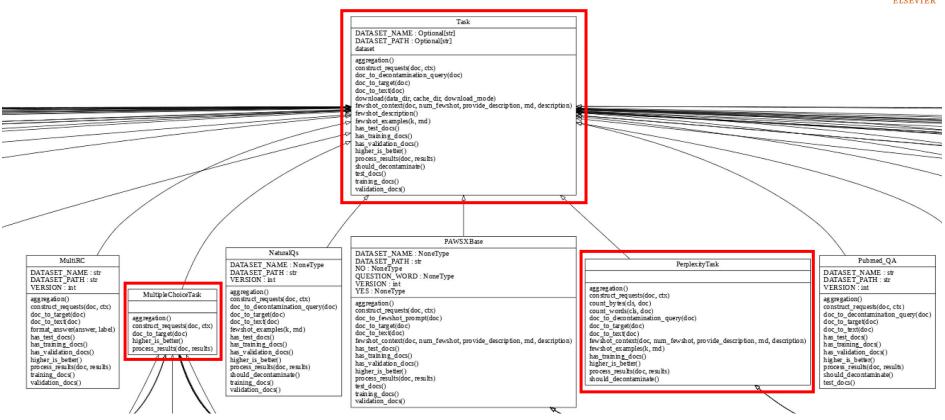
https://github.com/EleutherAI/lm-evaluation-harness











```
import datasets
from lm_eval.base import PerplexityTask
from lm_eval.utils import escaped_split
class ICSRPerplexity(PerplexityTask):
   We re-use here the ICSR data, but we throw away the labels.
   We ask the model to generate the abstracts, and measure LM perplexity.
   VERSION = "0.1.0"
   DATASET PATH = "icsr"
   DATASET NAME = "icsr-perplexity"
   def __init (self, data dir=None, cache dir=None, download mode=None):
        self.data dir = data dir
        self.dataset = datasets.load dataset(
            path=self.DATASET_PATH,
           name=self.DATASET_NAME,
           data dir=data dir.
           cache_dir=cache_dir,
            download mode=download mode,
           field="data",
       self. training docs = None
        self. fewshot docs = None
   def download(self, data_dir=None, cache_dir=None, download_mode=None):
        raise TypeError("cannot download an arbitrary JSON dataset")
   def has validation docs(self):
        return False
   def has training docs(self):
        return False
   def has test docs(self):
        return True
   def test docs(self):
        return map(self._process_doc, self.dataset["test"])
   def process doc(self, doc):
       return doc['Abstract']
```



- Extend task (sub) class
- Implement custom constructor
- Implement or override super-class methods



```
MODEL_REGISTRY = {
    "hf":
                               gpt2.HFLM,
    "hf-causal":
                               gpt2.HFLM,
    "hf-causal-experimental": huggingface.AutoCausalLM,
    "hf-seq2seq":
                               huggingface.AutoSeq2SeqLM,
    "gpt2":
                               gpt2.GPT2LM,
    "gpt3":
                               gpt3.GPT3LM,
    "anthropic":
                               anthropic llms.AnthropicLM,
    "textsynth":
                               textsynth.TextSynthLM,
    "dummy":
                               dummy.DummyLM,
```

Library supports:

- HF-supported decoder-only models + architectures
- HF-supported encoder-decoder models + architectures
- OpenAl legacy models (GPT3, GPT4)
- Anthropic Claude v1

Datasets need to be formatted into HF Dataset JSON format

Models and datasets by default fetched from HF remote servers

But they can also be deployed locally, provided HF formats are satisfied



lm-eval-harness leaderboard [2022-]



https://huggingface.co/spaces/HuggingFaceH4/open llm leaderboard

Open LLM Leaderboard LLM Benchmark Metrics through time About FAO Submit Model types Search models or licenses (e.g., 'model_name; license: MIT') and press ENTER... continuously pretrained fine-tuned on domain-specific datasets Select columns to show base merges and moerges Average 🔝 TruthfulQA ✓ Winogrande Precision Architecture **Hub License** Model sha ✓ bfloat16 Hide models Model sizes (in billions of parameters) Private or deleted Contains a merge/moerge ✓ Flagged ▲ Model Average 🔝 ▲ HellaSwag ▲ GSM8K A TruthfulQA Winogrande zhengr/MixTAO-7Bx2-MoE-v8.1 77.5 89.22 78.57 87.37 71.11 73.81 yunconglong/Truthful DPO TomGrc FusionNet 7Bx2 MoE 13B 77.44 74.91 89.3 64.67 78.02 88.24 69.52 yunconglong/DARE_TIES_13B 77.1 74.32 89.5 64.47 78.66 88.08 67.55 77.08 89.51 yunconglong/13B_MATH_DPO 74.66 64.53 78.63 88.08 67.1 yunconglong/MoE 13B DPO 77.05 74.32 89.39 64.48 78.47 67.63 yam-peleg/Experiment26-7B 76.74 89.15 84.93 70.43 73.38 64.32 78.24 MTSAIR/multi_verse_model } 76.74 72 87 89.2 64.4 77.92 84.77 71 27 chihoonlee10/T30-Mistral-Orca-Math-DPO 76.7 72.95 89.23 64.42 78.41 84.93 70.28 yam-peleg/Experiment26-78 76.67 89.12 64.3 78.04 85 70.43 73.12 rwitz/experiment26-truthy-iter-0 76.65 73 29 89.11 64.35 77.86 84.93 70.36 nbeerbower/bophades-mistral-truthy-DPO-7B 76.63 73.38 89.28 64.73 77.88 85.16 69.37 vam-peleg/Experiment30-7B 76.62 73.38 89.13 64.28 77.98 84.93 70.05

ChemLLM [2023]



- Open release of chemistry LLM evaluation benchmarks
- "Promptified" version of known benchmarks
 - USPTO-50k
 - PubChem

O ...

Ability	Task	Task Type	Dataset	#ICL candidates	#test	Evaluation Metrics
Understanding	Name Prediction	Generation	PubChem	500	100	Accuracy
	Property Prediction	Classification	BBBP, HIV, BACE, Tox21, ClinTox	2053, 41127, 1514, 8014, 1484	100	Accuracy, F1 score
Reasoning	Yield Prediction	Classification	Butchward-Hartwig, Suzuki-Miyaura	3957, 5650	100	Accuracy
	Reaction Prediction	Generation	USPTO-Mixed	409035	100	Accuracy, Validity
	Reagents Selection	Ranking	Suzuki-Miyaura	5760	100	Accuracy
	Retrosynthesis	Generation	USPTO-50k	40029	100	Accuracy, Validity
	Text-Based Molecule Design	Generation	ChEBI-20	26407	100	BLEU, Exact Match, etc
Explaining	Molecule Captioning	Generation	ChEBI-20	26407	100	BLEU, Chemists, etc

https://github.com/ChemFoundationModels/ChemLLMBench

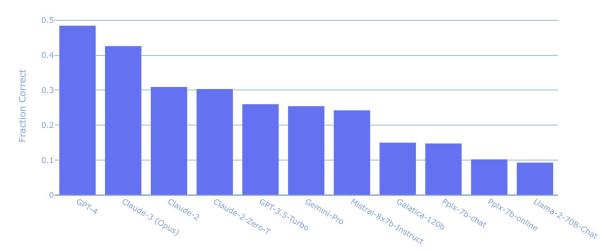
ChemBench [2024]



- Similar to ChemLLM, but focused on pharmaceutical domain
- Evaluation of open source models + leaderboard

https://lamalab-org.github.io/chem-bench/leaderboard/

Relies on completion "accuracy" for benchmarking!



Challenges of metrics



Benchmark metrics focus too much on lexical match, at the expense of topical relevancy!

doc_id: 3

prompt_0: "Question: Pyridine synthesis via hetero-Diels-Alder reaction of 1,2,4-triazines and dienophiles (e.g. enamine) followed by extrusion of N2 What is the name of the reaction? Answer: Boger Thermal Cycloaddition Question: How to synthesis L-valeric acid? Answer: The Marckwald Asymmetric Synthesis is a well-known reaction to synthesise L-valeric acid. It's a Chiral synthesis of L-valer ic acid by pyrolysis of brucine salt of racemic α-methyl-α-ethylmalonic acid Question: What is a Clauson-Kaas Synthesis? Answer: Synthesis of pyrrole derivative by the condensation between a prima ry aliphatic or aromatic amine and 2,5-dialkoxy-tetrahydrofuran in the presence of an acid catalyst Question: Which reaction type is able to convert trialkyl phosphite and alkyl halide to phosphon ate? Answer: until"

logit_0: "The reaction type that is able to convert trialkyl phosphite and alkyl halide to phosphonate is a Wacker reaction"

truth: " The Arbuzov Reaction is a well-known reaction to synthesise phosphonate. It's a Formation of a phosphonate from a trialkyl phosphite and an alkyl halide"

Completion1: "The reaction type that is able to convert trialkyl phosphite and alkyl halide to phosphonate is a Wacker reaction" **Completion2:** "The Arbuzov Reaction" [closer as it captures the subject!]

Truth: "The Arbuzov Reaction is a well-known reaction to synthesise phosphonate. It's a Formation of a phosphonate from a trialkyl phosphite and an alkyl halide"

Completion	BLEURT	BLEU	ROUGE-1	ROUGE-2
Completion 1	0.166	0.137	0.533	0.186
Completion 2	-0.644	0.001	0.207	0.148



Human Evaluation



Human evaluation [2023]



- Benchmark evaluation cannot cover all the quality dimensions of LLMs
- Human evaluations rely on A/B testing:
 - Pits two generations of two (blinded) models for same input
 - Humans express a preference (A is better, B is better, tie)

Two broad methods:

- 1. ELO scores
- 2. Votes w. IRR

The ELO score is a ranking score derived from chess. It is typically measured and updated after every single match. In the context of LLM "gamified" A/B testing it quantifies and aggregates user preferences towards a given model.

In what follows, $E_A \in [0,1]$ is the expected likelihood of A winning, whereas S_A is the "observed" likelihood with $S_A = 1$ if A wins, $S_A = 0$, if A loses and $S_A = 0.5$. Let ELO_A be the initial ELO score of A. After playing a game the new ELO score of A is

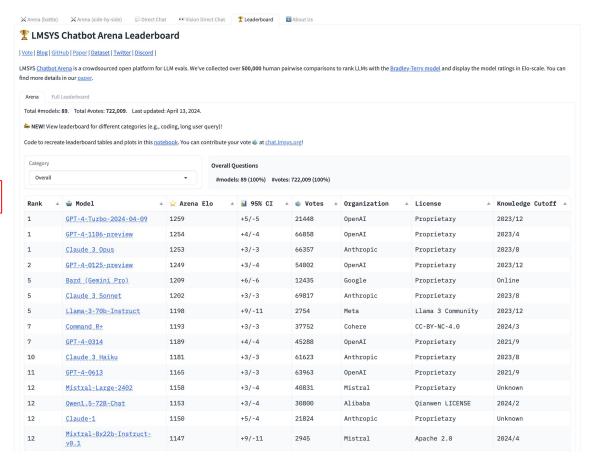
$$ELO'_A = ELO_A + k \cdot (S_A - E_A)$$

where k denotes a scaling factor that determines how much influence each particular match can have on the overall ELO rating of the player.

Chatbot arena leaderboard [2023-]



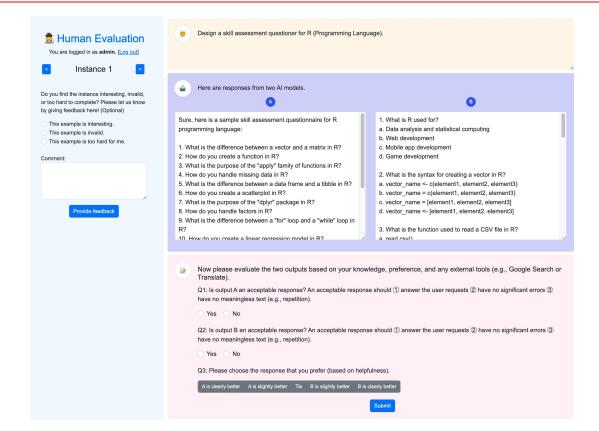
https://chat.lmsys.org/?leaderboard



Human evaluation – A/B testing



https://github.com/allenai/open-instruct/tree/main/human eval



Human evaluation – factuality [2024]

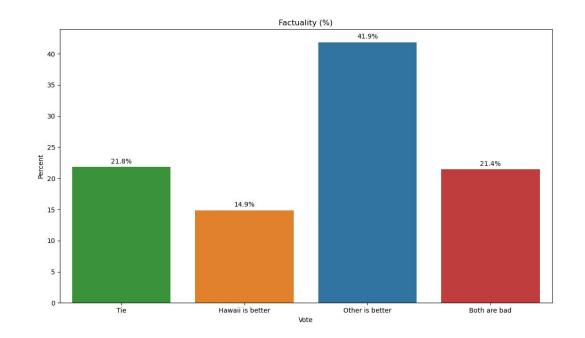


- Elsevier analysed the factuality of LLMs on custom LLS input prompts (NER, RE, open and closed book QA)
- ✓ >500 prompts judged by >40 content experts
- ✓ Alpha score / IRR > 0.6

Models

A) Hawaii: LSS-specific model (8b)

B) Other: LSS-fine tuned llama2 (7b), llama-2-chat (7b), GPT-3.5-turbo (175b)



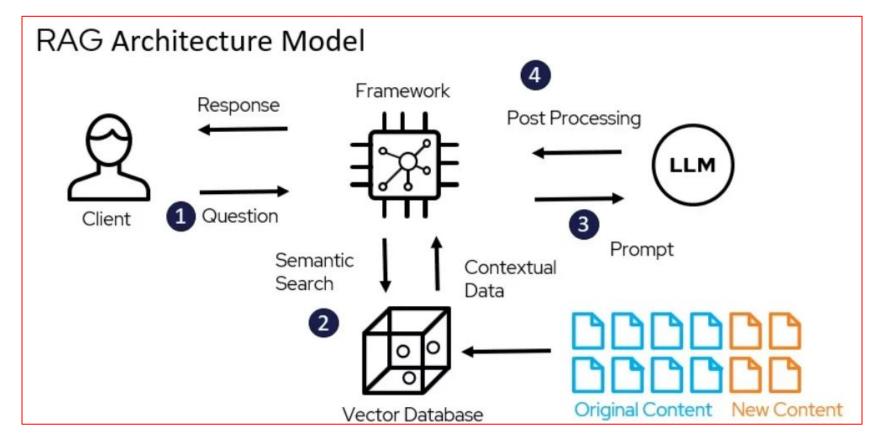


RAG Systems



Retrieval augmented generation [2022-]







References





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- 2024 Retrieval-Augmented Generation for Large Language Models: A Survey
 https://arxiv.org/abs/2312.10997



Thank you!

