Item Response Theory for NLP

EACL2024 Tutorial, 21st March 2024

John P. Lalor, Pedro Rodriguez, João Sedoc, Jose Hernandez-Orallo

https://eacl2024irt.github.io/

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Part 1. Evaluation for NLP

João Sedoc¹

¹ New York University

https://joaosedoc.com

What Do We Evaluate in NLP?

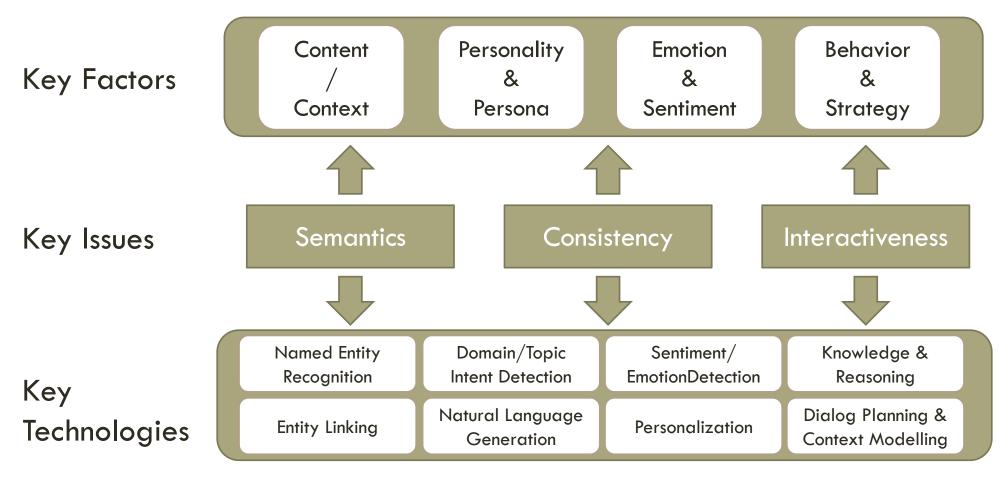
EVALUATIONS ARE AT SEVERAL LEVELS

- 1) System-level evaluations
 - This is probably the most common evaluation type (MT, Dialog, NLI, etc...)
- 2) Machine learning method evaluations
 - E.g., LSTM vs Transformer
- 3) Metrics
 - E.g., BLEU, BERTScore, etc
- 4) Annotations
 - Annotation error estimates
- 5) Data
 - Quality, domain similarity, toxicity

SYSTEM EVALUATIONS

- 1. Extrinsic task based evaluation
- 2. Intrinsic evaluation
- 3. Human evaluation
- 4. Automatic metric evaluation
- 5. A/B testing
- 6. Error analysis

CHALLENGES FOR DIALOG SYSTEMS



From Huang et al., 2019, "Challenges in Building Intelligent Open-Domain Systems"

COMMON TASK FRAMEWORK & LEADERBOARDS

There is general agreement that these competitive evaluations had a striking and beneficial effect on the performance of various systems tested over the years. However, it is also recognized (albeit less generally) that these evaluation experiments also had the, less beneficial, effect that the participating systems focused increasingly more narrowly on those few parameters that were measured in the evaluation, to the detriment of more general properties.

- Schwitter et al. 2000

Focusing on headline state-of-the-art numbers "provide(s) limited value for scientific progress absent insight into what drives them" and where they fail.

- Lipton and Steinhardt, 2019

LOTS OF IFANFRRAARDE B Spaces: mteb/leaderboard © © like 2 * Running on CPU UPGRADE



Y LMSYS Chatbot Arena Leaderboard

Files and versions 6 Community 2

The Stanfo

| Vote | Blog | GitHub | Paper | Dataset | Twitter | Discord |

Tota | Stag | Startus | Tuper | Successor | Twitter | Siscord

What is SQuAD?

Stanford Question Answering Dataset (SQuAD) is a reading comprehension dataset, consisting of questions posed by crowdworkers on a set of Wikipedia articles, where the answer to every question is a segment of text, or span, from the corresponding reading passage, or the question might be unanswerable.

SQuAD2.0 combines the 100,000 questions in SQuAD1.1 with over 50,000 unanswerable questions written adversarially by crowdworkers to look similar to answerable ones. To do well on SQuAD2.0, systems must not only answer questions when possible, but also determine when no answer is supported by the paragraph and abstain from answering.

Explore SQuAD2.0 and model predictions

SQuAD2.0 paper (Rajpurkar & Jia et al. '18)

SQuAD 1.1, the previous version of the SQuAD dataset, contains 100,000+ question-answer pairs on 500+ articles.

Explore SQuAD1.1 and model predictions

SQuAD1.0 paper (Rajpurkar et al. '16)

Arena Elo Full Leaderboard

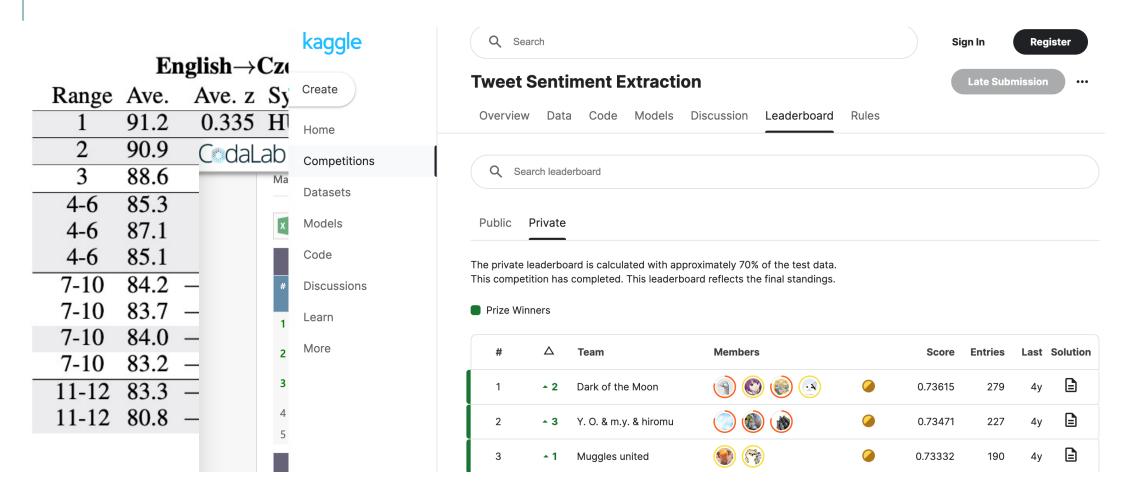
Total #models: 73. Total #votes: 408144. Last updated: March 13, 2024.

Contribute your vote at chat.lmsys.org! Find more analysis in the notebook.

Rank	Model	☆ Arena Elo	95% CI	♦ Votes ▲	Organization A	License	Knowledge Cutoff
1	GPT-4-1106-preview	1251	+5/-4	48226	OpenAI	Proprietary	2023/4
1	GPT-4-0125-preview	1249	+5/-6	22282	OpenAI	Proprietary	2023/12
1	Claude 3 Opus	1247	+6/-6	14854	Anthropic	Proprietary	2023/8
4	Bard (Gemini Pro)	1202	+6/-7	12623	Google	Proprietary	Online
4	Claude 3 Sonnet	1190	+6/-6	14845	Anthropic	Proprietary	2023/8
5	GPT-4-0314	1185	+4/-6	27245	OpenAI	Proprietary	2021/9
7	GPT-4-0613	1159	+4/-5	43783	OpenAI	Proprietary	2021/9

LMSYS Chatbot Arena is a crowdsourced open platform for LLM evals. We've collected over 400,000 human preference votes to rank LLMs with the Elo ranking system.

SHARED TASKS



LEADERBOARDS CAN IMPROVE

- 1. Questions with the Right Difficulty
- 2. Discriminative Questions
- 3. Minimize Ambiguity, Maximize Fairness
- 4. Don't be Overly Definitive
- 5. Be Flexible and Introspective

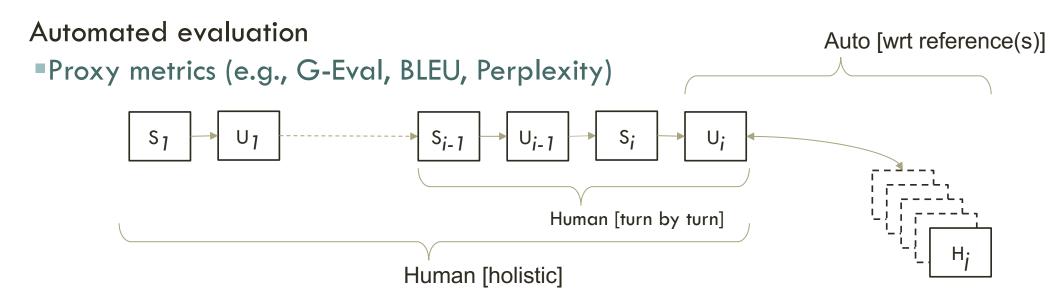
METHODS FOR RANKING

- 1. Average score
- 2. Z-scored ratings
- 3. Preference ranking
 - Bradley-Terry-Leech
 - Elo rating system
 - Trueskill
 - Item Response Theory

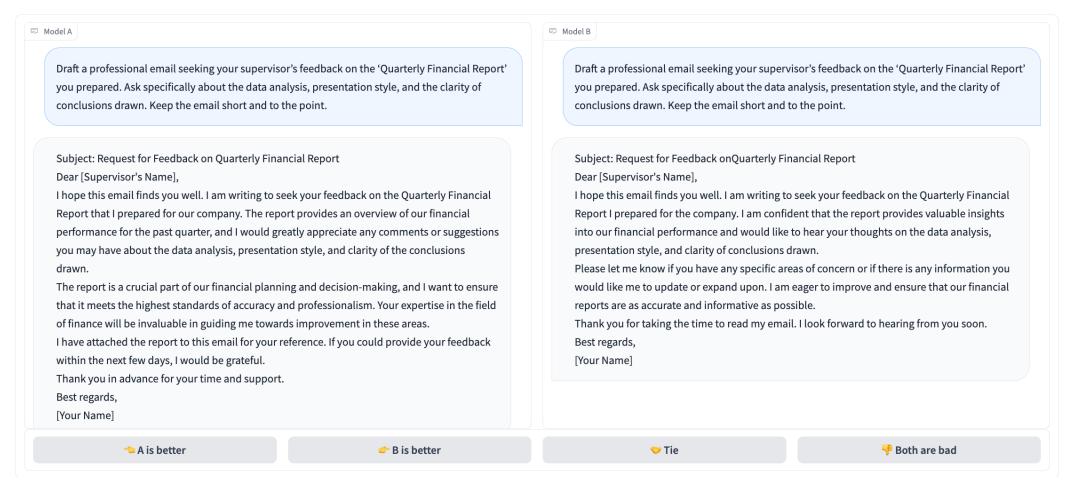
HUMAN / AUTOMATIC METRICS

Human evaluation

- Expert judges (WOCHAT, Alexa)
- Crowd-sourced (non-expert) judgments (DBDC)



A / B TESTING



ERROR ANALYSIS

- 1. Categorize error types
- 2. Investigate sources
- 3. Identify possible explanations

Annotations

EVALUATION OF ANNOTATIONS

- 1. Inter-annotator agreement (IAA)
 - Cohen's Kappa
 - Krippendorff's alpha
 - Fleiss' Kappa

- 2. Accuracy, Precision/Recall/F-score
- 3. Consistency checks
- 4. Error Analysis

Data

UNDERLYING DATA ANALYSIS

- 1. Quality of the examples
- 2. Difficulty of data
- 3. Usefulness for evaluation
- 4. Error Analysis

THANK YOU!

JOAO SEDOC

http://joaosedoc.com/ jsedoc@nyu.edu

NEXT UP

Next Section: Introduction to IRT