



Quality Assessment of Open-Source Language Model Generated Sentence Pairs in Rotorcraft Aviation Domain Emma L. McDaniel and Alicia I. Ruvinsky

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INTRODUCTION

- Rotorcraft decision support system
- Mining incident and accident reports for information deficiency landscape
- To data-mine, it is necessary to fine-tune for semantic similarity [1]
- fine-tune using contrastive learning, • To sentence pairs are required, not in dataset
- Endeavor to generate sentences pairs using open-source InstMixtral (8x7B SMoE) [2, 3] requires assessment due to hallucinations

METHODOLOGY

Evaluated generated sentences for 50 samples from our dataset across 7 prompts totaling 350 sentences.

Quality Assessment Metrics:

- 1) Stemmed word [4] similarity (BoW & Location [5])
- 2) Part of speech [6] similarity (BoW & Location)
- 3) Readability Scores [7] (Flesch-Kincaid Grade Level Score and Flesch Readability Ease)
- 4) Cosine Similarity [8] by embedding sentences using MPNet [9] (Semantic similarity)
- 5) Percentage of extra generated text

Table 1: Seven prompts utilized to generate sentences

Prompt Rewrite this sentence:

- Rewrite this sentence to be more informal:
- Rewrite this sentence to be more formal:
- Reorganize this sentence:
- Rewrite this sentence to be more concise:
- Rewrite this in another way:
- Rewrite this sentence using different vocabulary:

Table 2: is the m Metri Stem Un Stem Lev POS Uni POS Lev L

Flesch-Ki Flesc

Cosine Extra

- (96.43%)

- base-v2

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RESULTS

Vedian scores calculated across prompts (excluding Flesch-Kincaid, Flesh, and Extra %). "Average" nean of median scores per prompt. " <i>OS</i> " stands for original sentence score in readability metrics.								
ics	Average	Prompt 1	Prompt 2	Prompt 3	Prompt 4	Prompt 5	Prompt 6	Prompt 7
ion %	63.05%	71.42%	48.53%	64.17%	91.67%	71.43%	60.77%	33.33%
Dist %	80.67%	78.47%	81.82%	82.09%	73.03%	63.64%	89.17%	96.43%
on %	99.20%	100%	100%	100%	100%	94.38%	100%	100%
Dist %	57.02%	72.25%	50%	50%	64.06%	50%	62.83%	50%
incaid	OS:12.14	13.61	7.93	16.3	13.01	10.57	13.61	14.65
:h	OS:45.2	38.08	71.4	25.34	40.87	49.75	39.37	29.14
Sim	0.836	0.869	0.782	0.823	0.907	0.833	0.859	0.777
%	12%	8%	28%	18%	8%	14%	2%	6%

• Stemmed Union: Average across prompts is ~63%. Prompt 4 has the highest (91.67%); Prompt 7 has the lowest (33.33%)

• Stemmed Levenshtein Distance: Average across prompts is high (80.67%); Prompt 5 has the most similarity of word order (63.64%); Prompt 7 the most

• Part of Speech Union: high across all prompts due to low number of tags • Part of Speech Levenshtein Distance: ordering is on average 57% • Flesch-Kincaid Grade Level and Flesch Reading Ease: Prompt 2 had the lowest grade level score and the highest readability ease

• Cosine Similarity: Embeddings to original sentences produced low similarities despite perceived semantic closeness observed in manual evaluation

• Extra Sentences/Text: 12% had extraneous results in the prompt, with Prompt 2 having the highest at 28%

REFERENCES



CONCLUSION

generates the Prompt 7 most different with sentences and the evaluations, manual still similar sentences have semantic meanings.

Extra text generation occurs the most with Prompt 2.

Similarity Cosine between original sentence and generated did not result in sentences meaningful comparisons for the two sentences.

different Future work: utilize semantic similarity metrics and other generative LLMs.

REPRODUCIBILITY

The code and dataset utilized in this work are available on our open science framework project repository: <u>https://osf.io/9rtfy/</u>

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^[1] Emma L. McDaniel and Alicia I. Ruvinsky "Building Holistic Situation Awareness through Large Language Models." In proceedings of Cognitive Situation Management 2023 (CogSIMA '23); 16-20 October 2023 in Philadelphia, Pennsylvania. Proceedings to be published. [2] INSTMixtral. Mixtral-8x7B-Instruct-v0.1. Version: 4.36.2, Docs: https://huggingface.co/mistralai/Mixtral-8x7B-Instruct-v0.1 [3] Jiang, A. Q., Sablayrolles, A., Roux, A., Mensch, A., Savary, B., Bamford, C., Chaplot, D. S., Casas, D. d. l., Hanna, E. B., Bressand, F., et al (2024). Mixtral of Experts. arXiv preprint arXiv:2401.04088.

^[4] NLTK. Natural Language Toolkit. Version: 3.7.2, Docs: https://www.nltk.org/api/nltk.stem.html

^[5] Levenshtein Python C Extension Module. Version: 0.24.0, Docs: https://github.com/rapidfuzz/Levenshtein [6] spaCy. Version: 3.7.2, Docs: https://spacy.io/usage/linguistic-features

^[7] Readability Metrics. Version: 1.4.5, Docs: https://pypi.org/project/py-readability-metrics/

^[8] Scikit-Learn. Version: 1.4.0, Docs: https://scikit-learn.org/stable/modules/generated/sklearn.metrics.pairwise.cosine_ similarity.html [9] MPNet. sentence-transformers/all-mpnet-base-v2. Version: 4.36.2, Docs: https://huggingface.co/sentence-transformers/all-mpnet-