Evaluation of Large Language Models in Tailoring Educational Content for Underserved Cancer Survivors and Their Caregivers

Darren Liu[1,2], Xiao Hu[1,2,3], Canhua Xiao[1,4], Jinbing Bai[1,4], Zahra A. Barandouzi[1,4], Stephanie M. Lee[1], Caitlin I. Webster[1], La-Urshalar Brock[1,4], Lindsay J. Lee[5], Delgersuren Bold[1,2], Yufen Lin[1,4]

[1]Nell Hodgson Woodruff School of Nursing, Emory University [2]Center for Data Science, Emory University [3]The Wallace H. Coulter Department of Biomedical Engineering, Georgia Institute of Technology [4]Winship Cancer Institute, Emory University, [5]Department of Medicine, University of Florida

INTRODUCTION

- Underserved cancer survivors and their caregivers face a disproportionately increased risk of **symptom burden** from cancer and its treatments.
- **Large language models (LLMs)** offer researchers an opportunity to develop educational materials tailored to these populations.
- This study aimed to **evaluate different LLMs in tailoring educational content for underserved cancer survivors and their caregivers**.

METHODS

Data sources:

- Education materials are from the **national guidelines** (e.g., National Cancer Institute, National Comprehensive Cancer Network).
- We selected **30 distinct topics** such as fatigue, depression, anxiety, physical activity, coping skills, and more.

Models:

GPT-3.5 Turbo, GPT-4, and GPT-4 Turbo were utilized through Microsoft Azure OpenAI API to generate tailored content.

Prompts:

- At a low reading level (FKG ≤ 6);
- At a word limit of 250;
- In both textual and bulleted formats;
- Providing Spanish and Chinese translations for each topic.

Annotation:

- A panel of **nine oncology experts**, comprising four oncology professors, four doctoral students, and one medical resident.
- Each expert was assigned 10 topics to evaluate based on **seven criteria** and was required to provide feedback on the **errors**.

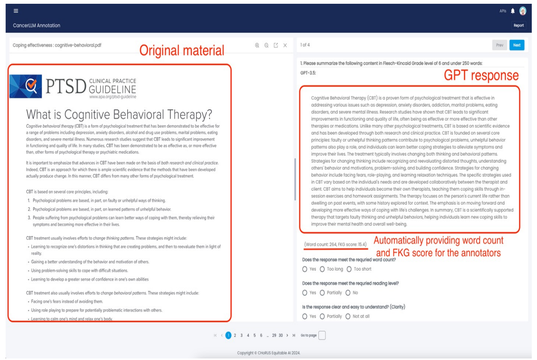


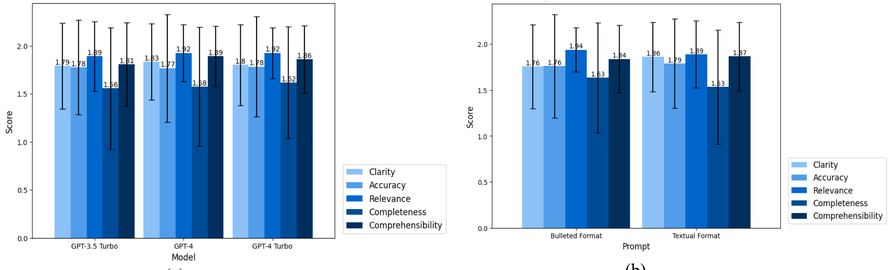
Figure 1. Screenshot of the Cohort Adjudication and Data Annotation application

- ANOVA or Chi-square analyses were employed to compare differences among the various GPT models and prompts.

RESULTS

Table 1. Performance of All Models, Prompts on the Summarization Tasks

	GPT-3.5 Turbo		GPT-4		GPT-4 Turbo	
Prompt	**Textual Format**	**Bullet Points**	**Textual Format**	**Bullet Points**	**Textual Format**	**Bullet Points**
Word Limit (%)	0.467	0.967	0.917	0.767	0.517	0.817
Reading Level (%)	0.183	0.283	0.217	0.217	0.533	0.317
Accuracy	1.767±0.500	1.783±0.49	1.800±0.480	1.733±0.634	1.800±0.48	1.767±0.563
Clarity*	1.833±0.418	1.750±0.474	1.867±0.389	1.800±0.403	1.883±0.324	1.717±0.49
Relevance	1.883±0.415	1.900±0.303	1.883±0.372	1.967±0.181	1.900±0.303	1.950±0.22
Completeness	1.533±0.623	1.583±0.645	1.483±0.624	1.667±0.601	1.583±0.619	1.650±0.547
Comprehensibility	1.817±0.469	1.800±0.403	1.883±0.324	1.900±0.303	1.900±0.303	1.817±0.39
Total Score	8.833±1.748	8.817±1.546	8.917±1.239	9.067±1.26	9.067±1.087	8.900±1.298
Spanish Translation (%)	0.933		0.967		1	
Chinese Translation (%)	0.767		0.867		0.800	



(a) (b)

Figure 2. Average Scores on Each Criterion Between: (a) Different Models; (b) Different Prompts.

- 74.2% (n=360) adhering to the specified word limit and achieving an average quality assessment score of 8.933 out of 10
- Achieving an accuracy of 88.9% for Spanish and 81.1% for Chinese translations
- Errors: inaccurate scope, expression, definition, meaningless points

DISCUSSION

- Overall, it is proven that LLMs are highly effective in tailoring, condensing, and translating educational content for underserved cancer patients and their caregivers.
- The findings from this study can inform the development and implementation of interventions in cancer symptom management and health equity.

CONCLUSION

- This study highlights the application of LLMs in cancer care and education while acknowledging their potential limitations.

ACKNOWLEDGMENT

- The study was supported by the Oncology Nursing Foundation Research Grant (2022 RE03).

yufen.lin@emory.edu