Lessons from a User Experience Evaluation of NLP Interfaces

Eduardo Calò, Lydia Penkert, Saad Mahamood

e.calo@uu.nl

NAACL 2025 (Findings)





Imagine having to go through this...



Motivations

- How human evaluation is presented impacts the quality of the data collected
 [3, 6]
- However this **aspect** is often **overlooked** by researchers
- \circ Can result in unreliable annotations \Rightarrow One factor hindering reproducibility [7]

What we do

- Bridge natural language processing (NLP) and human-centered design (HCD)
- Evaluate user interfaces (UIs) used in past NLP human evaluations based on HCD principles [4]
- o Draw **convenient recommendations** for designing Uls for human evaluations

HCD Interaction Principles

Suitability for the user's tasks: the UI supports the users in the completion of their tasks.

Self-descriptiveness: appropriate information is presented in the UI to make its capabilities and use immediately obvious.

Conformity with user expectations: the UI's behavior is predictable based on the context of use and commonly accepted conventions in that context.

Learnability: the UI supports the discovery of its capabilities, allows exploration, provides support, and minimizes the need for learning.

Controllability: the user maintains control of the UI and the interactions' speed, sequence, and individualization.

Use error robustness: the UI tolerates and assists the user in avoiding and recovering from errors.

User engagement: functions and information are presented in an inviting and motivating manner.

Uls and Participants

Four UIs from papers in ReproHum:

- 1. "It's not Rocket Science: Interpreting Figurative Language in Narratives" [2] (FL)
- 2. "Data-to-text Generation with Macro Planning" [5] study 1 (MLBF)
- 3. "Data-to-text Generation with Macro Planning" [5] study 2 (MLBC)
- 4. "NeuralREG: An end-to-end approach to referring expression generation" [1] (REG)

Three user experience (**UX**) **experts** (7-16 years of professional expertise)

Evaluation Procedure

INTERFACE: Link to the interface

Are the following principles met?

Suitability: Not met, Partially met, Met

If you answered *Not met* or *Partially met*, why do you think the principle is not (fully) met?

Self-descriptiveness: Not met, Partially met, Met

If you answered *Not met* or *Partially met*, why do you think the principle is not (fully) met?

. . .

IAA: from low to moderate (overall $\alpha = 0.339$) \Rightarrow not surprising (highly subjective)

 $\textbf{Ranking}: \ \mathsf{REG} > \mathsf{FL} > \mathsf{MLBC} > \mathsf{MLBF}$

Examples of Flaws



MLBF - Controllability issue

a) She needed to make it clear what she wanted

1. plausible

2. not plausible

FL - Self-descriptiveness issue

Press "Click to begin the HIT" to continue. Click to begin the HIT $\hat{a}-\P$

 $\mathsf{MLBC}\,\text{-}\,\mathsf{Self}\text{-}\mathsf{descriptiveness}\,\,\mathsf{issue}$

Recommendations

Principle	Recommendations
Self-descriptiveness	 Avoid confusing/subjective/judgmental/technical/redundant language Avoid long instructions, but if needed explain/present them properly Explain any part that may turn out to be unclear
Conformity	 Ensure uniformity in layout (e.g., length of the input fields) Use proper/consistent colors (e.g., brightness, palette, etc.) Organize/structure and position text in the right way Use the appropriate type of question based on the data you want to collect
Learnability	 Provide the right amount of examples Explain the terminology Give feedback Explain how to interact with the system
Controllability	 Provide users with the ability to revisit the instructions Enable empty state revert
Robustness	Clearly mark mandatory information Provide proper error messages (e.g., not too early, not persistent, not generic) Check input data in the backend Check if unwanted interactions with UI/text may occur Avoid default answers that may be misleading (e.g., default value of a slider)
Engagement	 Add a progress bar Do not use aggressive language (e.g., all-caps) Avoid heavy text/content/tables Give positive feedback after completion

Takeaway

RETHINK YOUR HUMAN EVALUATION INTERFACE!!!

References I

- [1] Thiago Castro Ferreira et al. "NeuralREG: An end-to-end approach to referring expression generation". In: Proceedings of the 56th Annual Meeting of the Association for Computational Linguistics (Volume 1: Long Papers). Ed. by Iryna Gurevych and Yusuke Miyao. Melbourne, Australia: Association for Computational Linguistics, July 2018, pp. 1959–1969. DOI: 10.18653/v1/P18-1182. URL: https://aclanthology.org/P18-1182/.
- [2] Tuhin Chakrabarty, Yejin Choi, and Vered Shwartz. "It's not rocket science: Interpreting figurative language in narratives". In: Transactions of the Association for Computational Linguistics 10 (2022), pp. 589–606.
- [3] Jessica Huynh, Jeffrey P. Bigham, and Maxine Eskénazi. "A Survey of NLP-Related Crowdsourcing HITs: what works and what does not". In: CoRR abs/2111.05241 (2021). arXiv: 2111.05241. URL: https://arxiv.org/abs/2111.05241.
- [4] ISO-9241-110. Ergonomics of human-system interaction Part 110: Interaction principles. May 2020. URL: https://www.iso.org/obp/ui/en/#iso:std:iso:9241:-110:ed-2:v1:en.

References II

- [5] Ratish Puduppully and Mirella Lapata. "Data-to-text Generation with Macro Planning". In:

 Transactions of the Association for Computational Linguistics 9 (May 2021), pp. 510-527. ISSN: 2307-387X. DOI: 10.1162/tacl_a_00381. eprint: https://direct.mit.edu/tacl/article-pdf/doi/10.1162/tacl\a_00381/1924176/tacl\a_00381.pdf. URL:

 https://doi.org/10.1162/tacl%5C_a%5C_00381.
- [6] Jamar Sullivan Jr. et al. "Explaining Why: How Instructions and User Interfaces Impact Annotator Rationales When Labeling Text Data". In: Proceedings of the 2022 Conference of the North American Chapter of the Association for Computational Linguistics: Human Language Technologies. Ed. by Marine Carpuat, Marie-Catherine de Marneffe, and Ivan Vladimir Meza Ruiz. Seattle, United States: Association for Computational Linguistics, July 2022, pp. 521–531. DOI: 10.18653/v1/2022.naacl-main.38. URL: https://aclanthology.org/2022.naacl-main.38/.

References III

[7] Craig Thomson, Ehud Reiter, and Anya Belz. "Common Flaws in Running Human Evaluation Experiments in NLP". In: Computational Linguistics (Mar. 2024), pp. 1–11. ISSN: 0891-2017. DOI: 10.1162/coli_a_00508. eprint: https://direct.mit.edu/coli/article-pdf/doi/10.1162/coli_a_00508/2348458/coli_a_00508.pdf. URL: https://doi.org/10.1162/coli%5C_a%5C_00508.