IRT for NLP: Structured Reading List

EACL 2024 Tutorial

1 Introductory

Frank B Baker. *The basics of item response theory*. ERIC, 2001

James E Carlson and Matthias von Davier. Item response theory. *ETS Research Report Series*, 2013(2):i–69, 2013

Prathiba Natesan, Ratna Nandakumar, Tom Minka, and Jonathan D Rubright. Bayesian prior choice in irt estimation using mcmc and variational bayes. *Frontiers in psychology*, 7:1422, 2016

2 IRT in NLP

Naoki Otani, Toshiaki Nakazawa, Daisuke Kawahara, and Sadao Kurohashi. IRT-based aggregation model of crowdsourced pairwise comparison for evaluating machine translations. In *Proceedings of the 2016 Conference on Empirical Methods in Natural Language Processing*, pages 511–520, Austin, Texas, November 2016. Association for Computational Linguistics

John P. Lalor, Hao Wu, and Hong Yu. Building an evaluation scale using item response theory. In *EMNLP*, 2016

John P. Lalor, Hao Wu, Tsendsuren Munkhdalai, and Hong Yu. Understanding deep learning performance through an examination of test set difficulty: A psychometric case study. In *EMNLP*, 2018

João Sedoc and Lyle Ungar. Item response theory for efficient human evaluation of chatbots. In *Proceedings of the First Workshop on Evaluation and Comparison of NLP Systems*, pages 21–33, Online, November 2020. Association for Computational Linguistics

Clara Vania, Phu Mon Htut, William Huang, Dhara Mungra, Richard Yuanzhe Pang, Jason Phang, Haokun Liu, Kyunghyun Cho, and Samuel R. Bowman. Comparing test sets with item response theory. In Chengqing Zong, Fei Xia, Wenjie Li, and Roberto Navigli, editors, *Proceedings of the 59th Annual Meeting of the Association for Computational Linguistics and the 11th International Joint Conference on Natural Language Processing (Volume 1: Long Papers)*, pages 1141–1158, Online, August 2021. Association for Computational Linguistics

Pedro Rodriguez, Joe Barrow, Alexander Miserlis Hoyle, John P. Lalor, Robin Jia, and Jordan Boyd-Graber. Evaluation examples are not equally informative: How should that change NLP leaderboards? In *ACL*, 2021

John P. Lalor and Pedro Rodriguez. py-irt: A scalable item response theory library for python. *INFORMS Journal on Computing*, 2022

Pedro Rodriguez, Phu Mon Htut, John Lalor, and João Sedoc. Clustering examples in multi-dataset benchmarks with item response theory. In *Proceedings of the Third Workshop on Insights from Negative Results in NLP*, pages 100–112, Dublin, Ireland, May 2022. Association for Computational Linguistics

3 IRT in ML

Fernando Martinez-Plumed and Jose Hernandez-Orallo. Dual indicators to analyze ai benchmarks: Difficulty, discrimination, ability, and generality. *IEEE Transactions on Games*, 12(2):121–131, 2018

Fernando Martínez-Plumed, Ricardo BC Prudêncio, Adolfo Martínez-Usó, and José Hernández-Orallo. Item response theory in ai: Analysing machine learning classifiers at the instance level. *Artificial intelligence*, 271:18–42, 2019

M Wu, R Davis, B Domingue, C Piech, and Noah D Goodman. Variational item response theory: Fast, accurate, and expressive. 2020

4 Advanced

Jacopo Amidei, Paul Piwek, and Alistair Willis. Identifying annotator bias: A new IRT-based method for bias identification. In *Proceedings of the 28th International Conference on Computational Linguistics*, pages 4787–4797, Barcelona, Spain (Online), December 2020. International Committee on Computational Linguistics

José Hernández-Orallo, Bao Sheng Loe, Lucy Cheke, Fernando Martínez-Plumed, and Seán Ó hÉigeartaigh. General intelligence disentangled via a generality metric for natural and artificial intelligence. *Scientific reports*, 11(1):22822, 2021

Antonio Laverghetta Jr., Animesh Nighojkar, Jamshidbek Mirzakhalov, and John Licato. Can transformer language models predict psychometric properties? In *Proceedings of *SEM 2021: The Tenth Joint Conference on Lexical and Computational Semantics*, pages 12–25, Online, August 2021. Association for Computational Linguistics

Fernando Martínez-Plumed, David Castellano, Carlos Monserrat-Aranda, and José Hernández-Orallo. When ai difficulty is easy: The explanatory power of predicting irt difficulty. In *Proceedings of the AAAI Conference on Artificial Intelligence*, volume 36, pages 7719–7727, 2022

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[2] James E Carlson and Matthias von Davier. Item response theory. *ETS Research Report Series*, 2013(2):i–69, 2013.

[3] John P. Lalor, Hao Wu, and Hong Yu. Building an evaluation scale using item response theory. In *EMNLP*, 2016.

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[9] Prathiba Natesan, Ratna Nandakumar, Tom Minka, and Jonathan D Rubright. Bayesian prior choice in irt estimation using mcmc and variational bayes. *Frontiers in psychology*, 7:1422, 2016.

[10] Naoki Otani, Toshiaki Nakazawa, Daisuke Kawahara, and Sadao Kurohashi. IRT-based aggregation model of crowdsourced pairwise comparison for evaluating machine translations. In *Proceedings of the 2016 Conference on Empirical Methods in Natural Language Processing*, pages 511–520, Austin, Texas, November 2016. Association for Computational Linguistics.

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[12] João Sedoc and Lyle Ungar. Item response theory for efficient human evaluation of chatbots. In *Proceedings of the First Workshop on Evaluation and Comparison of NLP Systems*, pages 21–33, Online, November 2020. Association for Computational Linguistics.

[13] Clara Vania, Phu Mon Htut, William Huang, Dhara Mungra, Richard Yuanzhe Pang, Jason Phang, Haokun Liu, Kyunghyun Cho, and Samuel R. Bowman. Comparing test sets with item response theory. In

Chengqing Zong, Fei Xia, Wenjie Li, and Roberto Navigli, editors, *Proceedings of the 59th Annual Meeting of the Association for Computational Linguistics and the 11th International Joint Conference on Natural Language Processing (Volume 1: Long Papers)*, pages 1141–1158, Online, August 2021. Association for Computational Linguistics.

[14] M Wu, R Davis, B Domingue, C Piech, and Noah D Goodman. Variational item response theory: Fast, accurate, and expressive. 2020.

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[18] Fernando Martínez-Plumed, David Castellano, Carlos Monserrat-Aranda, and José Hernández-Orallo. When ai difficulty is easy: The explanatory power of predicting irt difficulty. In *Proceedings of the AAAI Conference on Artificial Intelligence*, volume 36, pages 7719–7727, 2022.