

Multi-Level AI Trustworthiness Labels Scale Potential Users' Perceptions and Evaluations of AI Products

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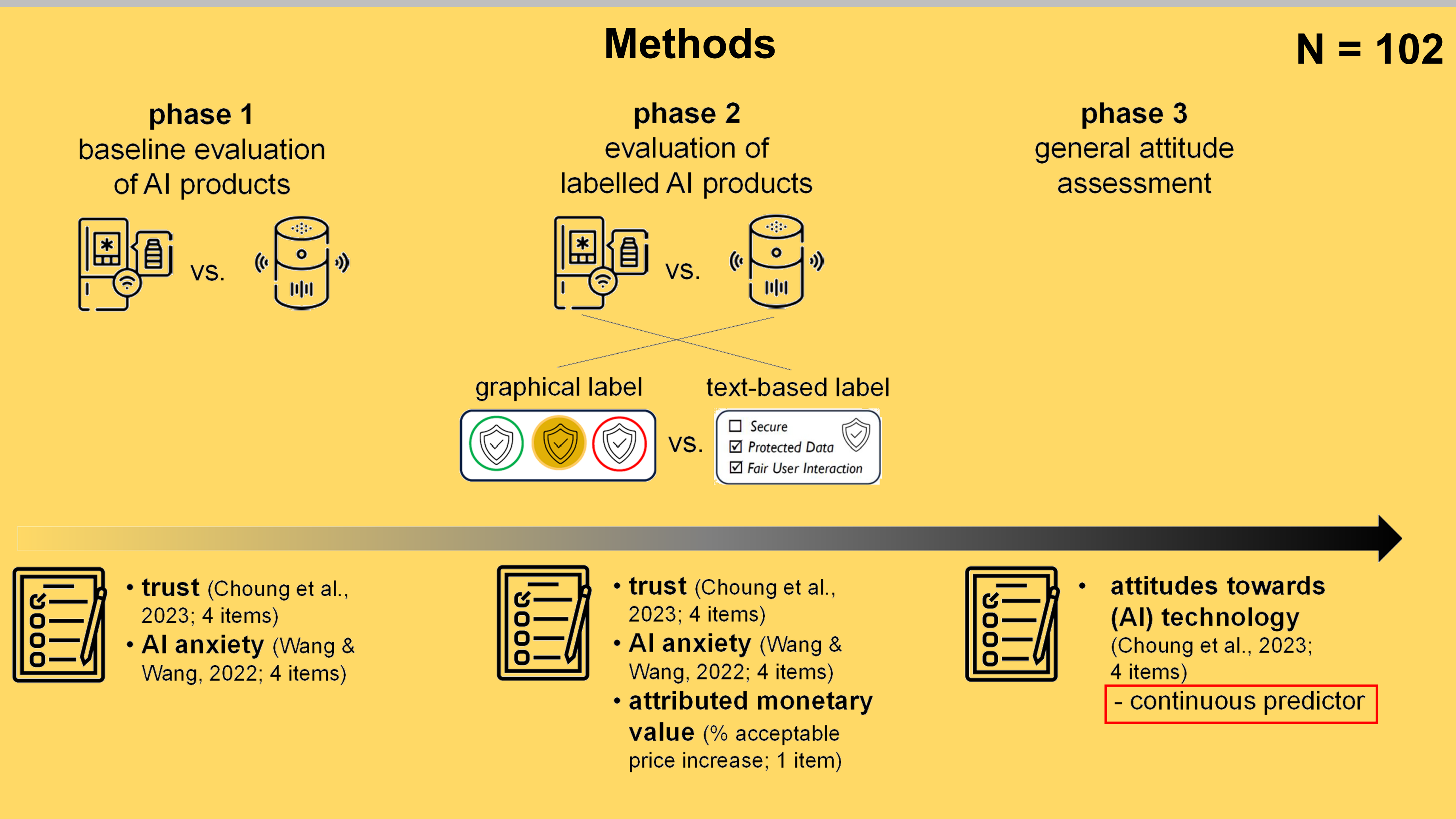


Theoretical Background

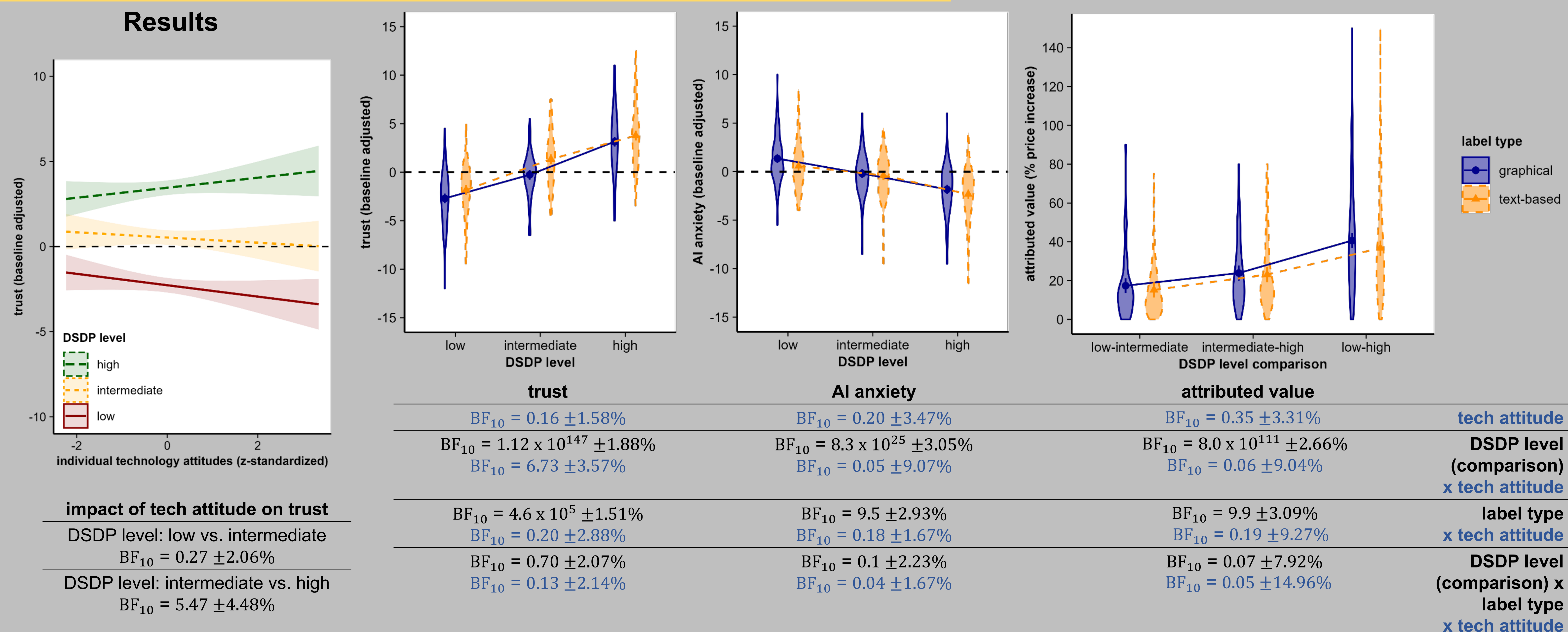
(Potential) users of artificial intelligence (AI) products (e.g., smart fridges, voice assistants) are rarely able to evaluate the trustworthiness of AI products accurately (Gillespie et al., 2023; Schlicker et al., 2025), as corresponding information is commonly not easily accessible. Concerns regarding AI trustworthiness, in particular, data security and data privacy (DSDP) concerns (Gillespie et al., 2023) jeopardize a further widespread acceptance and broader adoption of AI products (see e.g., Marangunić & Granić, 2015; Venkatesh et al., 2003, for theories on technology acceptance). Trust is an essential precursor of technology acceptance and adoption (e.g., Vorm & Combs, 2022).

As such, both users' misplaced distrust (e.g., Choung et al., 2023; Schlicker et al., 2025) as well as users' misplaced trust (due to expectancy violations, e.g., Hong, 2021; Rheu et al., 2024) prevent the further acceptance and adoption of new (and trustworthy) AI technologies and obstruct corresponding benefits of AI usage.

Here, I investigated whether multi-level AI trustworthiness labels communicating DSDP information suitably scale users' (dis-)trust in AI products – in particular when additionally considering individual attitudes towards (AI) technology. Importantly, participants rated both labeled as well as unlabeled (baseline) AI products, allowing for an additional assessment of biases.



Results



Discussion

Trust, AI anxiety, as well as the monetary value attributed to AI products scaled with an AI trustworthiness label's DSDP level. Importantly, participants' ratings of unlabeled AI products corresponded to their perceptions of AI products labeled with an intermediate DSDP level.

This apparent bias towards intermediate DSDP judgements in the absence of information on AI products underscores the relevance of explicitly communicating AI trustworthiness to (potential) users. Interestingly, differences in trust for AI products with a high as compared to intermediate DSDP level further increased with more positive attitudes towards AI technology.

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