Construct Validity Checklist

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This checklist follows the recommendations made in the paper:

https://openreview.net/pdf?id=mdA5lVvNcU Define the phenomenon □ Provide a precise and operational definition for the phenomenon being measured □ Specify the scope of the phenomenon being covered and acknowledge any excluded aspects □ Identify if the phenomenon has sub-components and ensure they are measured separately Measure only the phenomenon □ Control for unrelated tasks that may affect the results □ Assess the impact of format constraints on model performance □ Validate any automated output parsing techniques for accuracy, consistency and bias Construct a representative dataset for the task ☐ Employ sampling strategies to ensure task items are representative of the overall task space ☐ Verify the quality and relevance of all task items, especially for large or automatically generated datasets □ Include task items that test known LLM sensitivities (e.g. input permutations or variations) Acknowledge limitations of reusing datasets □ Document whether the benchmark adapts a previous dataset or benchmark ☐ If so, analyse and report the relevant strengths and limitations of the adapted prior work □ If so, report and compare performance on the new benchmark against the original □ Explain modifications to reused datasets and how they improve construct validity Prepare for contamination □ Implement tests to detect data contamination and apply them to the benchmark □ Maintain a held-out set of task items to facilitate ongoing, uncontaminated evaluation ☐ Investigate the potential pre-exposure of benchmark source materials or similar data in common LLM training corpora Use statistical methods to compare models □ Report the benchmark's sample size and justify its statistical power □ Report uncertainty estimates for all primary scores to enable robust model comparisons □ If using human raters, describe their demographics and mitigate potential demographic biases in rater recruitment and instructions □ Use metrics that capture the inherent variability of any subjective labels, without relying on singlepoint aggregation or exact matching. Conduct an error analysis □ Conduct a qualitative and quantitative analysis of common failure modes □ Investigate whether failure modes correlate with non-targeted phenomena (confounders) rather than the intended construct □ If so, identify and discuss any potential scoring biases revealed in the error analysis □ Conduct experiments or propose new directions to improve model scores on the benchmark Justify construct validity □ Justify the relevance of the benchmark for the phenomenon with real-world applications □ Provide a clear rationale for the choice of tasks and metrics, connected to the operational definition of the phenomenon □ Compare similarities and differences between the benchmark and existing evaluations of similar phe-

□ Discuss the limitations and design trade-offs of the benchmark concerning construct validity

Measuring what Matters: Construct Validity in Large Language Model Benchmarks