## **A Construct Validity Checklist**

For usability, we have reproduced the complete construct validity checklist here, grouped by recommendation. We recommend that checklist users consider each of these questions and whether the answer is adequately addressed in the benchmark and corresponding paper. We anticipate that it may be difficult to adopt every recommendation in every case. For example, computing confidence intervals may be prohibitively expensive. In these cases, considering and discussing the tradeoffs of adopting the recommendations as limitations would enable readers of these papers to better interpret the results.

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

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