# Construct Validity Checklist

This checklist follows the recommendations made in the paper:

Measuring what Matters: Construct Validity in Large Language Model Benchmarks https://openreview.net/pdf?id=mdA51VvNcU

#### 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 single-point 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 phenomena

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