

# Performance-Efficiency Scaling in Code Generation Models from 0.5B to 13B Parameters

Assignee Research

May 30, 2026

## Abstract

This report synthesises findings from 14 peer-reviewed papers addressing the following research question: How does the Performance-Efficiency Ratio scale with model size (0.5B to 13B parameters) when tested on the original vs. progressively harder versions of HumanEval and MBPP benchmarks under the same. We introduce self-invoking code generation, a new task designed to evaluate the progressive reasoning and problem-solving capabilities of LLMs. In this task, models are presented with a base problem and a related, more complex problem. 11 claims were extracted from source literature; 2 were independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 4.5/10. This report is a machine-generated literature synthesis and does not constitute original research.

## 1 Introduction

This paper examines: HumanEval Pro and MBPP Pro: Evaluating Large Language Models on Self-invoking Code Generation. Research question: How does the Performance-Efficiency Ratio scale with model size (0.5B to 13B parameters) when tested on the original vs. progressively harder versions of HumanEval and MBPP benchmarks under the same inference budget constraints?.

## 2 Methodology

Systematic literature search across multiple databases yielded 14 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 4.5/10.

### 3 Results

14 papers retrieved. 11 claims extracted; 2 independently verified. Quality review score: 4.5/10.

### 4 Limitations

This report is a machine-generated literature synthesis and does not constitute original research. Automated retrieval and verification may introduce errors or omissions. Review scores reflect automated assessment, not human peer review. Readers should consult primary sources for authoritative information.

### 5 Extracted Claims

Claim	Verified	Confidence
o1-mini achieves 96.2% pass@1 on HumanEval but only 76.2% on HumanEval Pro.	✓	0.25
Instruction-tuned models are less efficient on self-invoking code generation than traditional code generation tasks.	✓	0.31
HumanEval and MBPP serve as fundamental benchmarks, focusing on Python function completion tasks with test-driven evaluation	×	0.06
Several benchmarks have expanded code evaluation benchmarks to encompass multiple programming languages, complex tasks	×	0.09
Deepseek-V2.5 is used to generate self-invoking problems, candidate solutions, and test inputs.	×	0.07
The benchmark construction process involves three steps: self-invoking problem generation, solution generation, and test	×	0.13
The final execution results are used to construct complete test cases with assert commands.	×	0.03
Qwen2.5-Coder-7B-base achieves 59.6% on HumanEval Pro and 38.6% on MBPP Pro.	×	0.12
Qwen2.5-Coder-7B-instruct achieves 64.9% on HumanEval Pro and 35.1% on MBPP Pro.	×	0.09
DeepseekCoder-33B-base achieves 71.9% on HumanEval Pro and 38.6% on MBPP Pro.	×	0.12
DeepseekCoder-33B-instruct achieves 80.7% on HumanEval Pro and 43.9% on MBPP Pro.	×	0.10

## References

- <http://arxiv.org/abs/2505.03786v1>
- <http://arxiv.org/abs/2605.11209v1>
- <http://arxiv.org/abs/2412.21199v2>