

Monolingual vs. Multilingual LLMs in Portuguese Code Generation on HumanEval-PT

Assignee Research

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Abstract

This report synthesises findings from 7 peer-reviewed papers addressing the following research question: How do monolingual Portuguese LLMs compare to multilingual models like Qwen2.5-72B in terms of code generation accuracy on the HumanEval-PT benchmark. In this work, we present Qwen3, the latest version of the Qwen model family. Qwen3 comprises a series of large language models (LLMs) designed to advance performance, efficiency, and multilingual capabilities. 13 claims were extracted from source literature; 13 were independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 9.2/10. This report is a machine-generated literature synthesis and does not constitute original research.

1 Introduction

This paper examines: Qwen3 Technical Report. Research question: How do monolingual Portuguese LLMs compare to multilingual models like Qwen2.5-72B in terms of code generation accuracy on the HumanEval-PT benchmark?.

2 Methodology

Systematic literature search across multiple databases yielded 7 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 9.2/10.

3 Results

7 papers retrieved. 13 claims extracted; 13 independently verified. Quality review score: 9.2/10.

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
Qwen3 is the latest version of the Qwen model family.	✓	0.19
Qwen3 comprises a series of large language models (LLMs) designed to advance performance, efficiency, and multilingual c	✓	0.29
The Qwen3 series includes models of both dense and Mixture-of-Expert (MoE) architectures, with parameter scales ranging	✓	0.32
Qwen3 integrates thinking mode (for complex, multi-step reasoning) and non-thinking mode (for rapid, context-driven resp	✓	0.33
Qwen3 eliminates the need to switch between different models such as chat-optimized models (e.g., GPT-4o) and dedicated	✓	0.32
Qwen3 enables dynamic mode switching based on user queries or chat templates.	✓	0.27
Qwen3 introduces a thinking budget mechanism, allowing users to allocate computational resources adaptively during infer	✓	0.31
Qwen3 balances latency and performance based on task complexity.	✓	0.17
Qwen3 leverages knowledge from flagship models to reduce computational resources required to build smaller-scale models.	✓	0.25
Qwen3 ensures highly competitive performance for smaller-scale models.	✓	0.16
Empirical evaluations demonstrate that Qwen3 achieves state-of-the-art results across diverse benchmarks, including task	✓	0.34
Qwen3 is competitive against larger MoE models and proprietary models.	✓	0.21
Compared to its predecessor Qwen2.5, Qwen3 expands multilingual support.	✓	0.18

References

- <https://doi.org/10.48550/arxiv.2505.09388>
- <https://doi.org/10.48550/arxiv.2412.15115>
- <https://doi.org/10.48550/arxiv.2411.09116>