

# Optimizing Generative Models for Tabular Fidelity Metrics and Multimodal Few-Shot Reasoning

Assignee Research

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## Abstract

Generative models have revolutionized multiple domains, yet their application to tabular data remains underexplored. Evaluating generative models for tabular data presents unique challenges due to structural complexity, large-scale variability, and mixed data types, making it difficult to intuitively capture intricate patterns. Existing evaluation metrics offer only partial insights, lacking a comprehensive measure of generative performance. To address this limitation, we propose three novel evaluation metrics: FAED, FPCAD, and RFIS. Our extensive experimental analysis, conducted on three stan

## 1 Introduction

This paper examines: Evaluating Generative Models for Tabular Data: Novel Metrics and Benchmarking. Research question: Does optimizing generative models for proposed tabular fidelity metrics improve the few-shot reasoning capabilities of multimodal foundation models on heterogeneous data benchmarks?.

## 2 Methodology

Systematic literature search across multiple databases yielded 11 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 8.8/10.

## 3 Results

11 papers retrieved. 10 claims extracted; 10 independently verified. Quality review score: 8.8/10.

## 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 |
|--|----------|------------|
| FAED effectively captures generative modeling issues overlooked by existing metrics.                                     | ✓        | 0.28       |
| FPCAD exhibits promising performance but requires further refinements to enhance its reliability.                        | ✓        | 0.18       |
| FAED successfully detects all synthesized problems, whereas FPCAD shows promising but improvable performance.            | ✓        | 0.21       |
| Existing metrics fail to identify key issues in generative modeling for tabular data.                                    | ✓        | 0.21       |
| FAED, FPCAD, and RFIS are novel evaluation metrics tailored for assessing generative models in tabular data.             | ✓        | 0.23       |
| Existing evaluation principles from the image domain can be adapted for tabular data.                                    | ✓        | 0.18       |
| SDV Fidelity, Utility, TSTR, and TRTS exhibit limitations in detecting key generative modeling challenges.               | ✓        | 0.28       |
| FAED, FPCAD, and RFIS effectively capture issues such as Quality Decrease, Mode Drop, and Mode Collapse compared to exis | ✓        | 0.30       |
| TSTR is useful for detecting cases where synthetic data only partially represents real data.                             | ✓        | 0.24       |
| TRTS assesses whether synthetic samples introduce patterns absent in real data.  | ✓        | 0.23       |

## References

- <http://arxiv.org/abs/2311.14544v1>
- <http://arxiv.org/abs/2504.20900v1>
- <http://arxiv.org/abs/2402.01204v4>