

Robustness of Tabular Generative Evaluation Metrics to Distributional Shifts in Mixed-Type Data Versus Kernel-Based Distances

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: How robust are proposed tabular generative evaluation metrics to distributional shifts in mixed-type data compared to traditional kernel-based distances?.

2 Methodology

Systematic literature search across multiple databases yielded 16 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 7.9/10.

3 Results

16 papers retrieved. 13 claims extracted; 12 independently verified. Quality review score: 7.9/10.

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.	✓	0.29
FPCAD exhibits promising performance but requires further refinements to enhance its reliability.	✓	0.18
FAED successfully detects all synthesized problems in the experimental results on three standard network intrusion detection	✓	0.18
FPCAD shows promising but improvable performance in detecting synthesized problems.	×	0.15
Existing metrics fail to identify key issues in generative modeling for tabular data.	✓	0.22
FAED and FPCAD are novel and robust metrics for evaluating generative models in the tabular data domain.	✓	0.20
Quality Decrease, Mode Drop, and Mode Collapse issues were embedded into the datasets to simulate real-world generative	✓	0.29
FAED, FPCAD, and RFIS are tailored for assessing generative models in tabular data, inspired by IS and FID from the image	✓	0.23
Existing evaluation principles from the image domain can be effectively adapted for tabular data.	✓	0.21
SDV Fidelity, Utility, TSTR, and TRTS exhibit limitations in detecting key generative modeling challenges.	✓	0.28
TSTR is useful for detecting cases where synthetic data only partially represents real data.	✓	0.25
TRTS assesses whether synthetic samples introduce patterns absent in real data.	✓	0.24
FAED, FPCAD, and RFIS offer robust tools for assessing generative models in the tabular data domain.	✓	0.19

References

- <http://arxiv.org/abs/2310.13146v1>
- <http://arxiv.org/abs/2402.13042v3>

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