



# TabStruct

## Measuring Structural Fidelity of Tabular Data

Xiangjian Jiang<sup>1</sup>, Nikola Simidjievski<sup>2,1</sup>, Mateja Jamnik<sup>1</sup>

<sup>1</sup>Department of Computer Science and Technology, University of Cambridge, UK

<sup>2</sup>Télécom Paris, Institut Polytechnique de Paris, France

xj265@cam.ac.uk, nikola.simidjievski@telecom-paris.fr, mateja.jamnik@cl.cam.ac.uk



 Paper



 Code

1

# Which image is real?

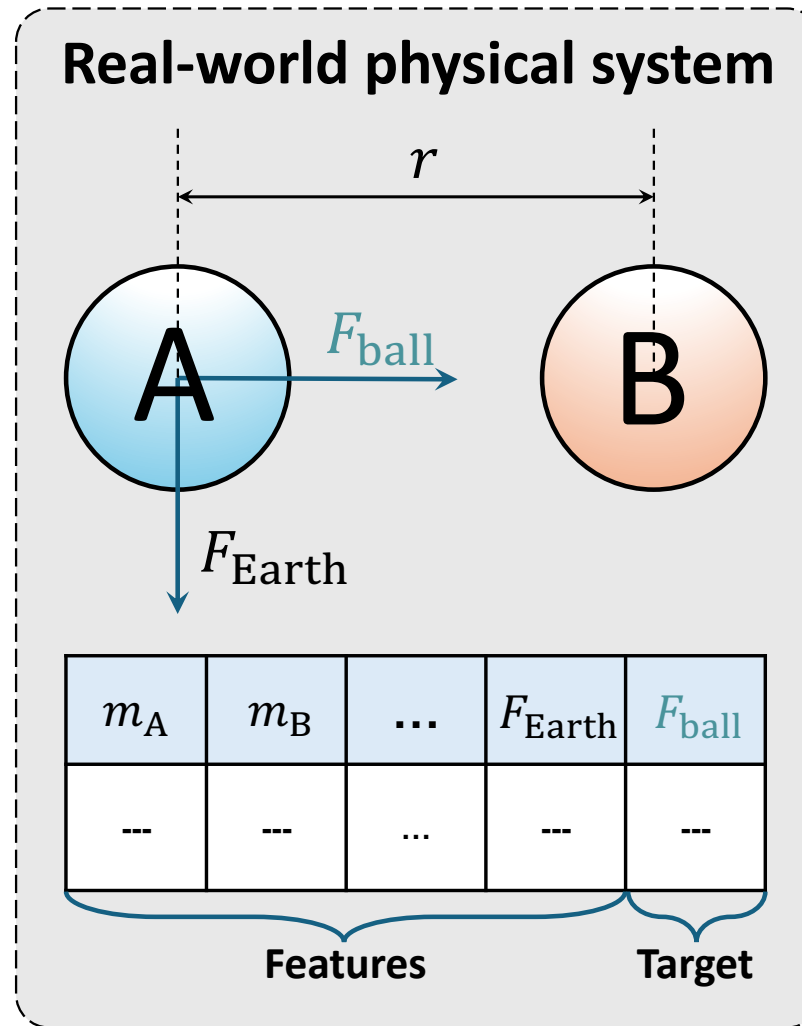


Background

Methodology

Experiments

# An example table of gravity system



# 3 Which table is real?

$m_A$	$m_B$	...	$F_{\text{Earth}}$	$F_{\text{ball}}$
0.3kg	2.2kg	...	3.4N	5.2N
0.5kg	3.4kg	...	2.9N	2.5N
0.6kg	6.4kg	...	1.5N	7.2N

$m_A$	$m_B$	...	$F_{\text{Earth}}$	$F_{\text{ball}}$
0.9kg	1.1kg	...	1.4N	0.7N
0.7kg	2.6kg	...	1.9N	1.8N
0.3kg	3.3kg	...	1.8N	3.0N

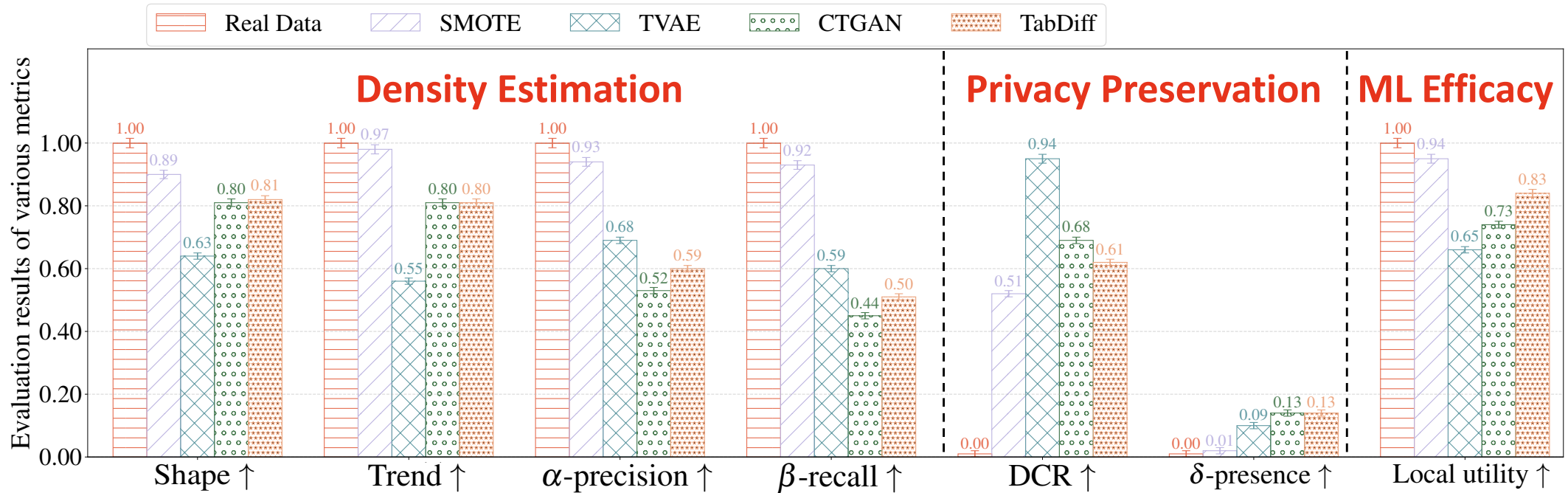
# Tabular data evaluation remains vague



$m_A$	$m_B$	...	$F_{\text{Earth}}$	$F_{\text{ball}}$	$m_A$	$m_B$	...	$F_{\text{Earth}}$	$F_{\text{ball}}$
0.3kg	2.2kg	...	3.4N	5.2N	0.9kg	1.1kg	...	1.4N	0.7N
0.5kg	3.4kg	...	2.9N	2.5N	0.7kg	2.6kg	...	1.9N	1.8N
0.6kg	6.4kg	...	1.5N	7.2N	0.3kg	3.3kg	...	1.8N	3.0N

- Humans (we 😊) are born with cognitive advantages to parse:
  - Sequential data: text, speech, etc.
  - Spatial data: image, etc.
- **Then, what about structured tabular data?**

# Is interpolation all we need?



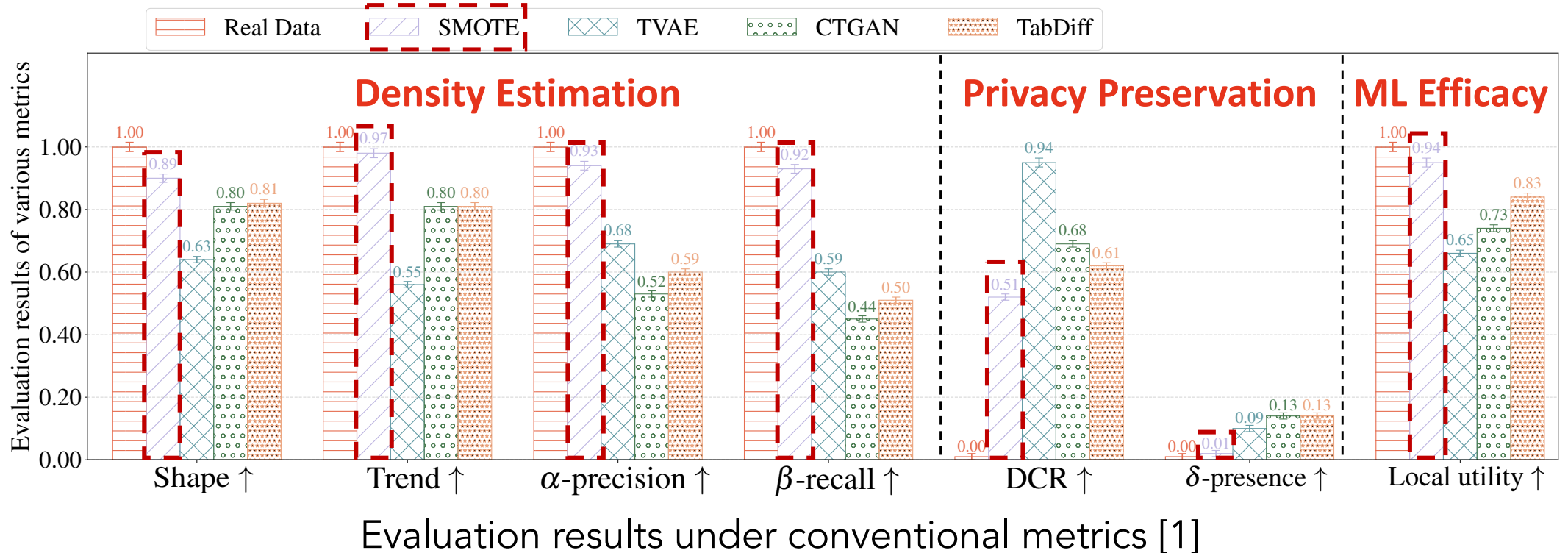
Evaluation results under conventional metrics [1]

## ○ Conventional evaluation metrics

- Density Estimation: discrepancy between **statistics**
- Privacy Preservation: risk of **privacy** leakage
- ML Efficacy: **utility** in downstream predictive tasks

[1] Jiang, Xiangjian, et al. "TabStruct: Measuring Structural Fidelity of Tabular Data." The Twelfth International Conference on Learning Representations (ICLR 2026 Oral).

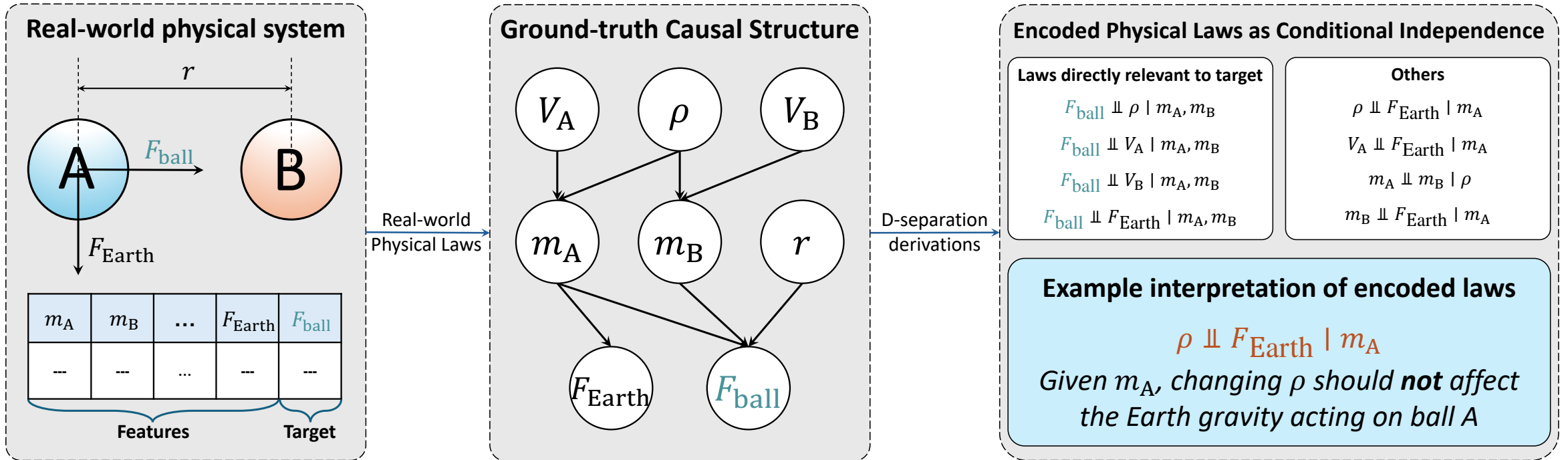
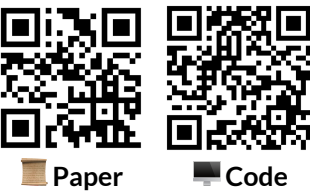
# Is interpolation all we need?



## ○ Conventional evaluation metrics

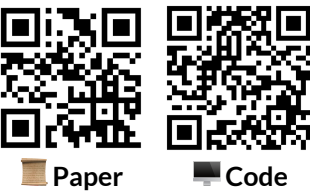
- Density Estimation: discrepancy between **statistics**
- Privacy Preservation: risk of **privacy** leakage
- ML Efficacy: **utility** in downstream predictive tasks

[1] Jiang, Xiangjian, et al. "TabStruct: Measuring Structural Fidelity of Tabular Data." The Twelfth International Conference on Learning Representations (ICLR 2026 Oral).



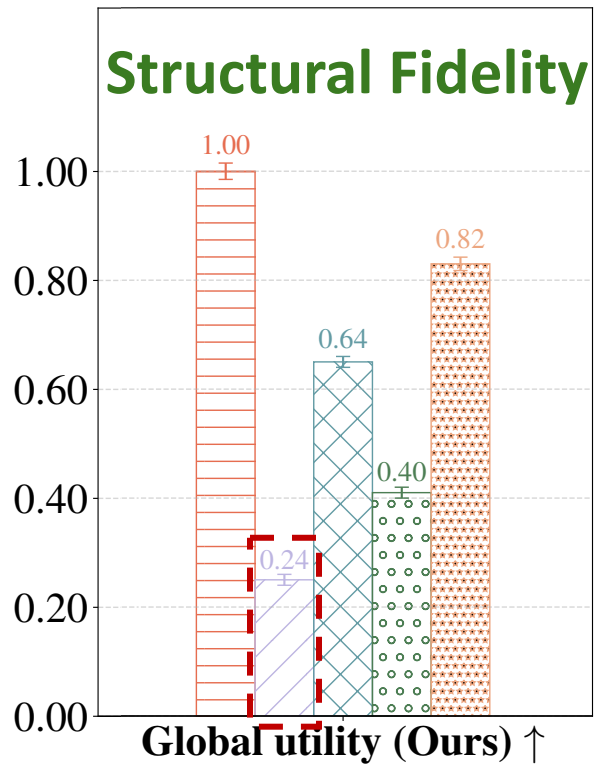
A physical system and its underlying causal structure

# Let's evaluate the causal structures!

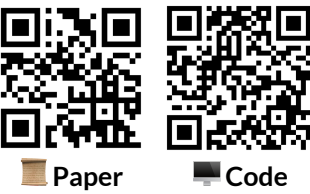


GT conditional independence		Real Data	SMOTE	TVAE	CTGAN	TabDiff
Global structure	Local structure					
	$F_{\text{ball}} \perp\!\!\!\perp \rho \mid m_A, m_B$	✓	✓	✓	✓	✓
	$F_{\text{ball}} \perp\!\!\!\perp V_A \mid m_A, m_B$	✓	✓	✓	✓	✓
	$F_{\text{ball}} \perp\!\!\!\perp V_B \mid m_A, m_B$	✓	✓	✗	✓	✓
	$F_{\text{ball}} \perp\!\!\!\perp F_{\text{Earth}} \mid m_A, m_B$	✓	✓	✓	✗	✗
	$\rho \perp\!\!\!\perp F_{\text{Earth}} \mid m_A$	✓	✗	✗	✓	✓
	$V_A \perp\!\!\!\perp F_{\text{Earth}} \mid m_A$	✓	✗	✓	✓	✓
	$m_A \perp\!\!\!\perp m_B \mid \rho$	✓	✗	✓	✗	✓
$m_B \perp\!\!\!\perp F_{\text{Earth}} \mid m_A$	✓	✗	✓	✗	✓	
<b>Obey physical laws (Global CI score)</b>		<b>100.0%</b>	<b>50.0%</b>	<b>75.0%</b>	<b>62.5%</b>	<b>87.5%</b>

Evaluation results of various metrics



- **Global CI**: golden standard, but **inapplicable** in many real-world datasets
- **Global Utility (Ours)**: empirically effective measure on **any dataset**

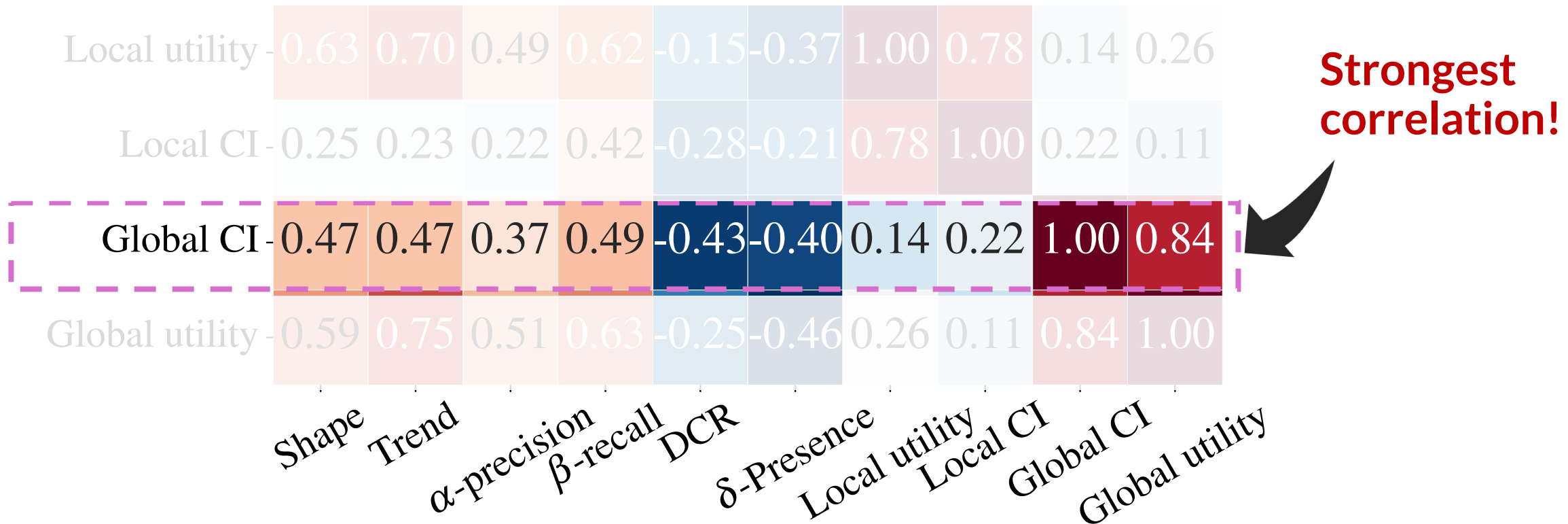
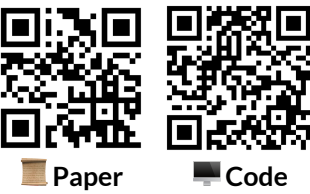


$$\text{Utility}_j(\mathcal{D}) := \begin{cases} \text{Perf}_j(\mathcal{D}_{\text{ref}})^{-1} \text{Perf}_j(\mathcal{D}), & \text{if } \mathbf{x}_j \text{ is categorical,} \\ \text{Perf}_j(\mathcal{D})^{-1} \text{Perf}_j(\mathcal{D}_{\text{ref}}), & \text{if } \mathbf{x}_j \text{ is numerical.} \end{cases}$$

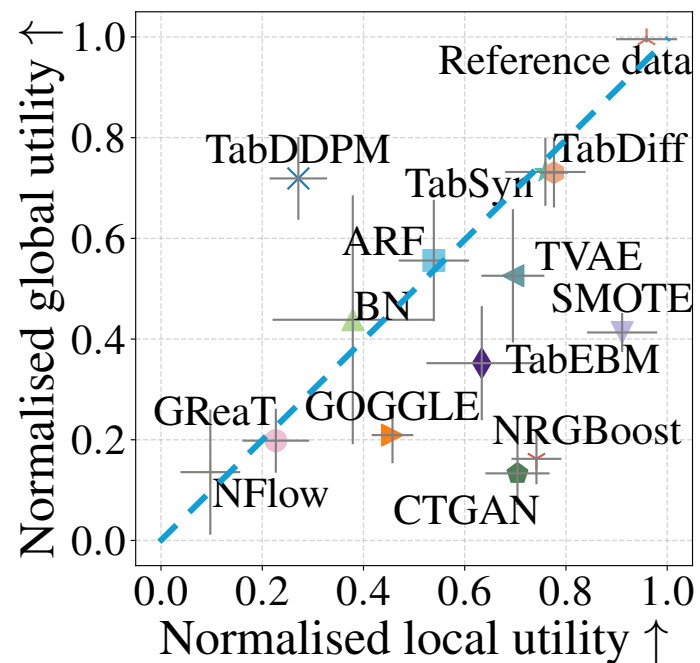
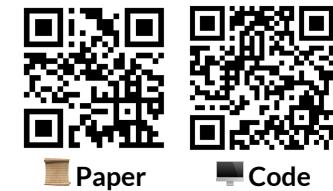
Computation of the utility per feature

- **Global Utility (Ours):** empirically effective measure on **any dataset**
- **Full mathematical details available in paper**

# Global utility provides a valid and novel lens

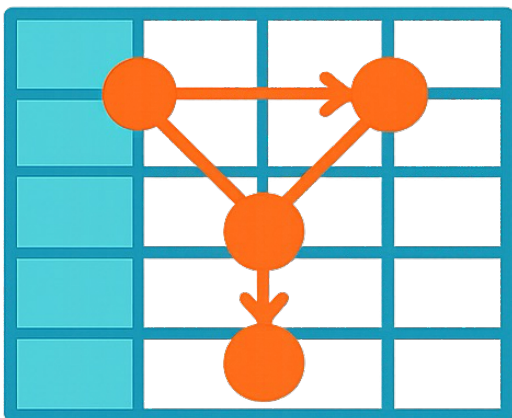


Spearman's rank correlation heatmap based on metric values on six SCM datasets



Mean normalised local utility vs. global utility on 23 real-world datasets

- Tabular generative models should capture the **global causal structures**
- More insights **available in paper**



# TabStruct

Structural Fidelity of Tabular Data



Code

- **29** challenging datasets × **13** generative models × **4** evaluation dimensions
- **150K** evaluations in total

# THANKS !

For more details, check our paper and code!



 Paper



 Code

Or come for a chat at  
**Poster Session 2**  
**Pavilion 3-#123**

2026/04/23 3:15 PM – 5:45 PM

Reach out to [xj265@cam.ac.uk](mailto:xj265@cam.ac.uk)



**Xiangjiang  
Jiang**



**Nikola  
Simidjievski**



**Mateja  
Jamnik**