

Geometry Based Data Generation

Ofir Lindenbaum^{1,*}, Jay S. Stanley III^{2,*}, Guy Wolf^{1,†} & Smita Krishnaswamy^{1,2,3,†}

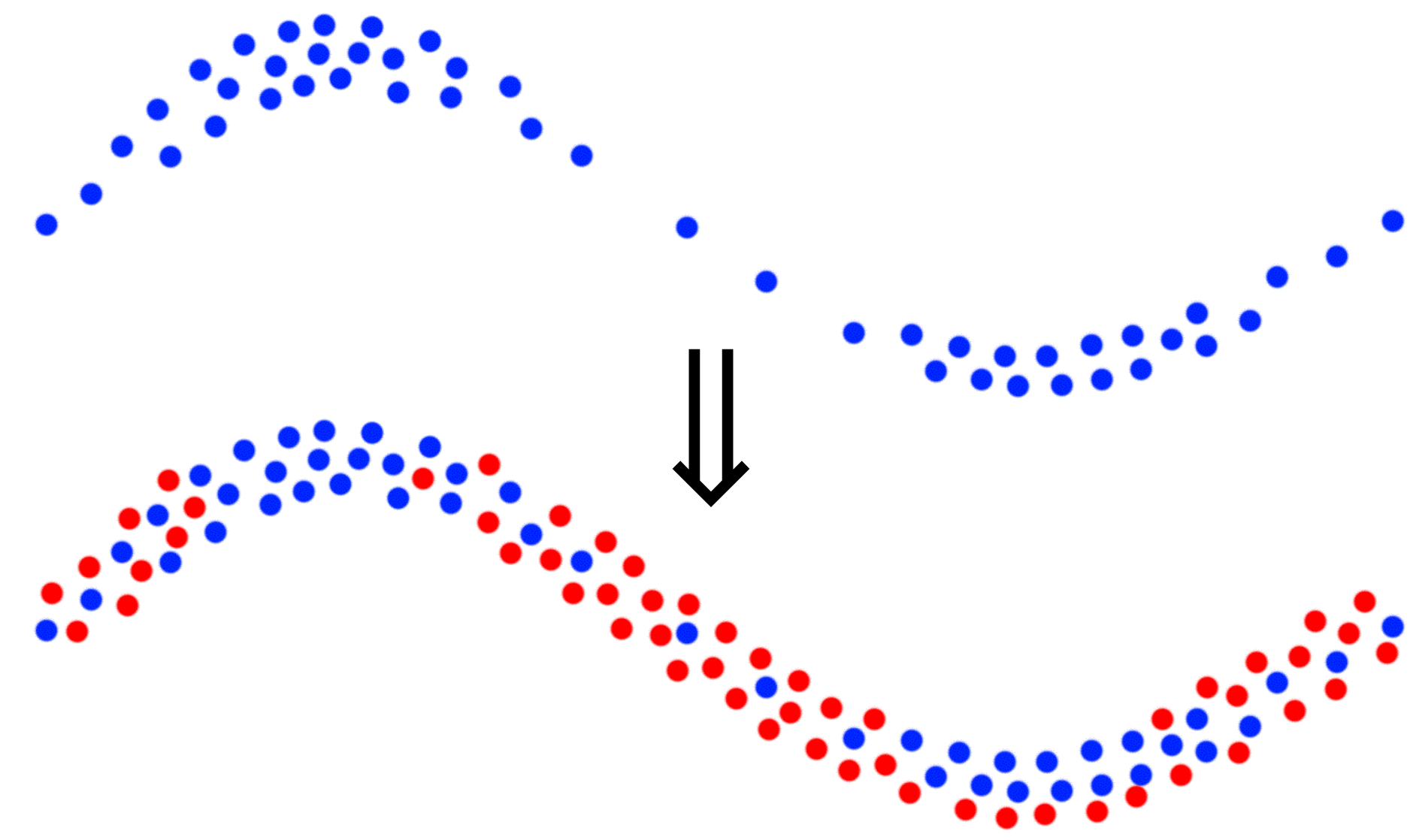
¹ Applied Mathematics Program, ² Comp. Bio. & Bioinformatics Program, ³ Depts. of Genetics & Computer Science, Yale University.

* Equal contributions; † equal contributions.

Yale

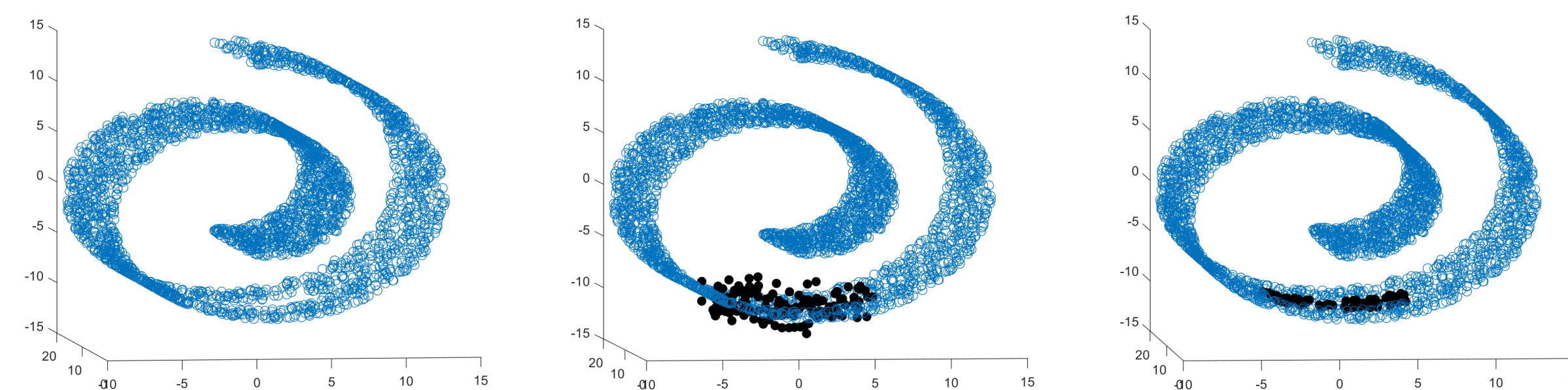
Motivation

Generate points uniformly from intrinsic data geometry / manifold:

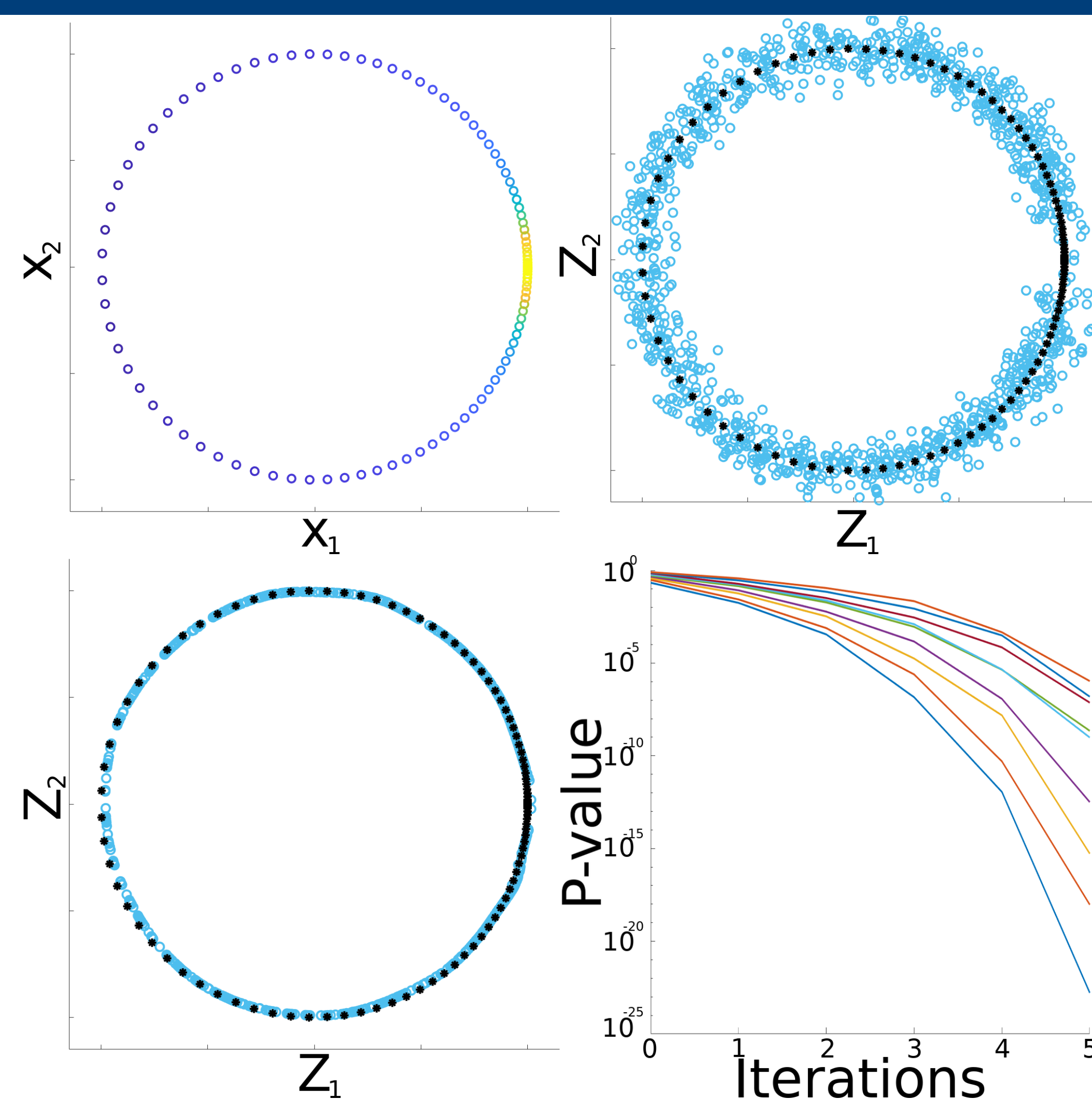


SUGAR

- Compute MGC [2] kernel: $\hat{K}(x, y) = \sum_{r \in \text{data}} \frac{\kappa(x, r) \kappa(r, y)}{\text{density}(r)}$
- Define Markovian random walks: $\Pr[x \xrightarrow{1 \text{ step}} y] = \frac{\hat{K}(x, y)}{\|\hat{K}(x, \cdot)\|_1}$
- Initialize new points: $x \sim \mathcal{N}(y, \Sigma_y)$, $y \in \text{data}$
- Walk towards the data manifold: $x \mapsto \sum_{y \in \text{data}} y \cdot \Pr[x \xrightarrow{t \text{ steps}} y]$

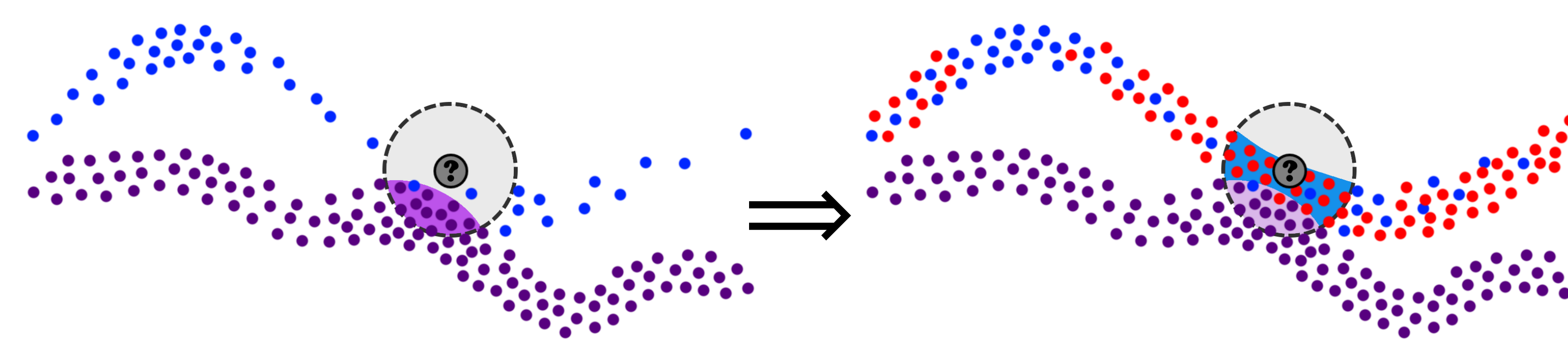


Manifold Equalization



SUGAR recovers undersampled regions according to manifold geometry.

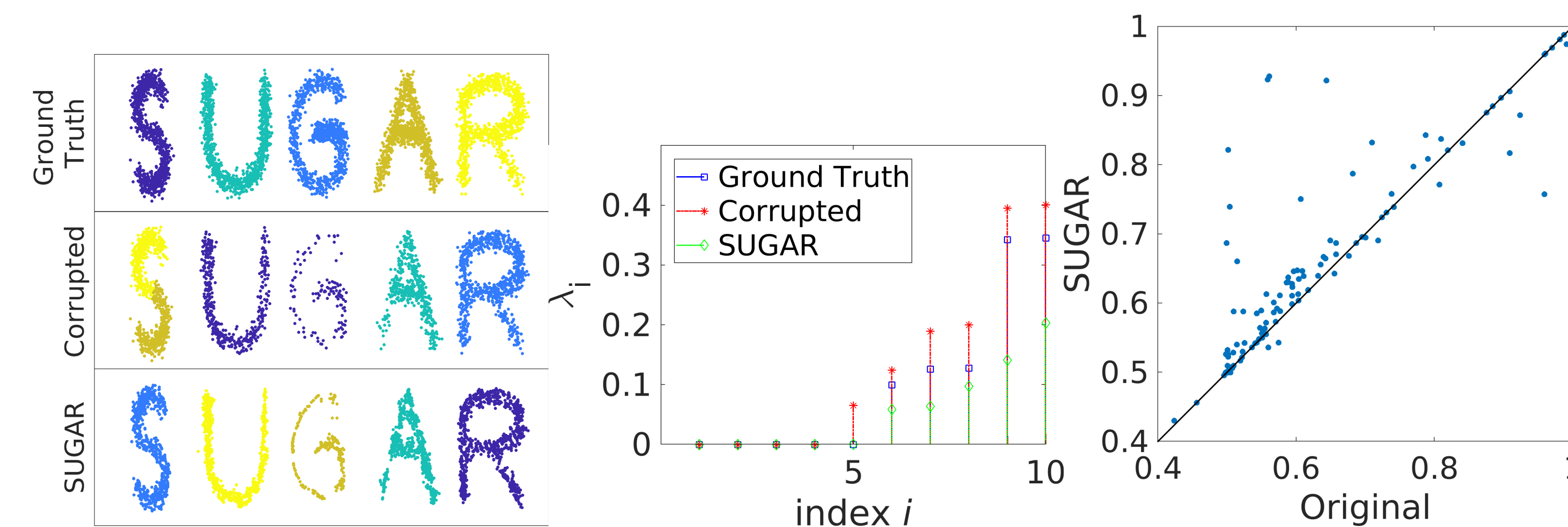
Classification



	k-NN			SVM			RUSBoost
	Orig	SMOTE	SUGAR	Orig	SMOTE	SUGAR	
ACP	0.67	0.76	0.78	0.77	0.77	0.78	0.75
ACR	0.64	0.73	0.77	0.78	0.78	0.84	0.81
MCC	0.66	0.74	0.78	0.78	0.78	0.84	0.80

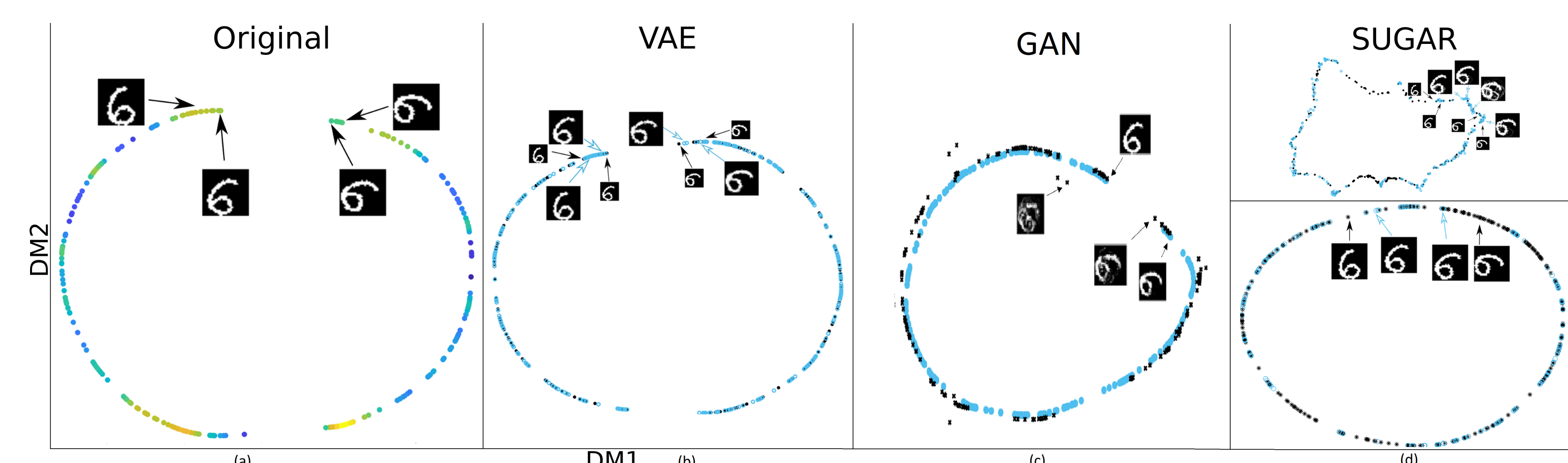
61 Imbalance datasets– Average class precision (ACP), class recall (ACR), and the Matthews correlation coefficient (MCC) for kNN and kernel SVM classifiers (using 10-fold cross validation) before / after SMOTE and SUGAR, and for RUSBoost classification.

Clustering



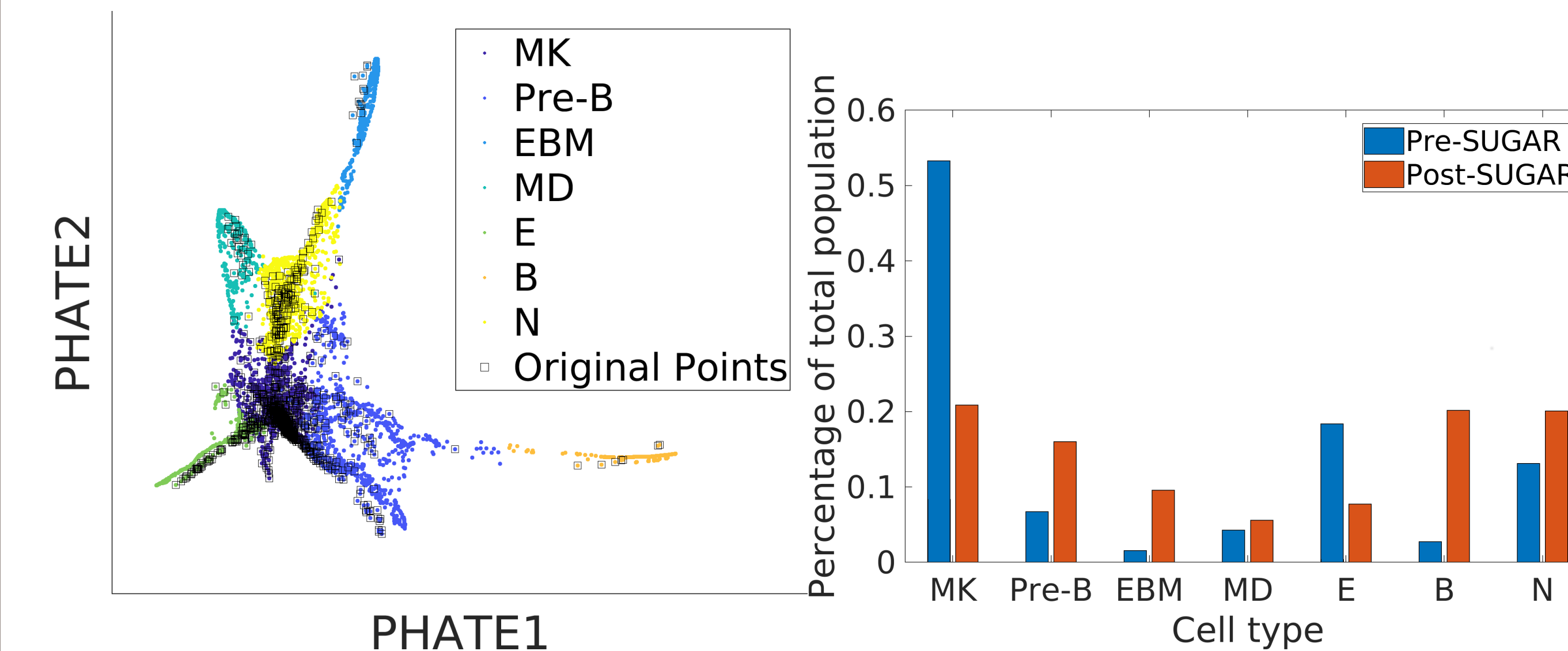
SUGAR enhances clustering accuracy **Left, top:** A ground truth SUGAR manifold was generated. **Left, middle:** Non uniform sampling of the ground truth manifold leads to inaccurate spectral clusters. **Left, bottom:** Spectral clusters are restored following SUGAR augmentation. **Middle:** The eigenvalues of the ground truth, corrupted, and SUGAR corrected graphs reveal that SUGAR restored the disconnected letters of the graph, evidenced by the multiplicity of the 0 eigenvalue. **Right:** Rand index scores of k-means before and after applying SUGAR on 115 datasets from [1].

MNIST Rotations



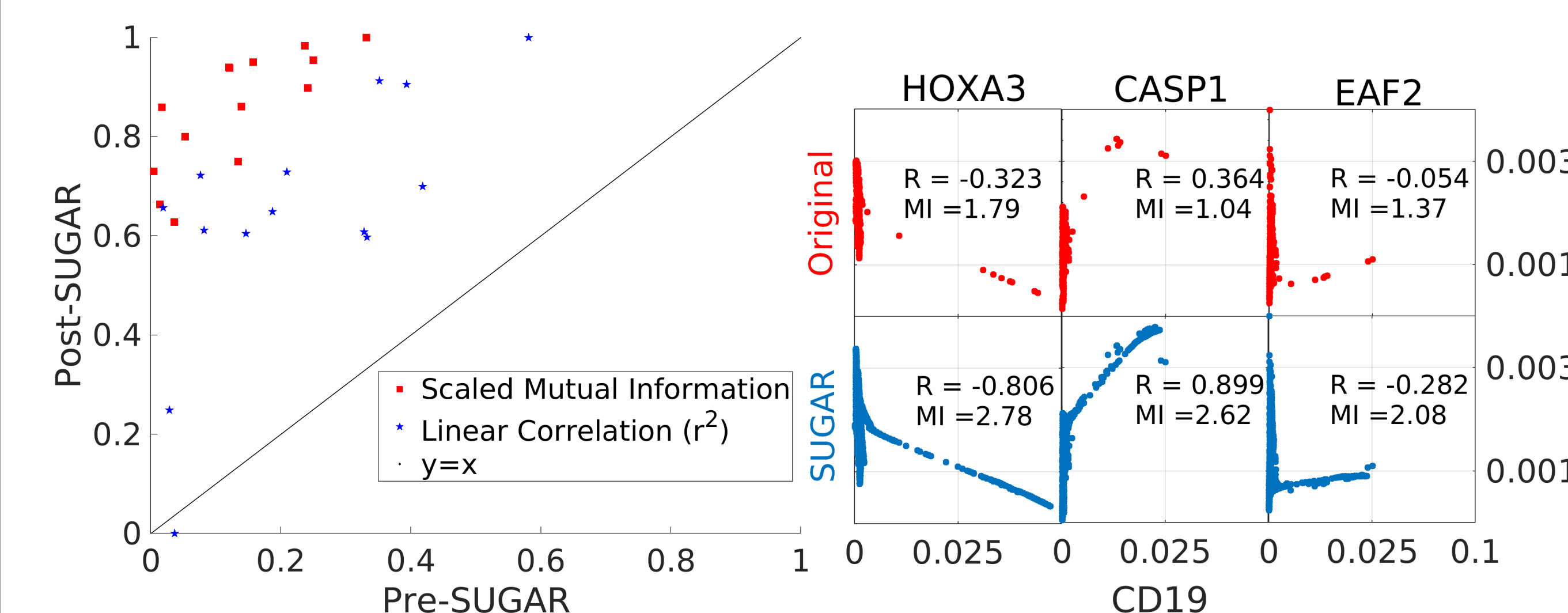
SUGAR recovers undersampled regions according to manifold geometry.

Hypothetical Population Exploration [4]



SUGAR illuminates hypothetical cell types in data. [4] explore a continuum model of early hematopoiesis. SUGAR generated new points in undersampled cell types and restored an absent late-B cell population. **Left:** PHATE embedding of the data. **Right:** Cell type distribution before and after SUGAR.

Canonical Gene-Gene Relationships [4]



SUGAR recovers gene-gene relationships. **Left:** Intra-module mutual information and linear correlation of modules discussed in [4]. **Right:** Two canonically correlated genes for B cell maturation (CD19), HOXA3 (negative) and CASP1 (positive), are recovered by SUGAR. [4] focus on EAF2 as a marker for neutrophils and monocytes, however, data after SUGAR illuminates a correlation in late B cells shown in [3].

References

- [1] J. Alcalá-Fdez, L. Sanchez, S. Garcia, M. J. del Jesus, S. Ventura, J. M. Garrell, J. Otero, C. Romero, J. Bacardit, V. M. Rivas, et al. Keel: a software tool to assess evolutionary algorithms for data mining problems. *Soft Computing*, 13(3):307–318, 2009.
- [2] A. Bermanis, M. Salhov, G. Wolf, and A. Averbuch. Measure-based diffusion grid construction and high-dimensional data discretization. *Applied and Computational Harmonic Analysis*, 40(2):207–228, 2016.
- [3] Y. Li, Y. Takahashi, S.-i. Fujii, Y. Zhou, R. Hong, A. Suzuki, T. Tsubata, K. Hase, and J.-Y. Wang. Eaf2 mediates germinal centre b-cell apoptosis to suppress excessive immune responses and prevent autoimmunity. *Nature communications*, 7:10836, 2016.
- [4] L. Velten, S. F. Haas, S. Raffel, S. Blaszkiewicz, S. Islam, B. P. Hennig, C. Hirche, C. Lutz, E. C. Buss, D. Nowak, et al. Human haematopoietic stem cell lineage commitment is a continuous process. *Nat. Cell Biol*, 19:271–281, 2017.