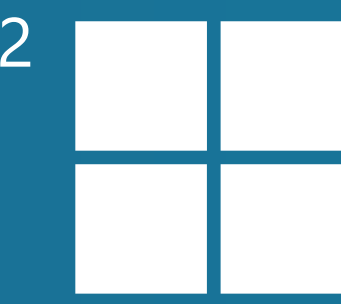


I-Con: A Unifying Framework for Representation Learning

Shaden Alshammari¹, John R. Hershey³, Axel Feldmann¹, William T. Freeman^{1,3}, Mark Hamilton^{1,2}



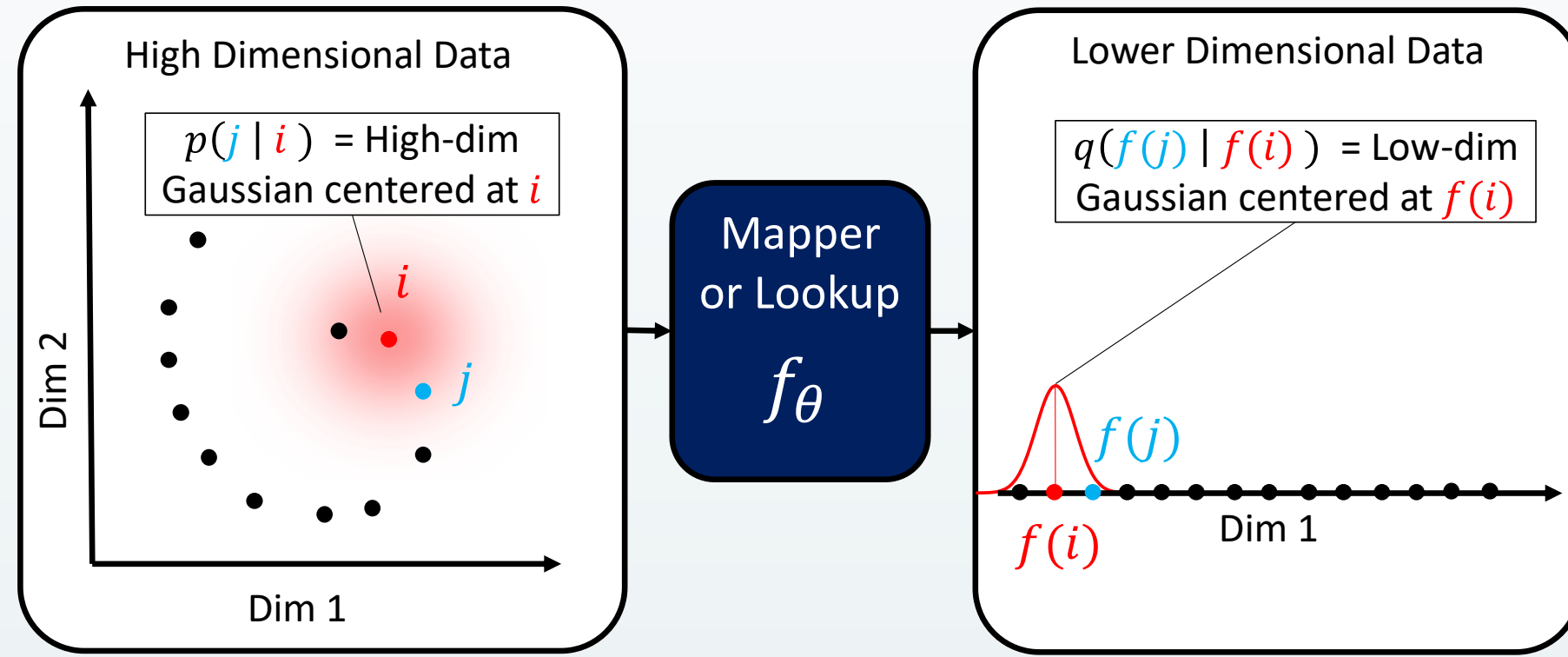
Microsoft



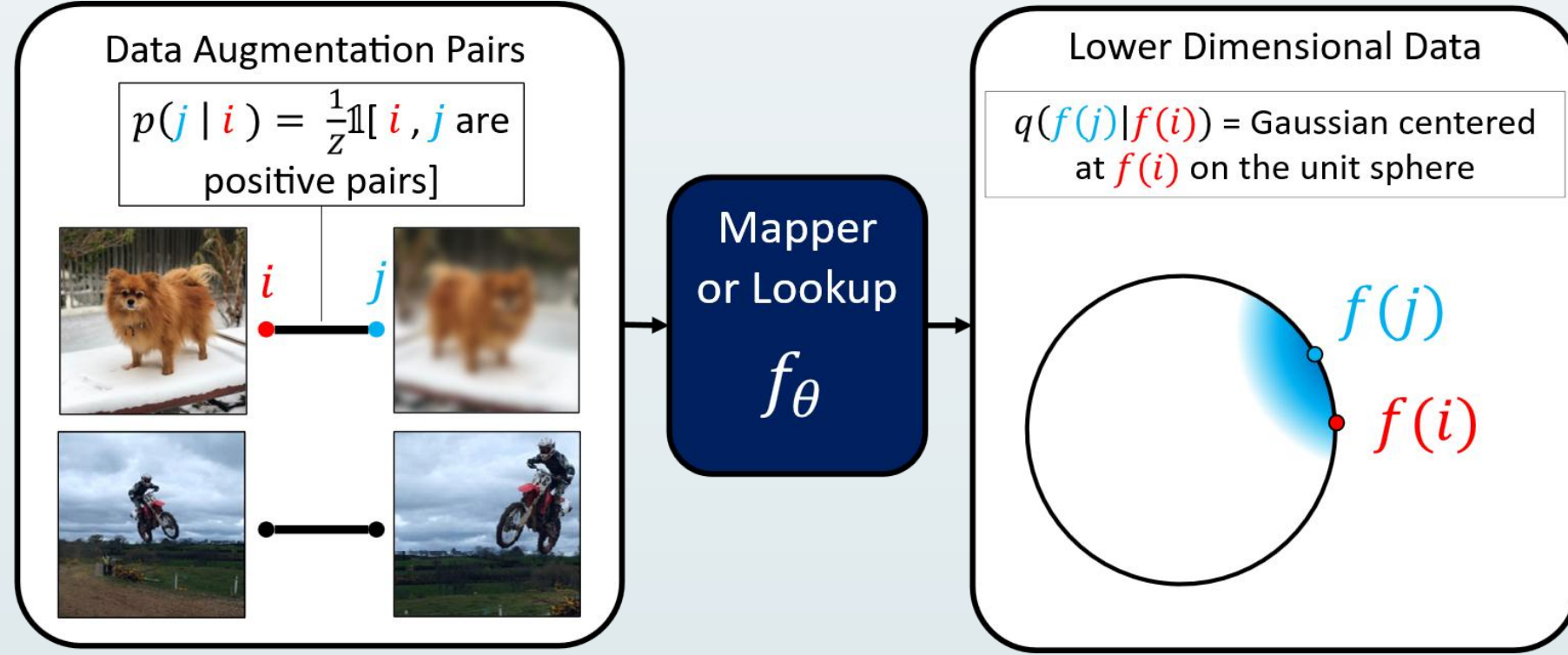
aka.ms/i-con

Unifying Clustering, Dim. Reduction, Contrastive Learning, and Spectral Graph Methods

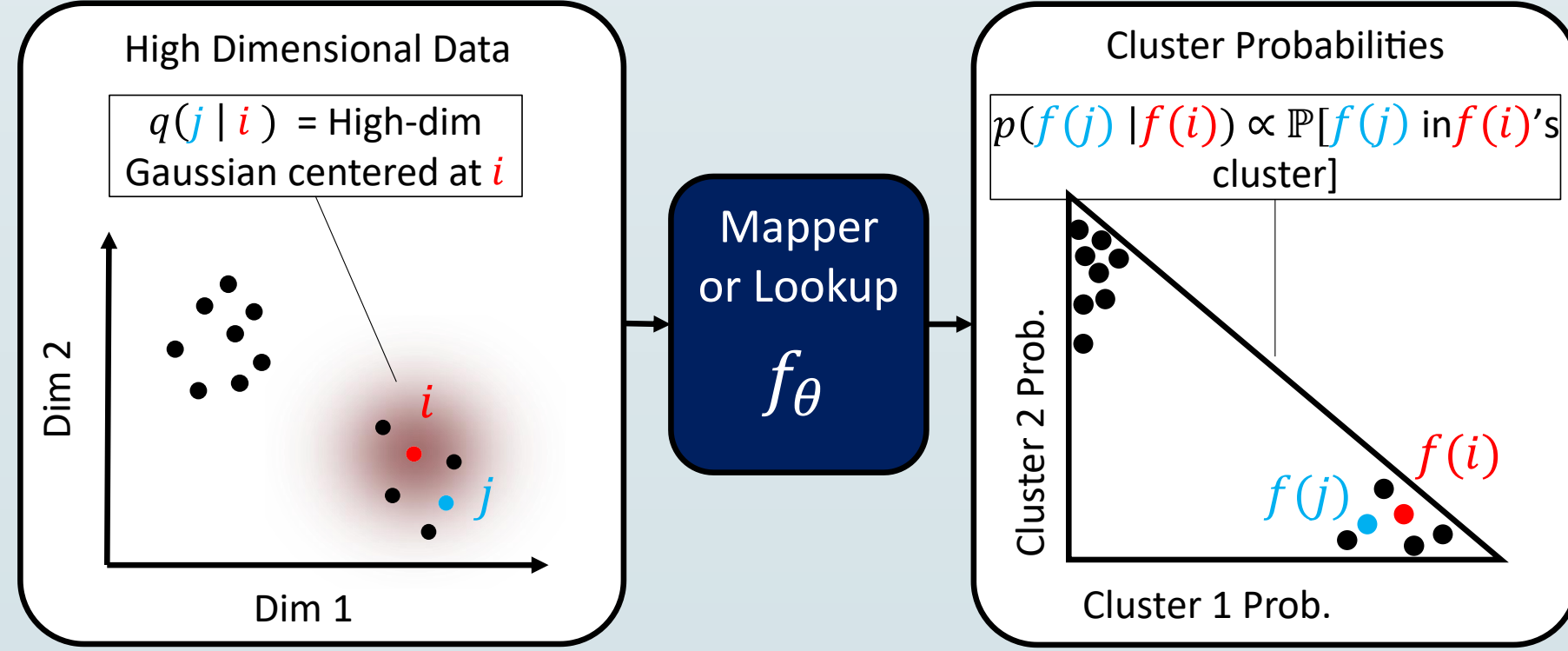
Dimensionality Reduction (SNE)



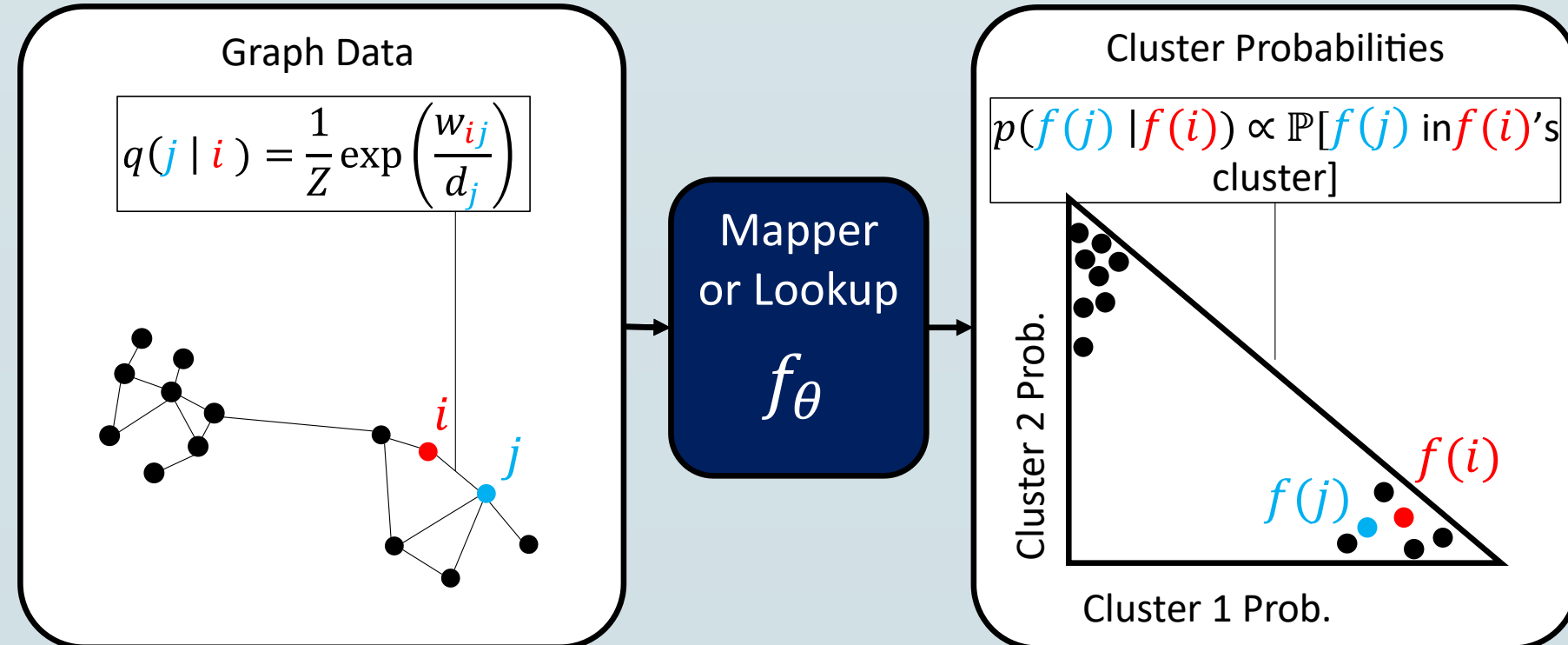
Contrastive Learning (InfoNCE)



Clustering (K-Means)



Spectral Clustering



We introduce a **periodic table of machine learning** based on a single equation that **unifies >20 loss functions**. We use this to derive a new state-of-the-art unsupervised image classifier.

The Information-Contrastive (I-Con) Loss:

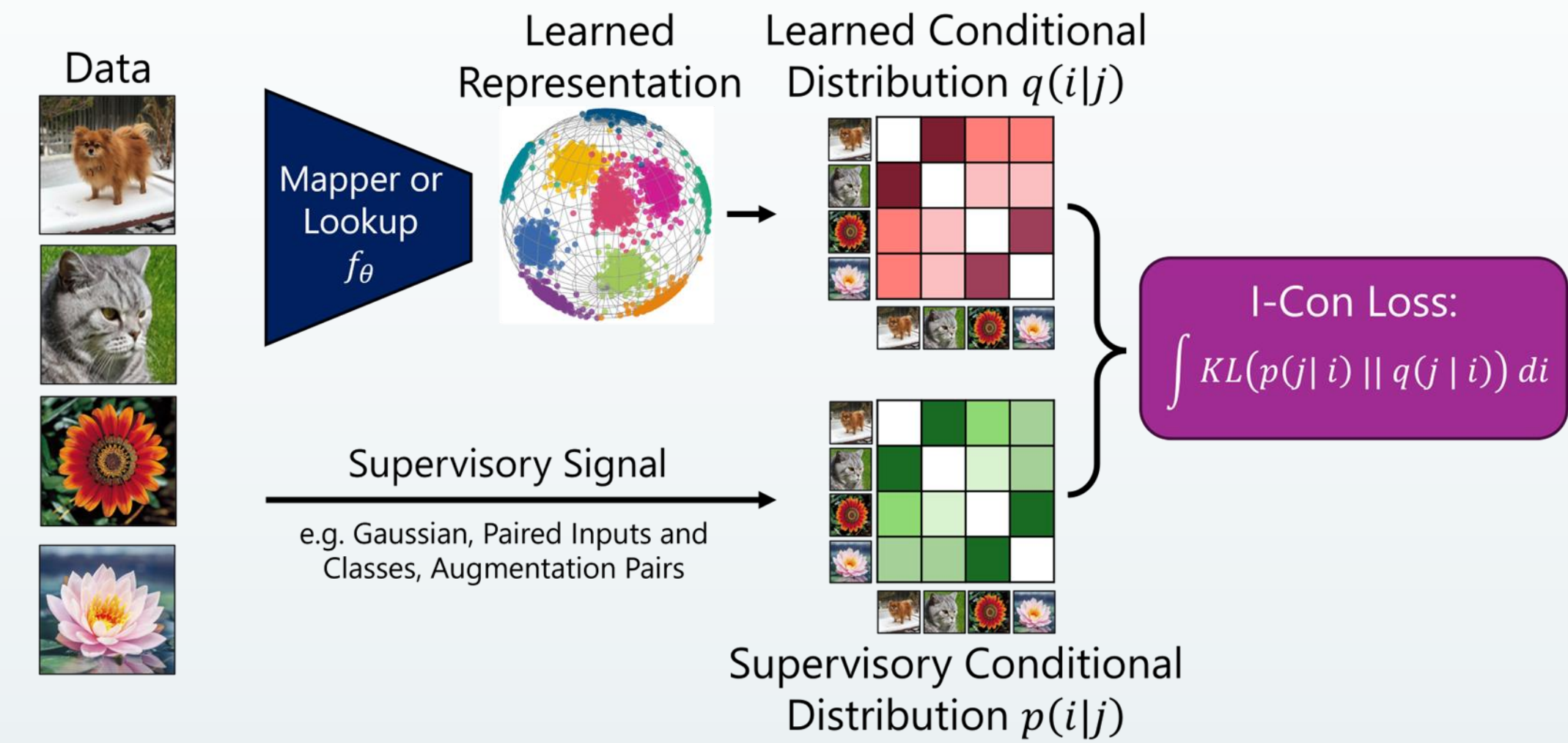
$$L = \int KL(p(j|i) \parallel q(j|i)) di$$

Supervisory Signal

	Gaussian	Student-T	Identity	Graph Kernel Weights	Uniform over K-Neighbors	Uniform over Positive Pairs	Cross-Modal Pairs	Uniform over Classes	Data-Label Pairs
Gaussian	SNE [Hinton 2002]	Dual t-SNE		SNE Graph Embeddings	SNE with Uniform Affinities	InfoNCE [Bachman 2019]	CLIP [Radford 2021]	SupCon [Khosla 2020]	Cross Entropy [Good 1963]
Gaussian $\sigma \rightarrow \infty$	X-Sample CL [Sobal 2025]				LGSimCLR [El Banani 2023]	SimCLR [Chen 2020]			
Gaussian $\sigma \rightarrow 0$						MoCoV3 [Chen 2021]	CMC [Tian 2020]		
Student-T	t-SNE [Van der Maaten 2008]	Doubly t-SNE		t-SNE Graph Embedding	t-SNE with Uniform Affinities	t-SimCNE [Böhm 2023]	t-CLIP	t-SupCon	Harmonic Loss [Baek 2025]
Cluster Probabilities	K-Means [MacQueen 1967]	t K-Means		Normalized Cuts [Shi 2000]	DCD [Yang 2012]	NEW InfoNCE Clustering [Ours]			

■ Dimensionality Reduction ■ Cluster Learning ■ Unimodal SSL ■ Multimodal SSL ■ Supervised Learning ■ Interpretation of Gaps

Architecture



Over 15 Theorems

Method	Choice of $p(j i)$	Choice of $q(j i)$
SNE [Hinton & Roweis, 2002]	Gaussian over data points, x_i	Gaussian over learned low-dimensional points, ϕ_i
t-SNE [Van der Maaten & Hinton, 2008]	Gaussian over data points, x_i	Gaussian over learned low-dimensional points, ϕ_i
PCA [Pearson, 1901]	$1[i=j]$	Wide Gaussian on linear projection features, $f_\theta(x_i)$
InfoNCE Loss [Bachman et al., 2019]	$\frac{1}{2} 1[i \text{ and } j \text{ are a positive pair}]$	Gaussian on deep normalized features, $f_\theta(x_i)$
Triplet Loss [Schultz et al., 2014]	$\frac{1}{2} 1[i \text{ and } j \text{ are a positive pair}]$	Gaussian on deep normalized features, $f_\theta(x_i)$
LGSimCLR [El Banani et al., 2023]	$\frac{1}{2} 1[i \text{ and } j \text{ are a positive pair}]$	Gaussian on deep normalized features, $f_\theta(x_i)$
X-Sample [Sobal et al., 2025]	$\frac{1}{2} 1[i \text{ and } j \text{ are a positive pair}]$	Gaussian on deep normalized features, $f_\theta(x_i)$
CMC & CLIP [Tian et al., 2020]	$\frac{1}{2} 1[i \text{ and } j \text{ are a positive pair}]$	Gaussian on deep normalized features, $f_\theta(x_i)$

Unsupervised Image Classification

Method	DiNo ViT-S/14	DiNo ViT-B/14	DiNo ViT-L/14
k-Means	51.84	52.26	53.36
Contrastive Clustering	47.35	55.64	59.84
SCAN	49.20	55.60	60.15
TEMI	56.84	58.62	—
Debiased InfoNCE Clustering (Ours)	57.8 ± 0.26	64.75 ± 0.18	67.52 ± 0.28

Unsupervised (Hungarian) Accuracy on ImageNet-1K Validation Set