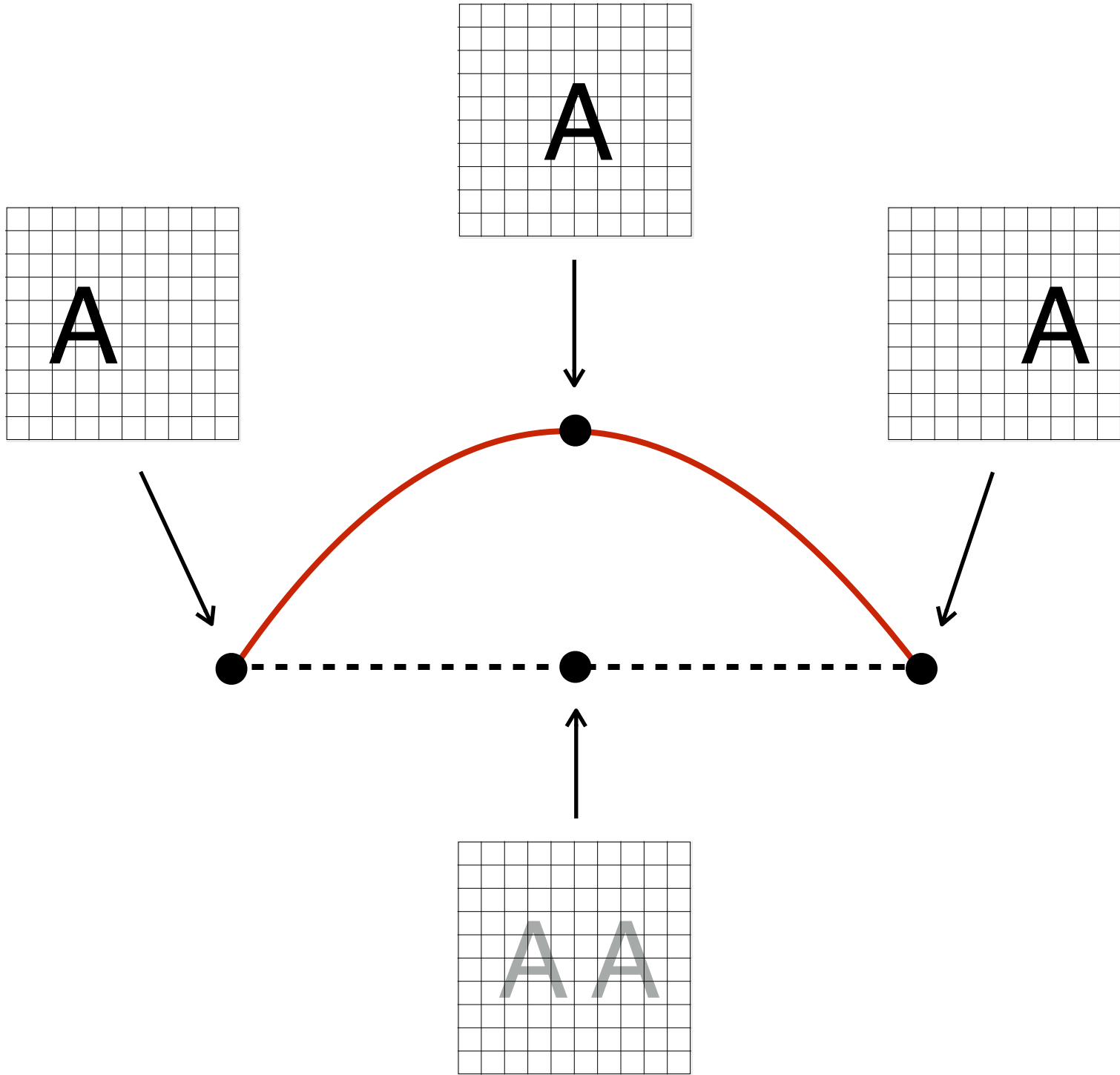
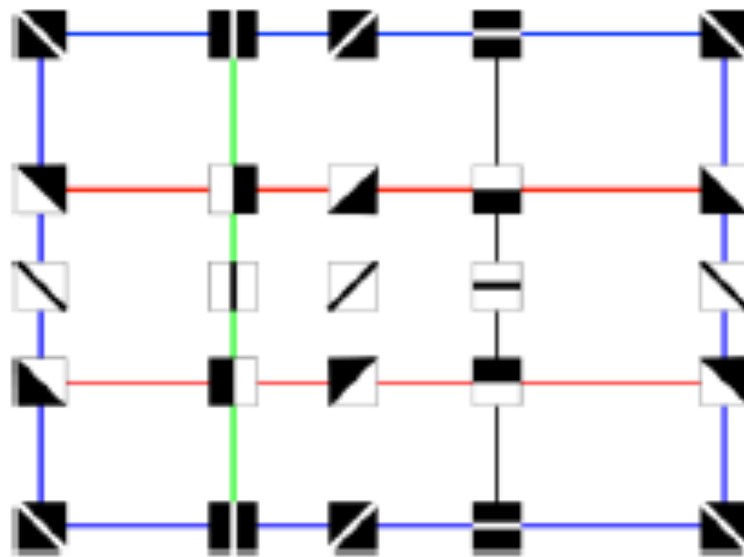
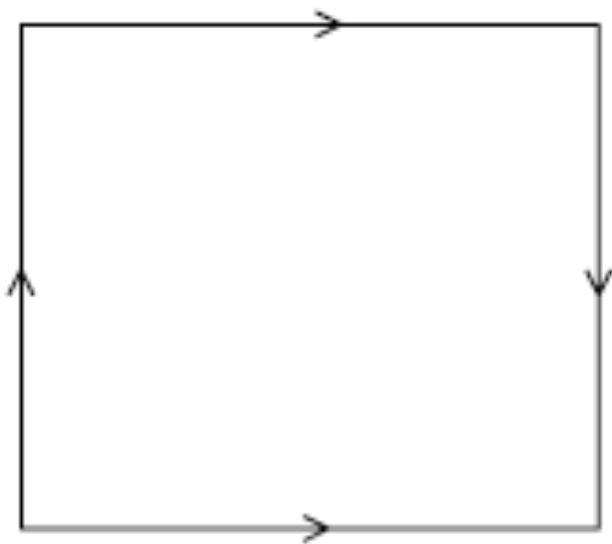
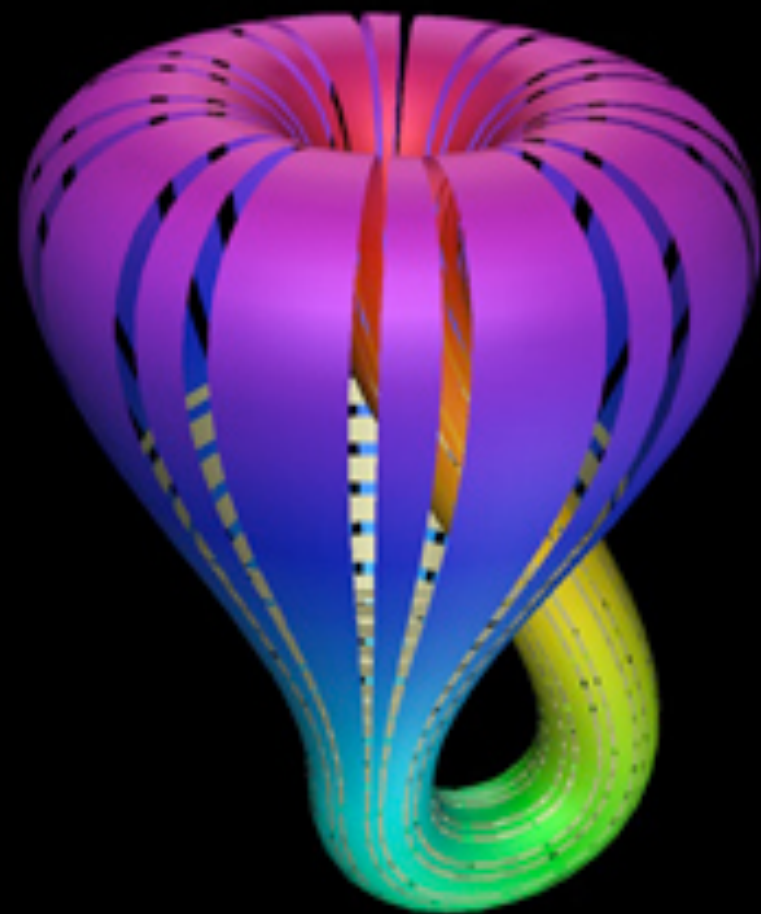


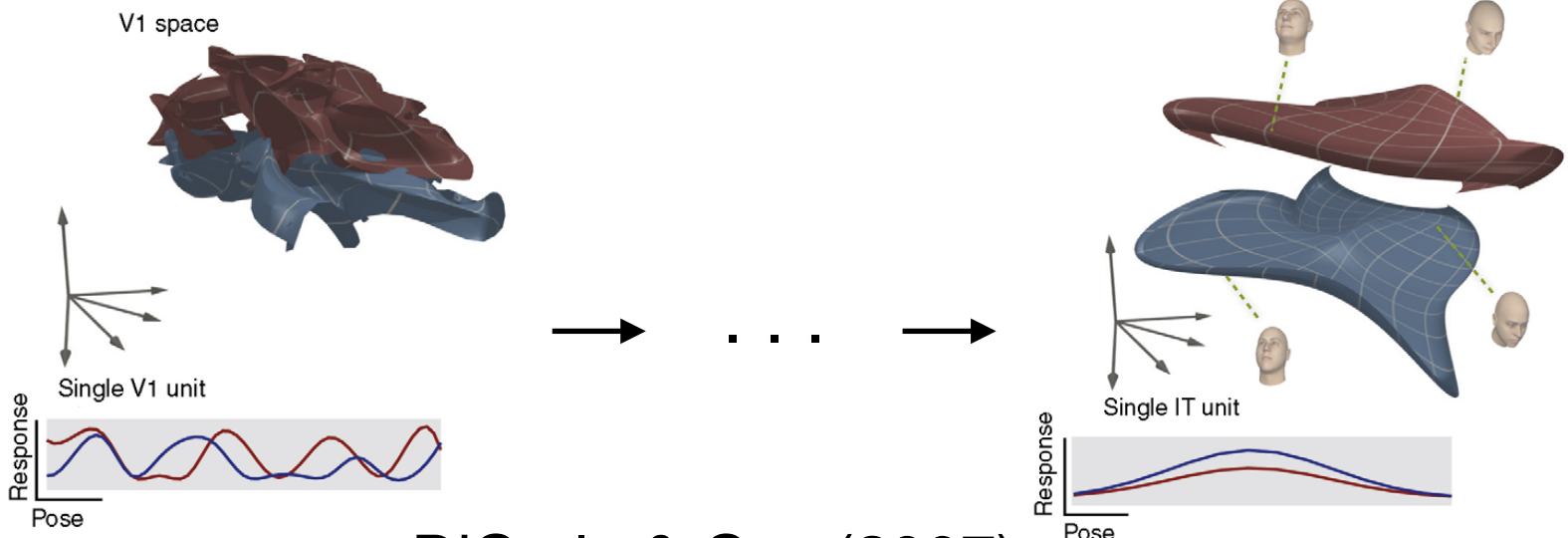
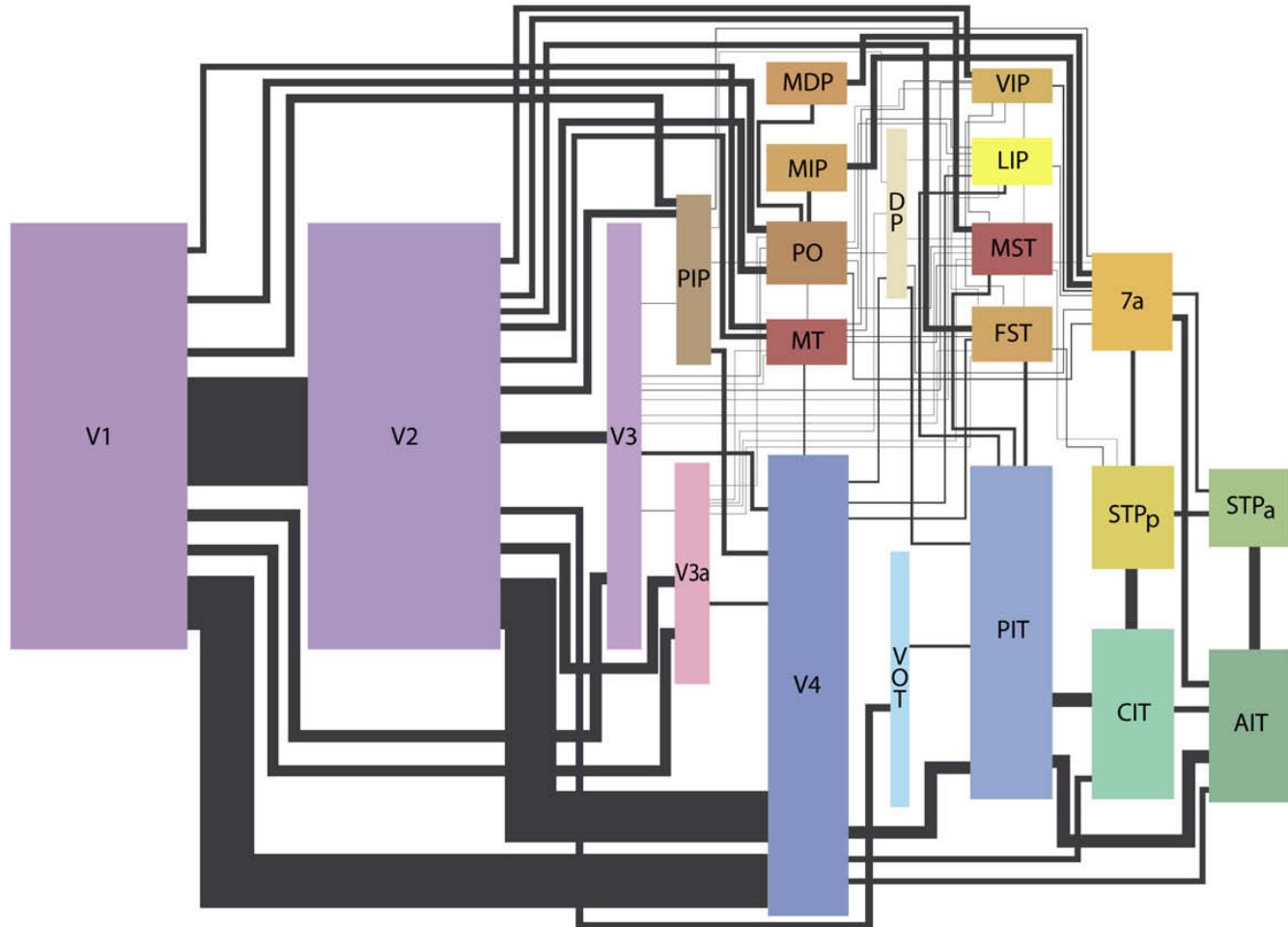
Manifold learning



Distribution of 3x3 pixel image patches drawn from natural scenes forms a Klein bottle
(Carlsson et al., 2008)







DiCarlo & Cox (2007)

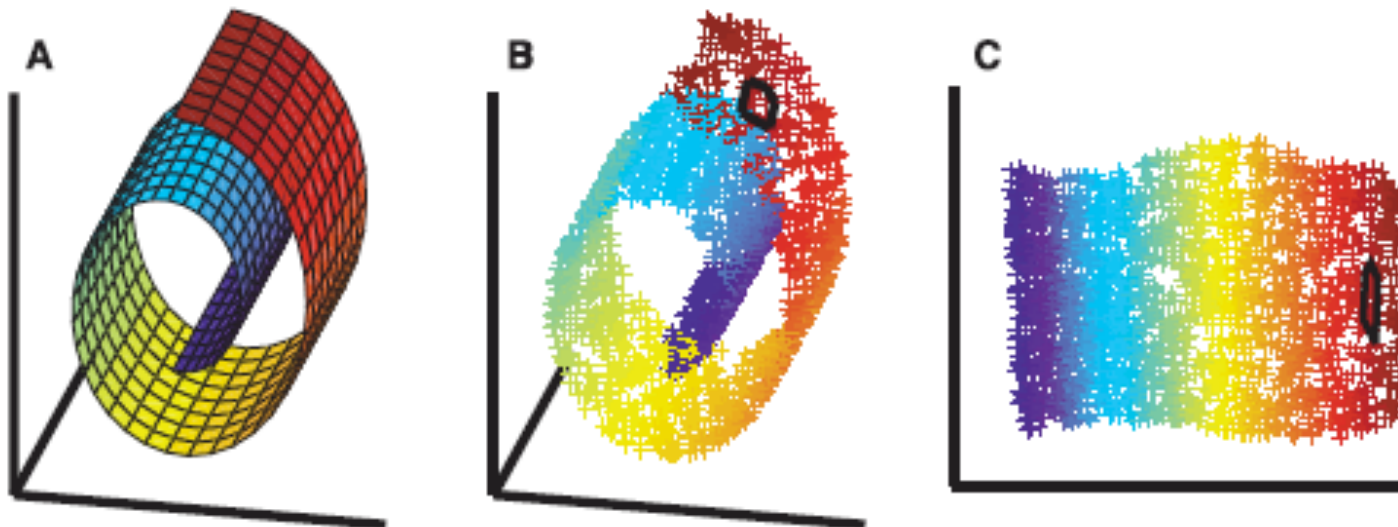
Nonlinear Dimensionality Reduction by Locally Linear Embedding

Sam T. Roweis¹ and Lawrence K. Saul²

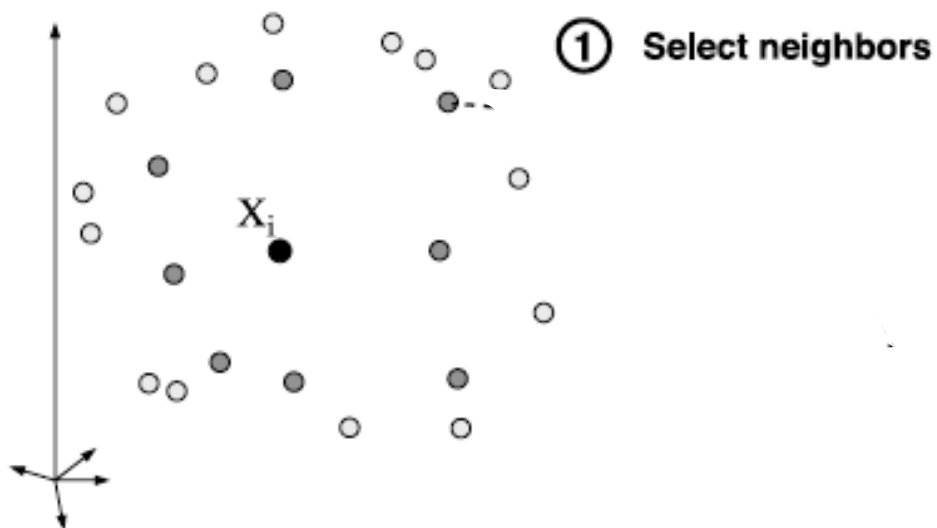
A Global Geometric Framework for Nonlinear Dimensionality Reduction

Joshua B. Tenenbaum,^{1*} Vin de Silva,² John C. Langford³

Science, 22 Dec. 2000



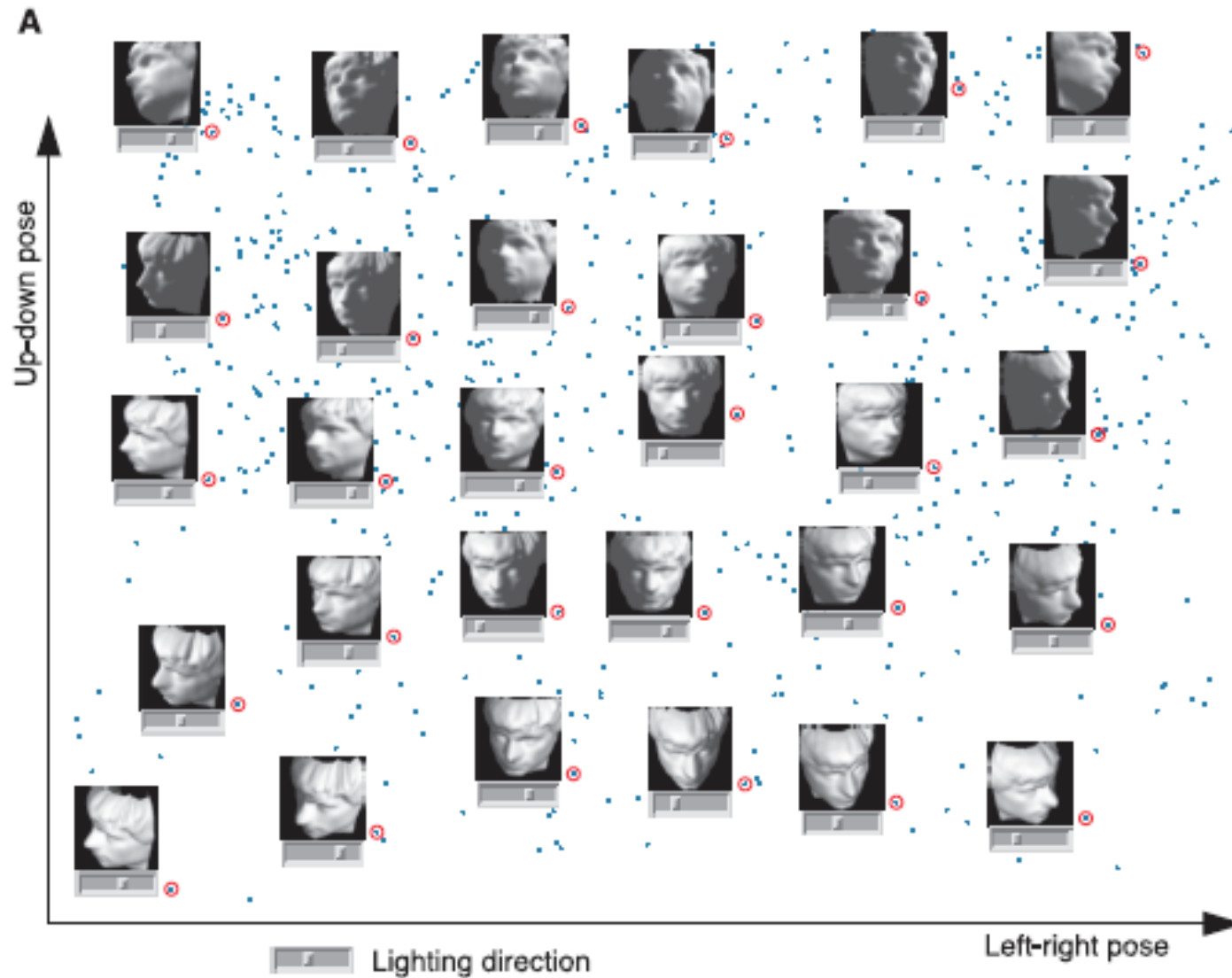
Local Linear Embedding (LLE)



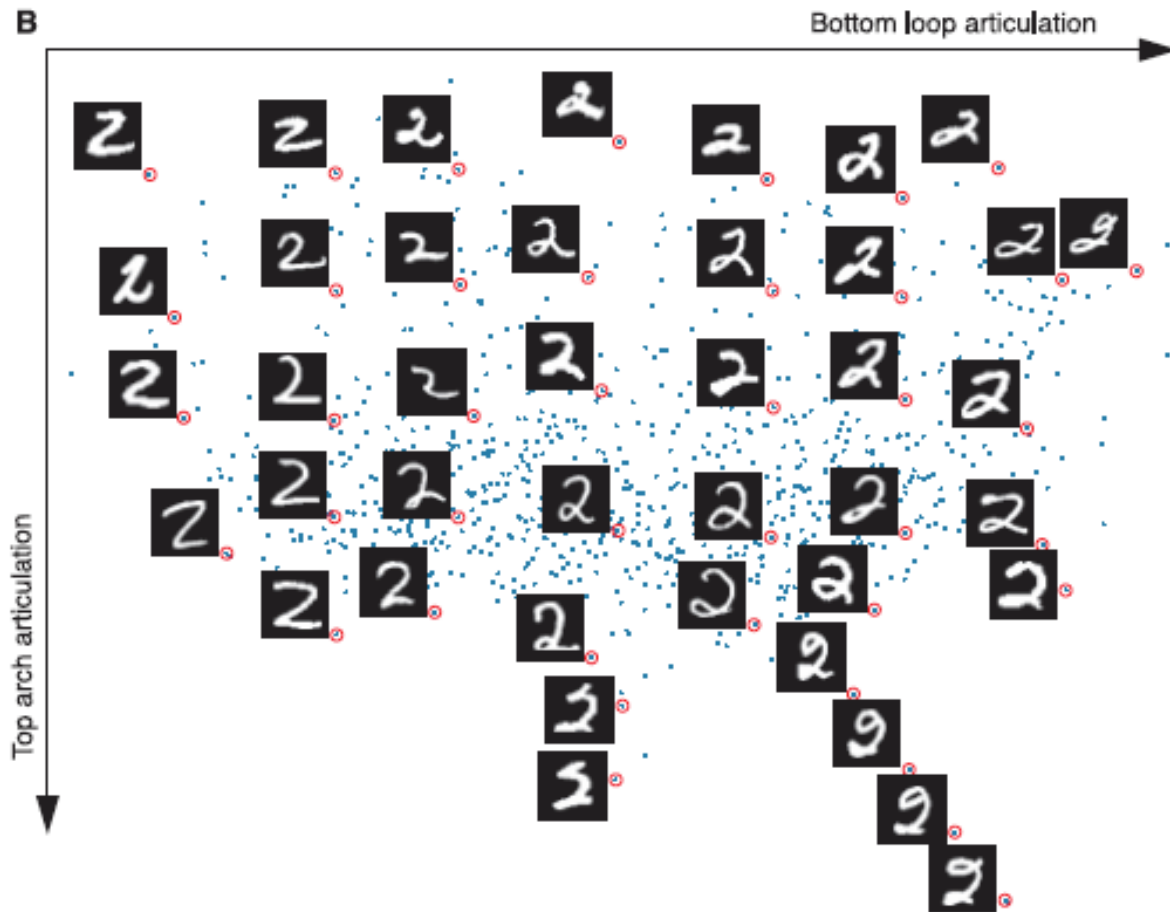
$$\varepsilon(W) = \sum_i \left| \vec{X}_i - \sum_j W_{ij} \vec{X}_j \right|^2$$

$$\Phi(Y) = \sum_i \left| \vec{Y}_i - \sum_j W_{ij} \vec{Y}_j \right|^2$$

Manifold of facial pose and lighting

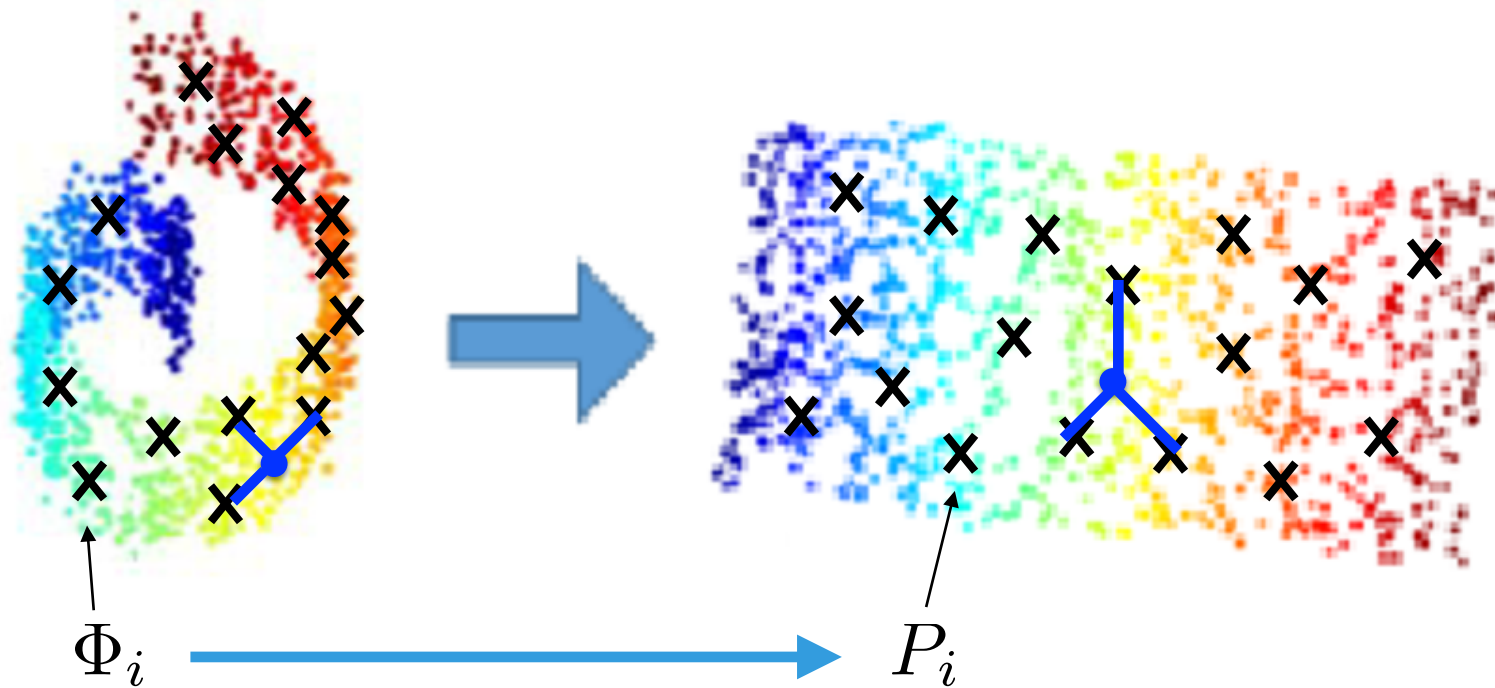


Hand-written digits



Local Linear Landmarks (LLL)

(Vladymyrov & Carreira-Perpinán, 2013)

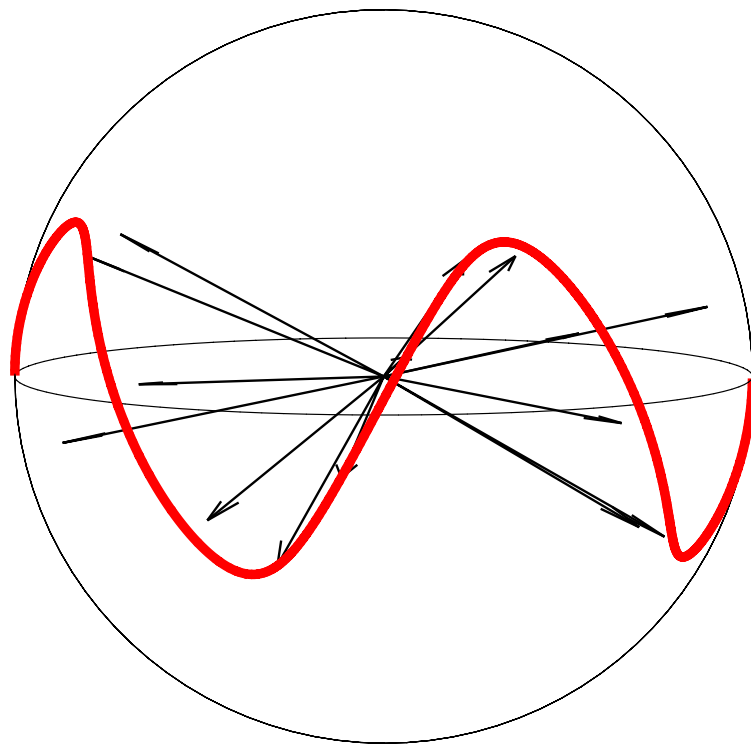


$$x = \Phi \alpha + n \quad \longrightarrow \quad y = P \alpha$$

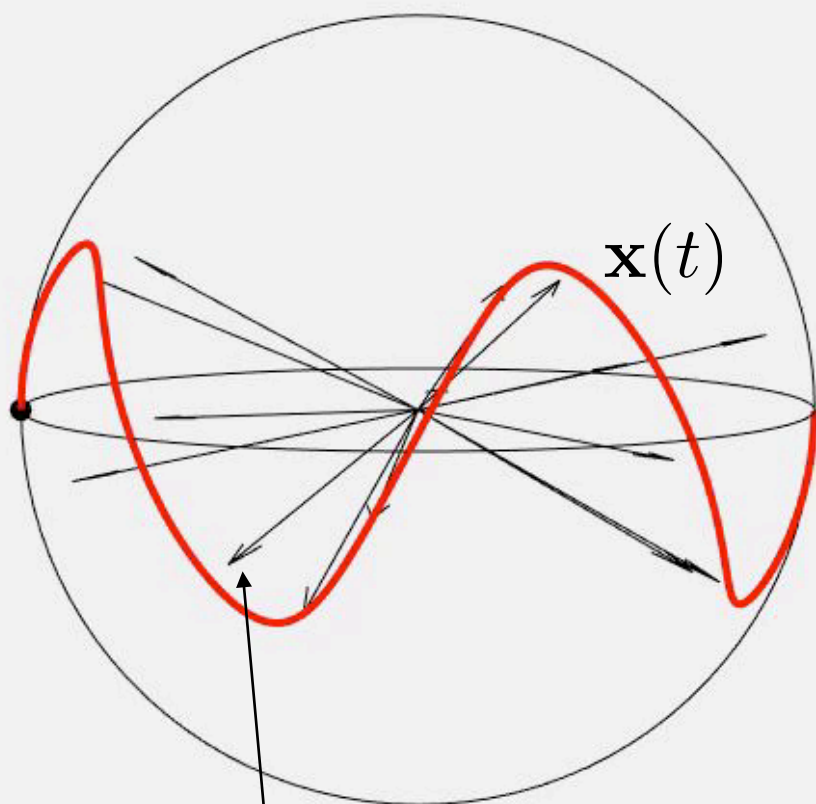
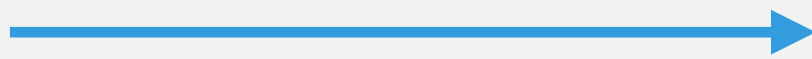
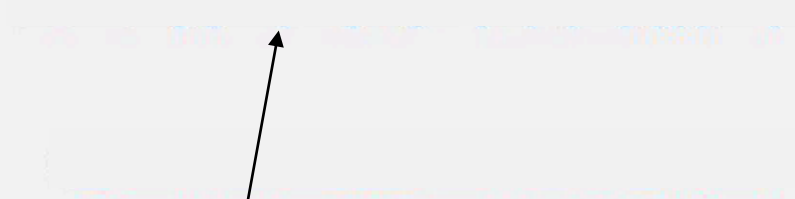
local linear approximation

embedding

Basis functions learned by sparse coding form a locally linear approximation to the manifold of natural images



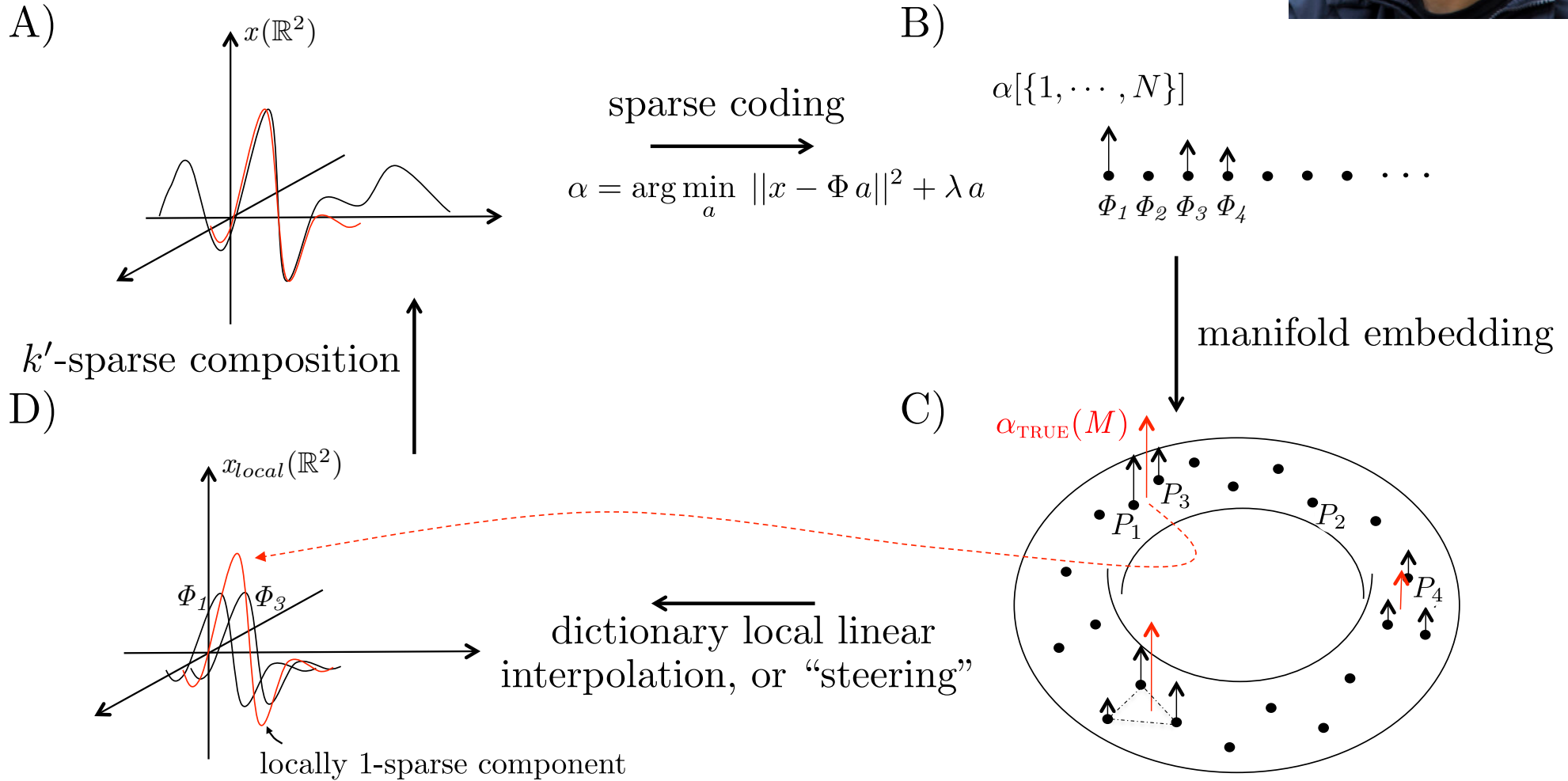
$$\mathbf{x}(t) = \Phi \alpha(t) + \mathbf{n}(t)$$

 Φ_i  P_i 

$$\beta(t) = P \alpha(t)$$

Sparse Manifold Transform

(Yubei Chen, Ph.D. thesis; Neurips 2018)



Unsupervised learning principles

- Linear Hebbian learning → PCA
- Competitive Hebbian learning → clustering
- Sparse coding → feature learning
- Self-organizing map → topographic maps
- Sparse manifold transform → manifold learning
- Slow feature analysis → invariance