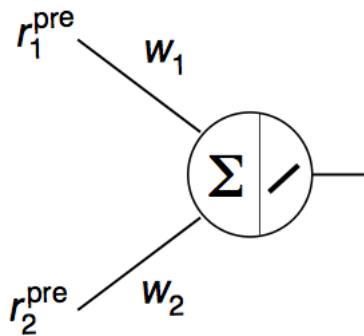
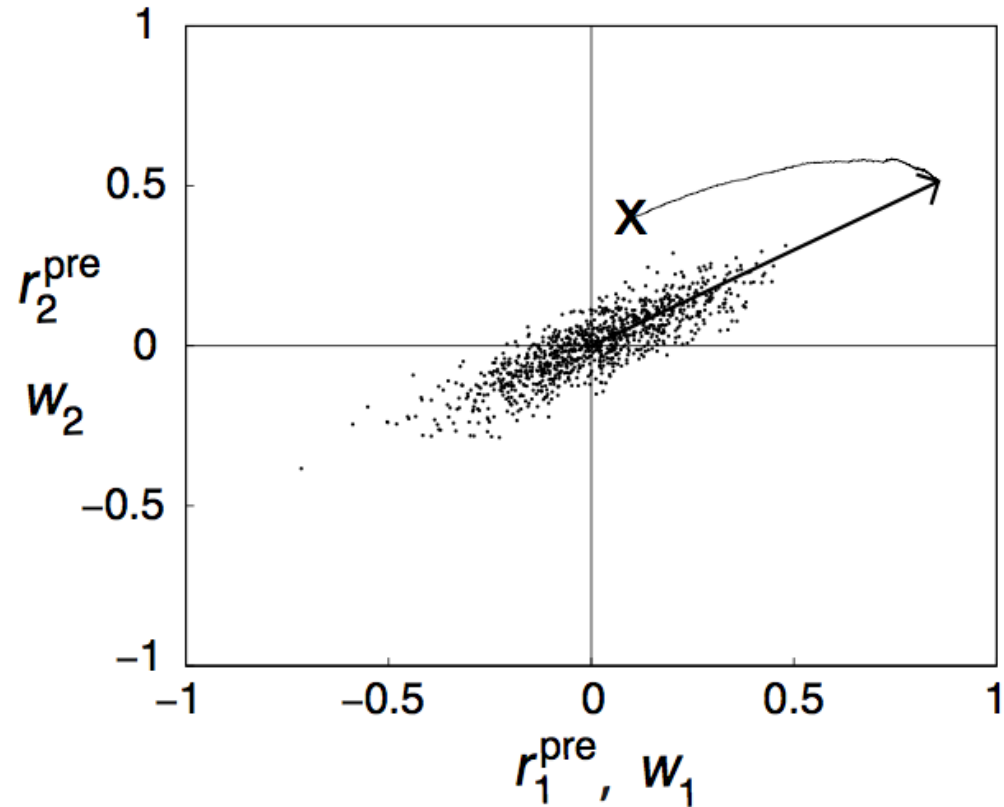


Dimensionality reduction
end
Representational learning

PCA: Principal Component Analysis



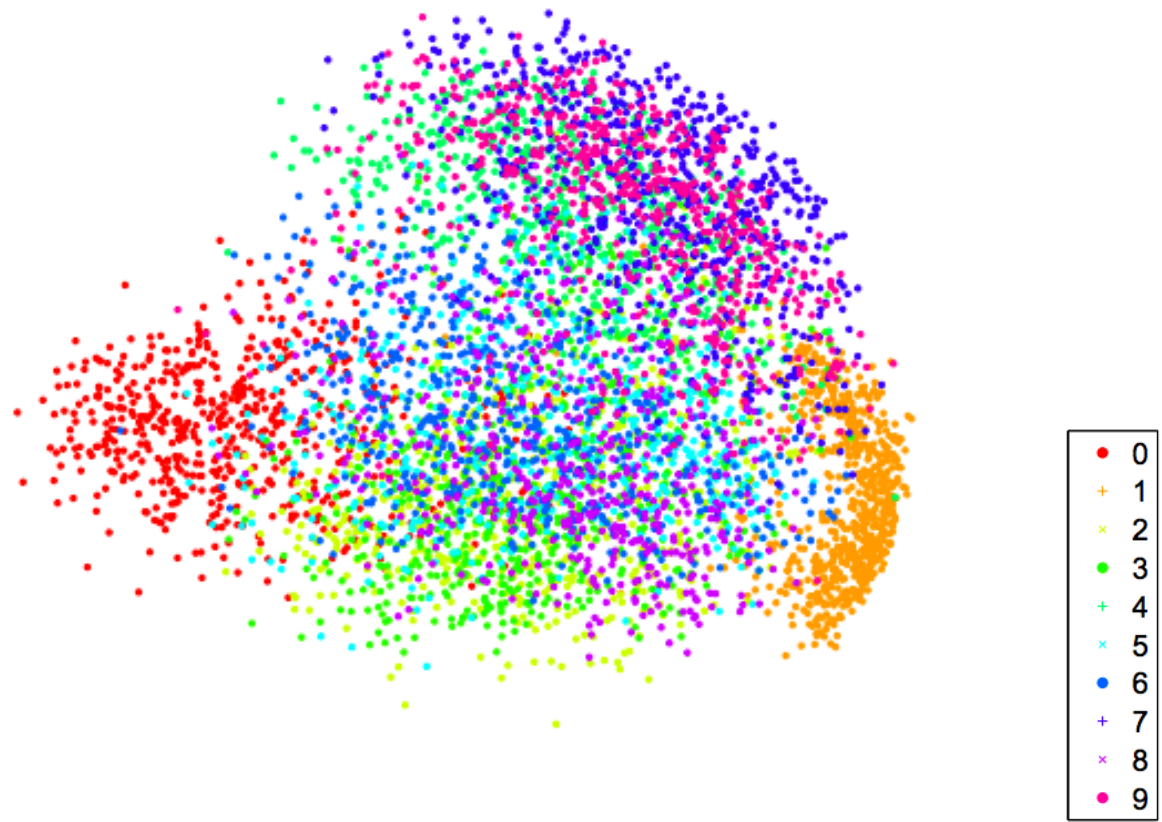
$$\Delta w_{ij} = r_i r_j - (r_i)^2 w_{ij}$$



Linear transformation that maximizes variance

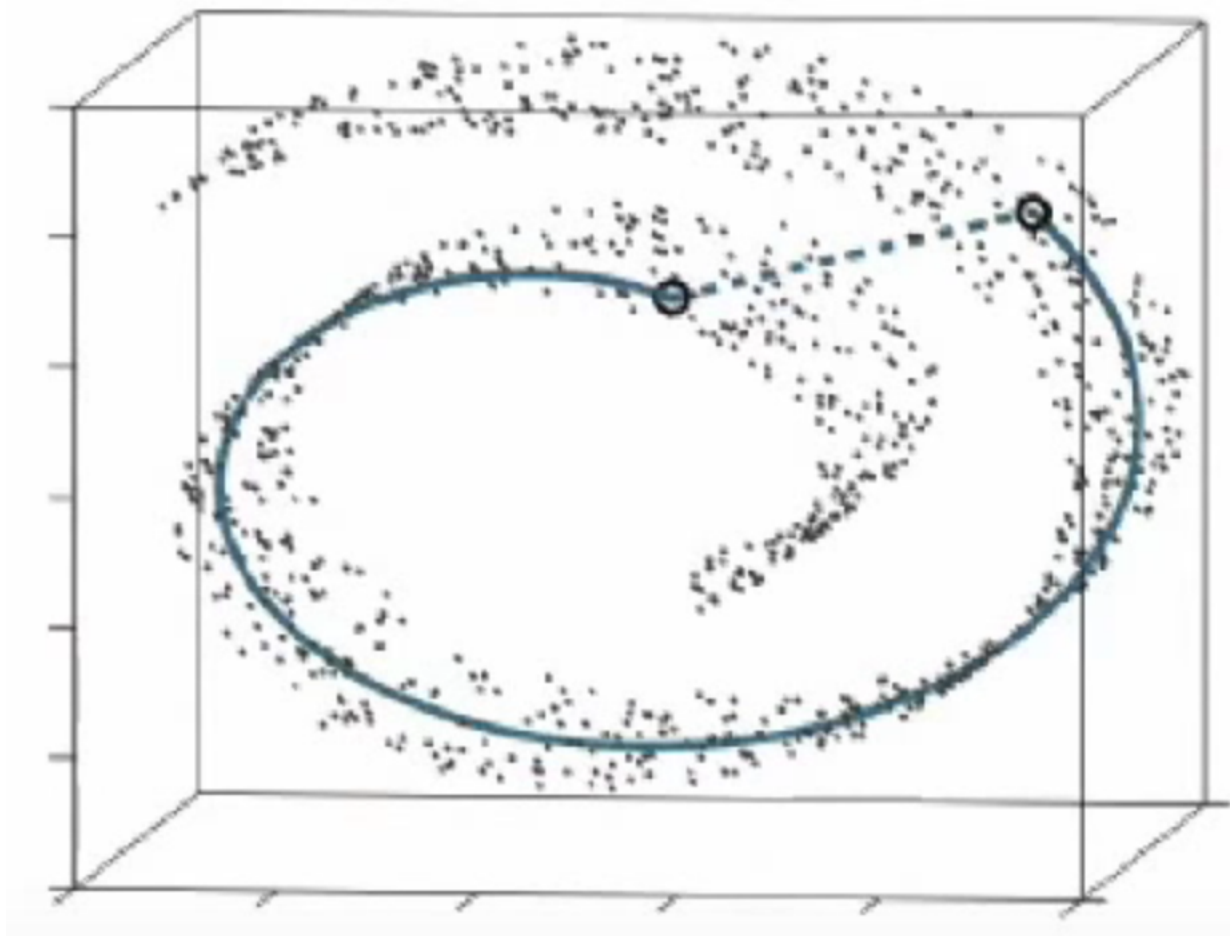
PCA of MNIST data

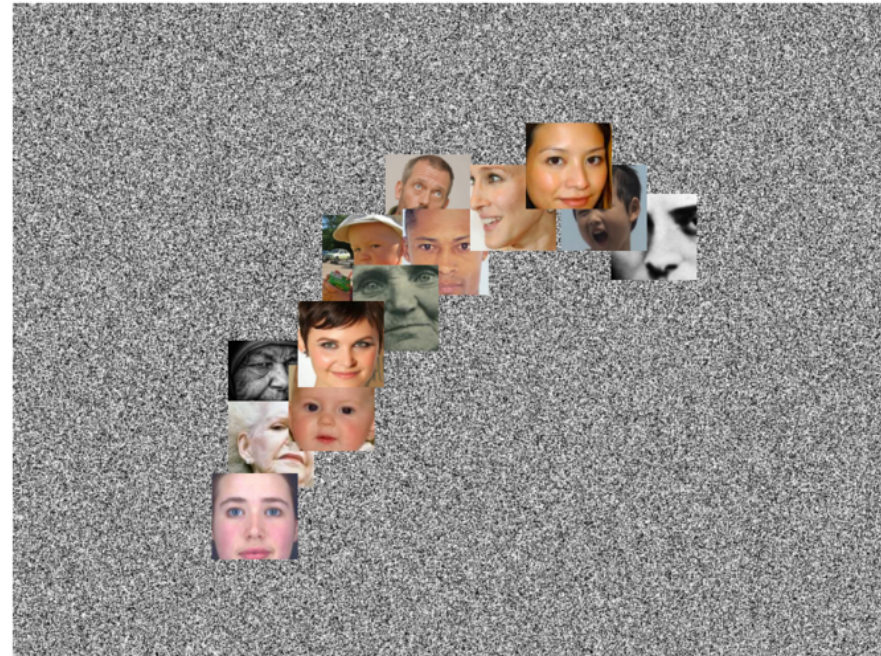
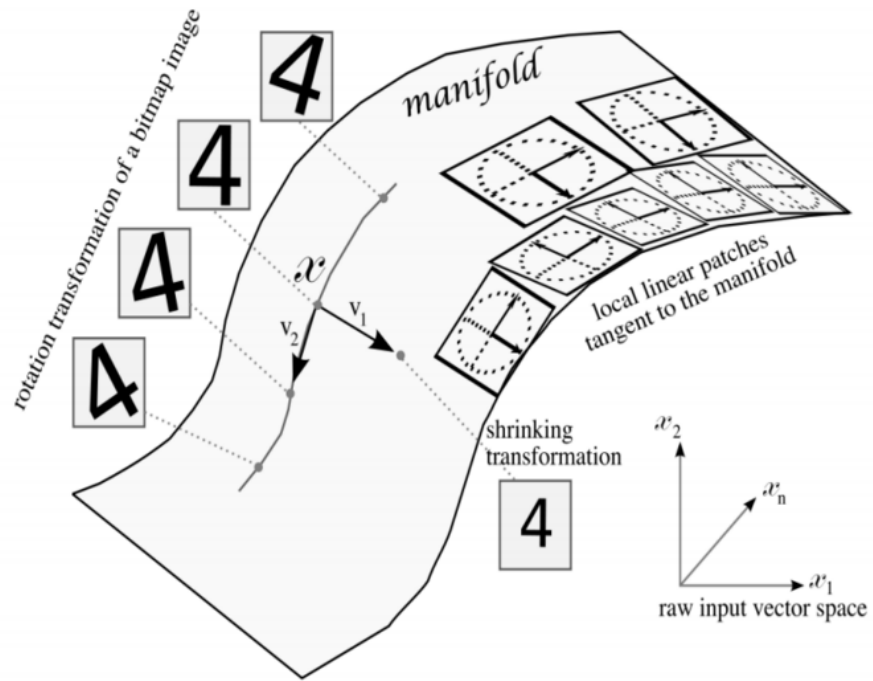
3 6 8 1 7 9 6 6 4 1
6 7 5 7 8 6 3 4 8 5
2 1 7 9 7 1 2 1 4 0
4 8 1 9 0 1 8 3 9 4
7 6 1 8 6 4 1 3 6 0
7 5 9 2 6 5 3 1 9 7
4 2 2 2 2 3 4 4 3 0
0 2 3 8 0 7 3 8 5 7
0 1 4 6 4 6 0 2 4 3
7 1 2 3 1 6 9 3 6 1

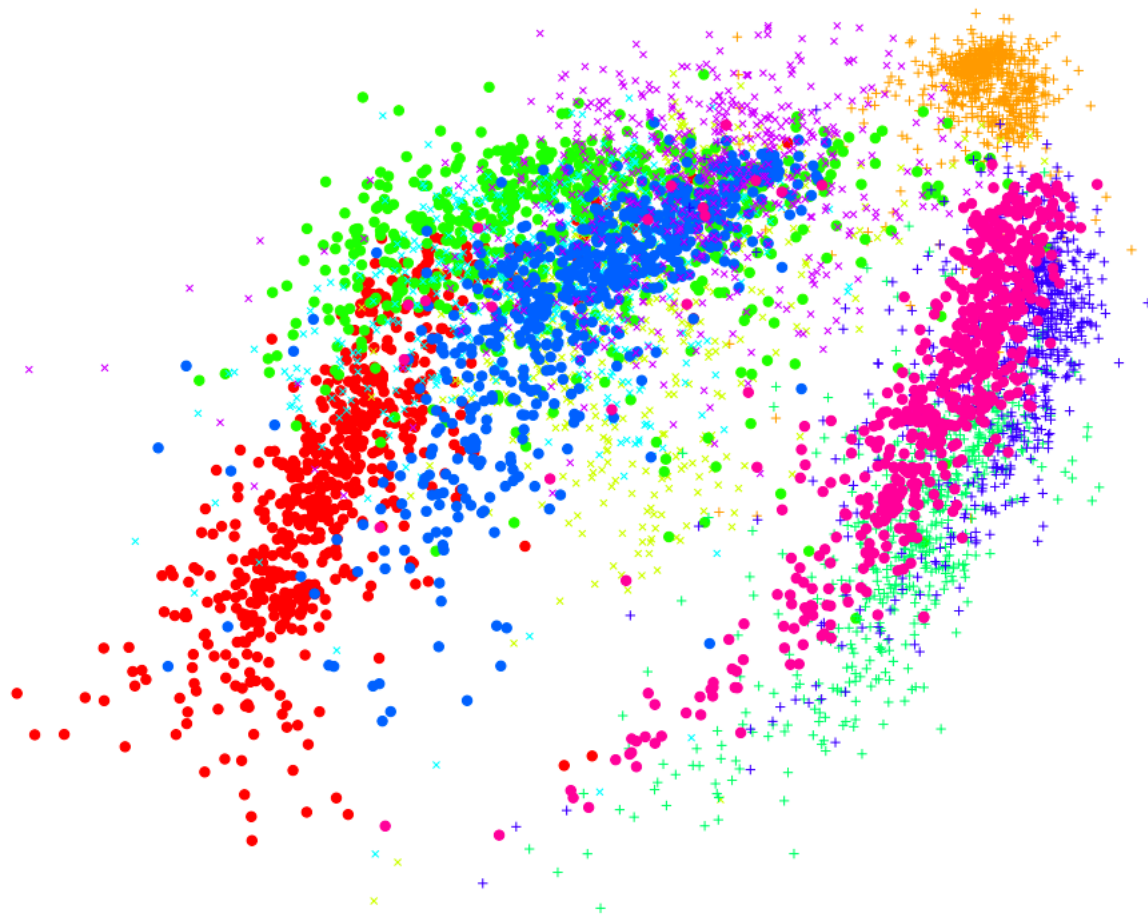


(a) Visualization by PCA.

Preserve large pairwise distances

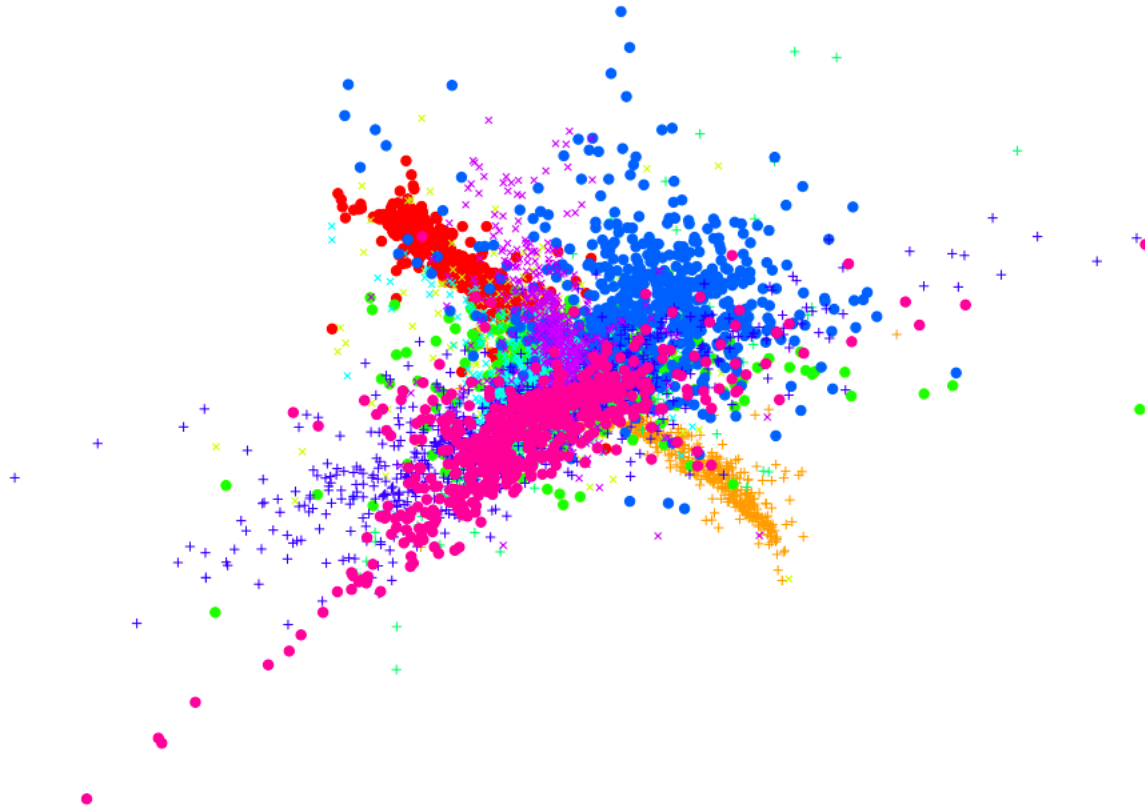
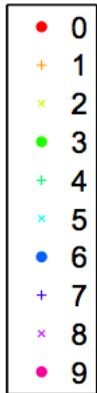






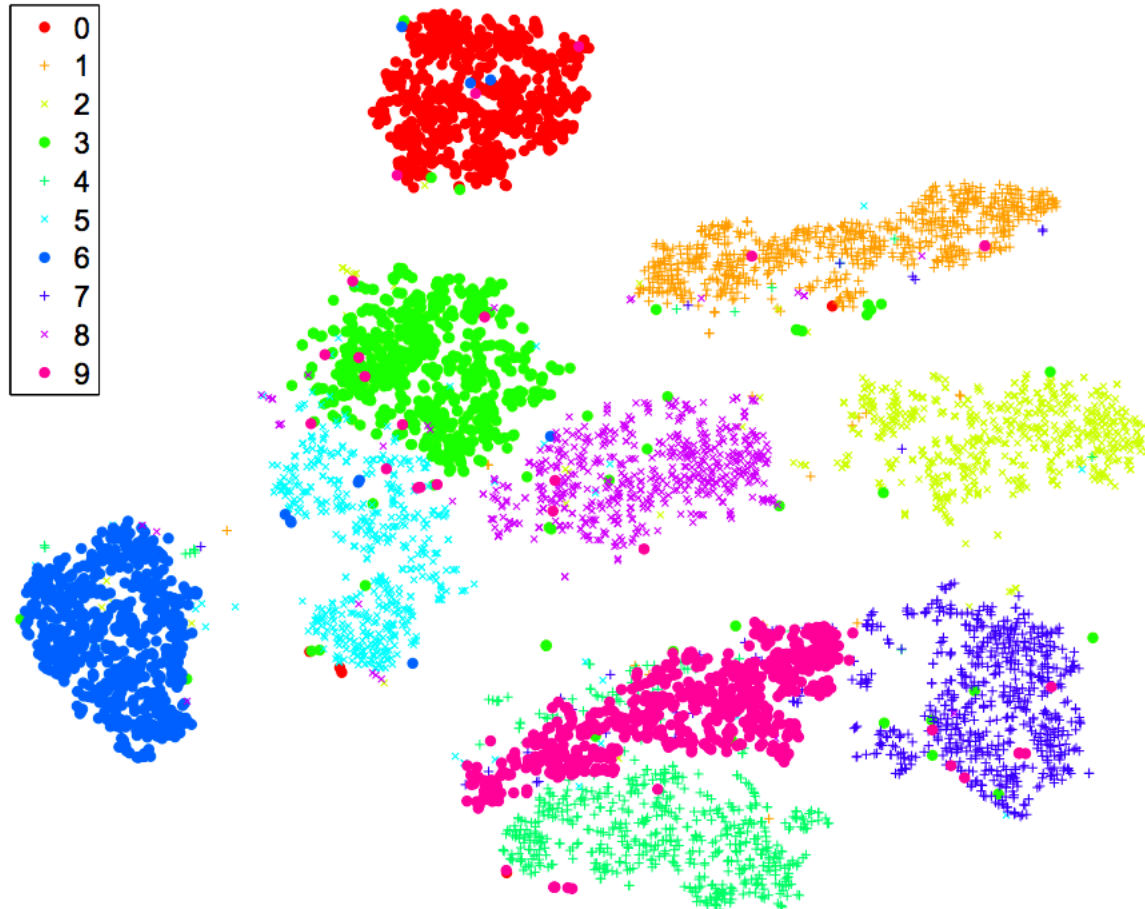
(a) Visualization by Isomap.

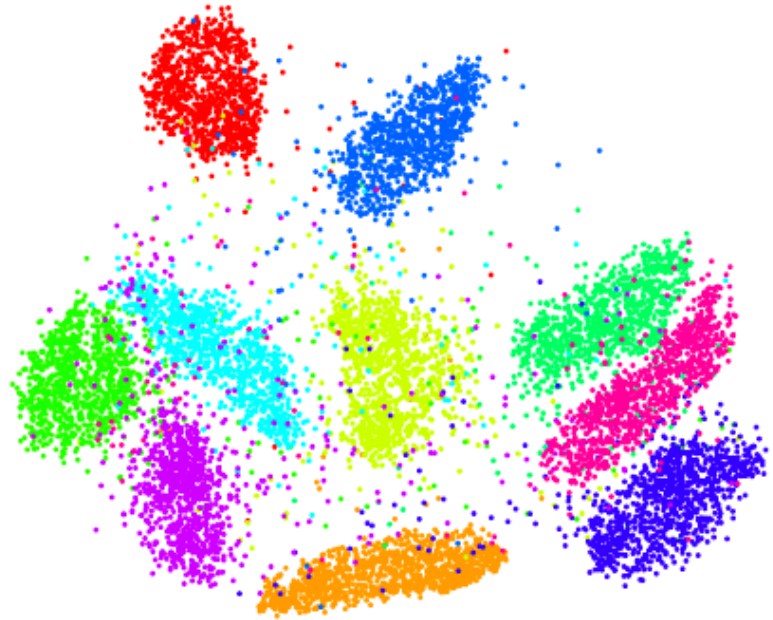
Local Linear Embedding



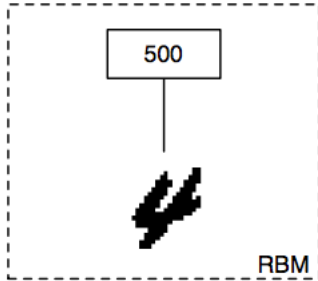
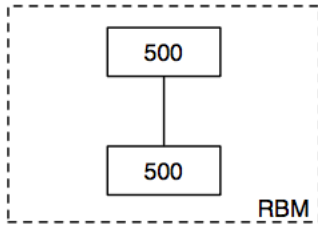
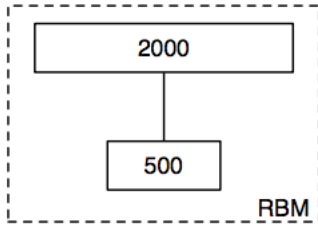
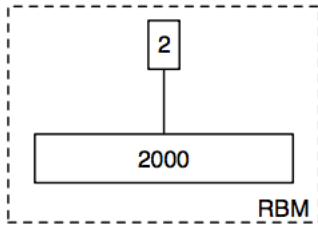
(b) Visualization by LLE.

T_SNE (van der Maaten, Hinton 2008)

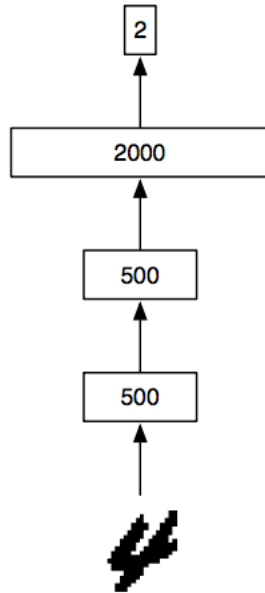




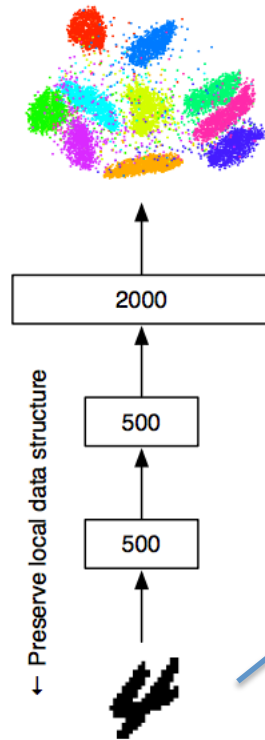
c) Visualization by parametric t-SNE.



Pretraining

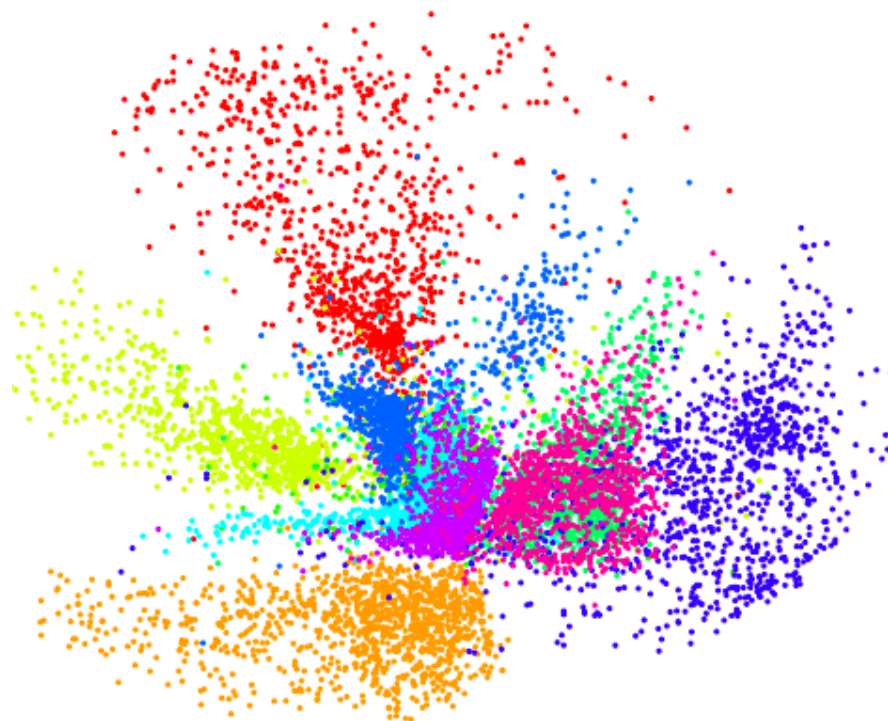


Construction



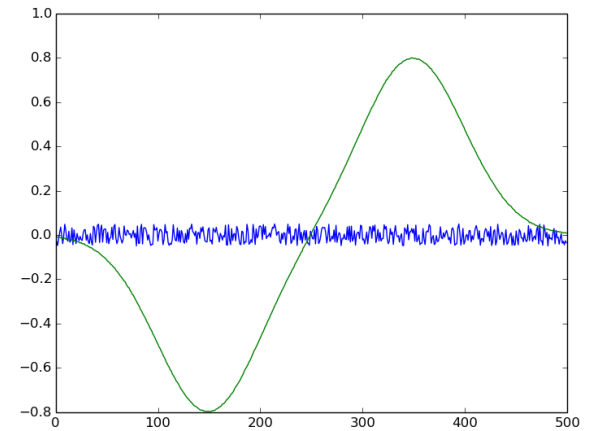
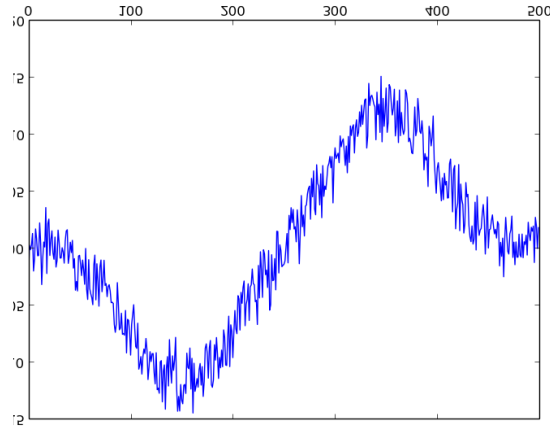
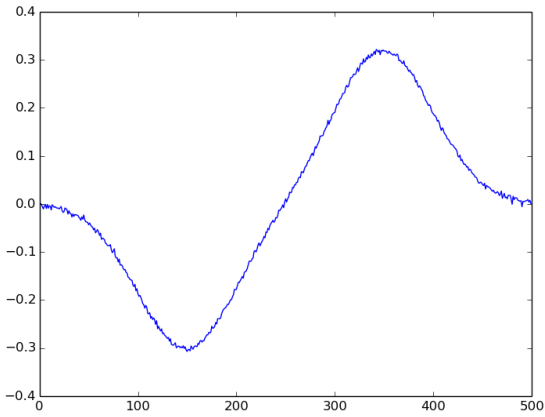
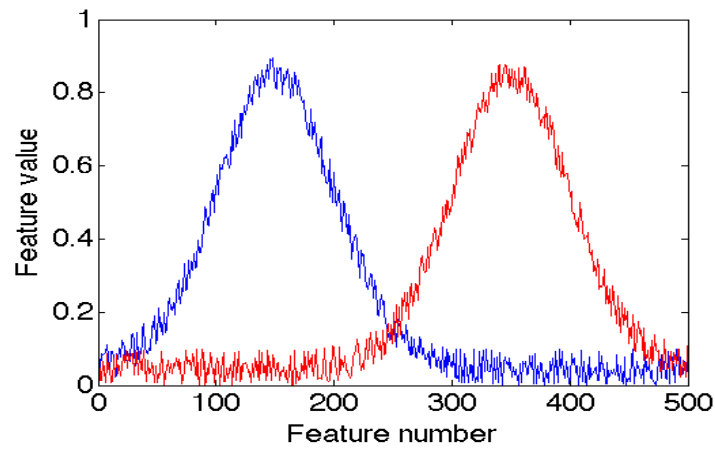
Finetuning

Objective function:
KL divergence between
T-distributed pairwise
distances

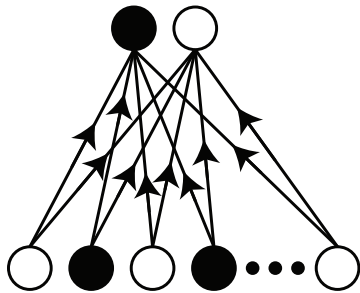


(b) Visualization by an autoencoder.

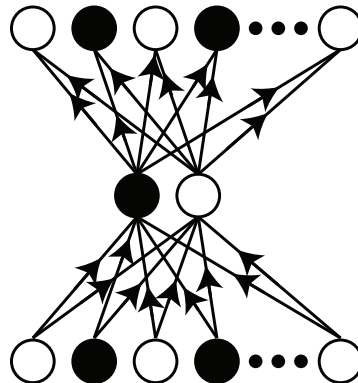
3 simple examples



Perceptron



Autoencoder



RBM

