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Subspace Projection Methods for Large Scale Image Data Analysis

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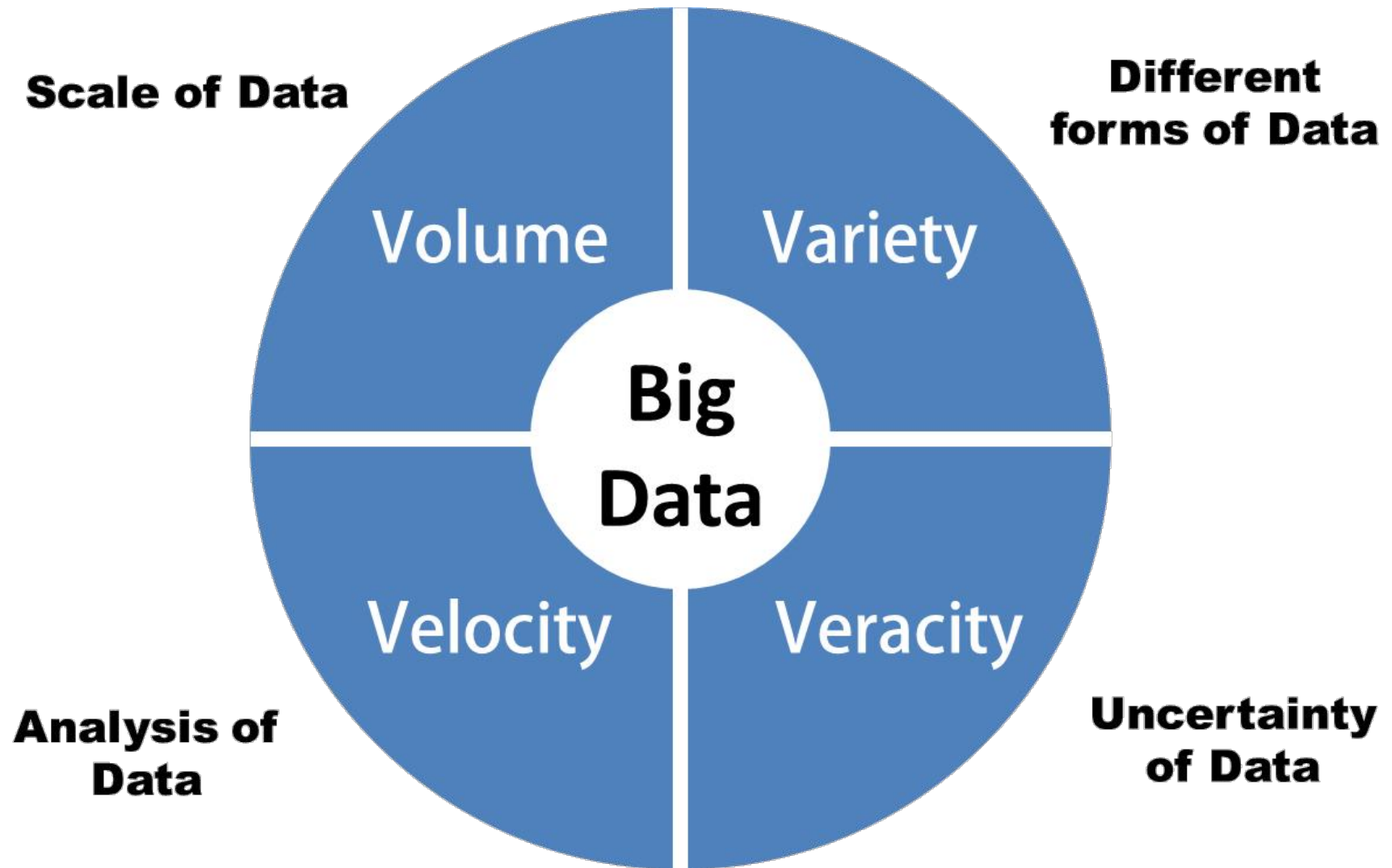


Outline

- Introduction
- Big Data & Sub-space analysis
- Visual data
- Semantic analysis & attributes
- Intrinsic dimensionality estimation
- Information theoretic measure of projection
- Sub-space projection methods
- Analysis and Conclusions

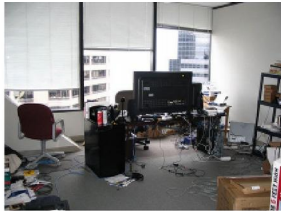


Big Data





Visual Data



bottle

car

chair

dog

plant

train

PASCAL Visual Object Classification
challenge dataset



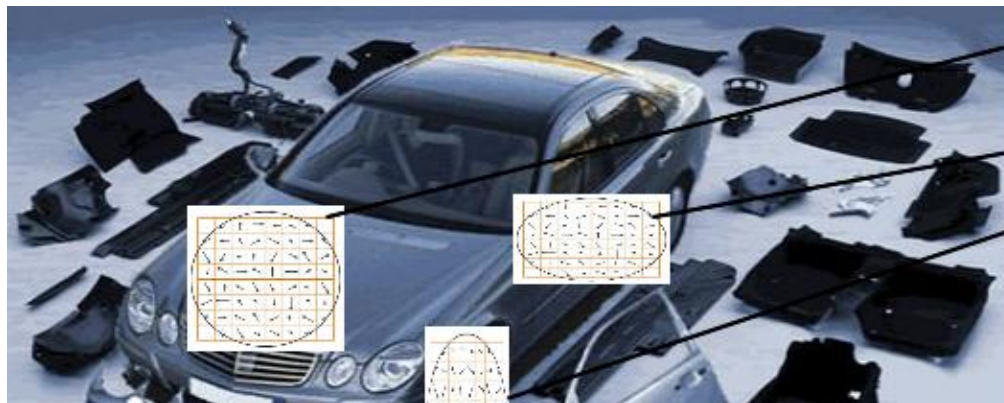
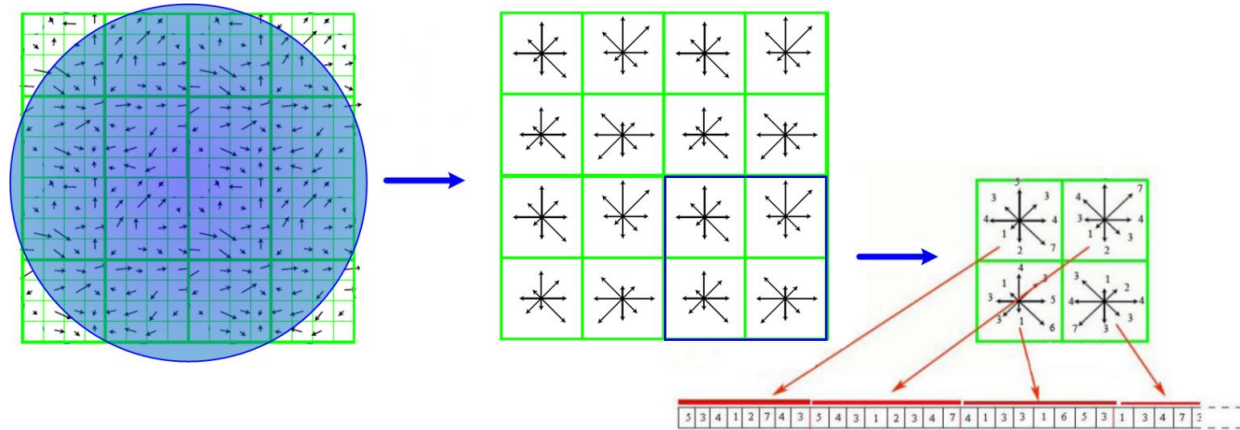


Intra-category appearance variation





Visual Feature & Category parts



7	45	72	0	0	0	0	0	0	0
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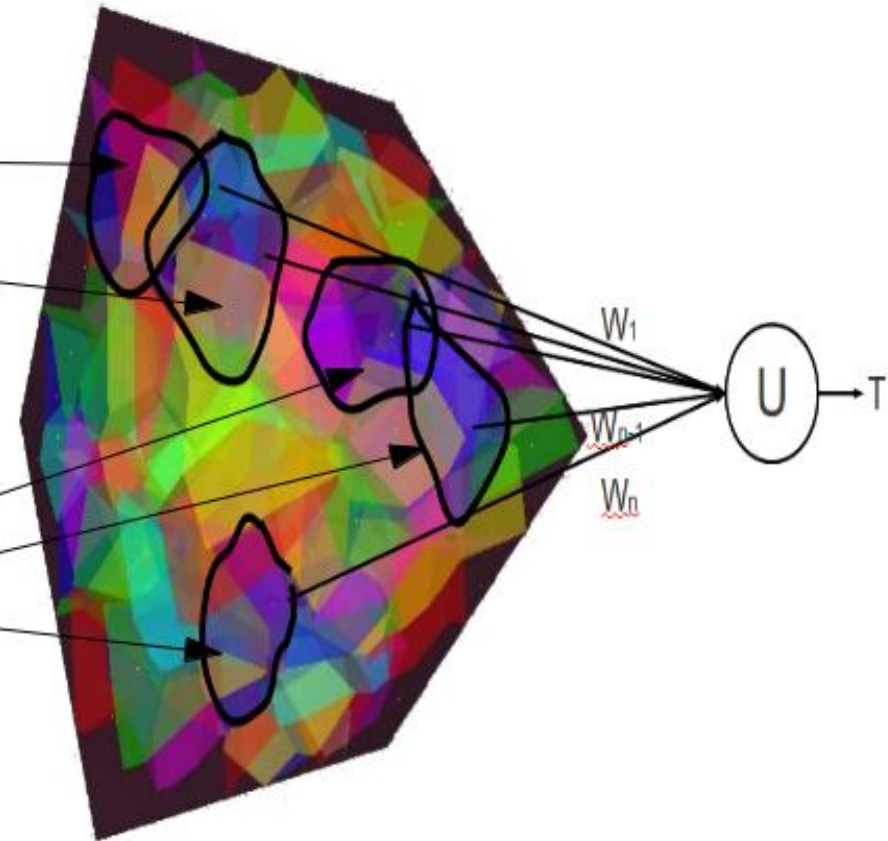
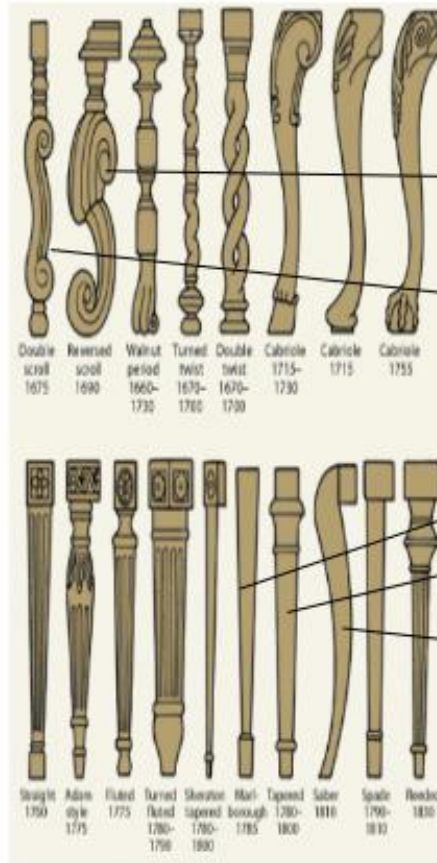
0	0	0	32	3	93	0	0	0	0
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0	0	0	0	0	0	6	42	15	11
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Feature Vectors

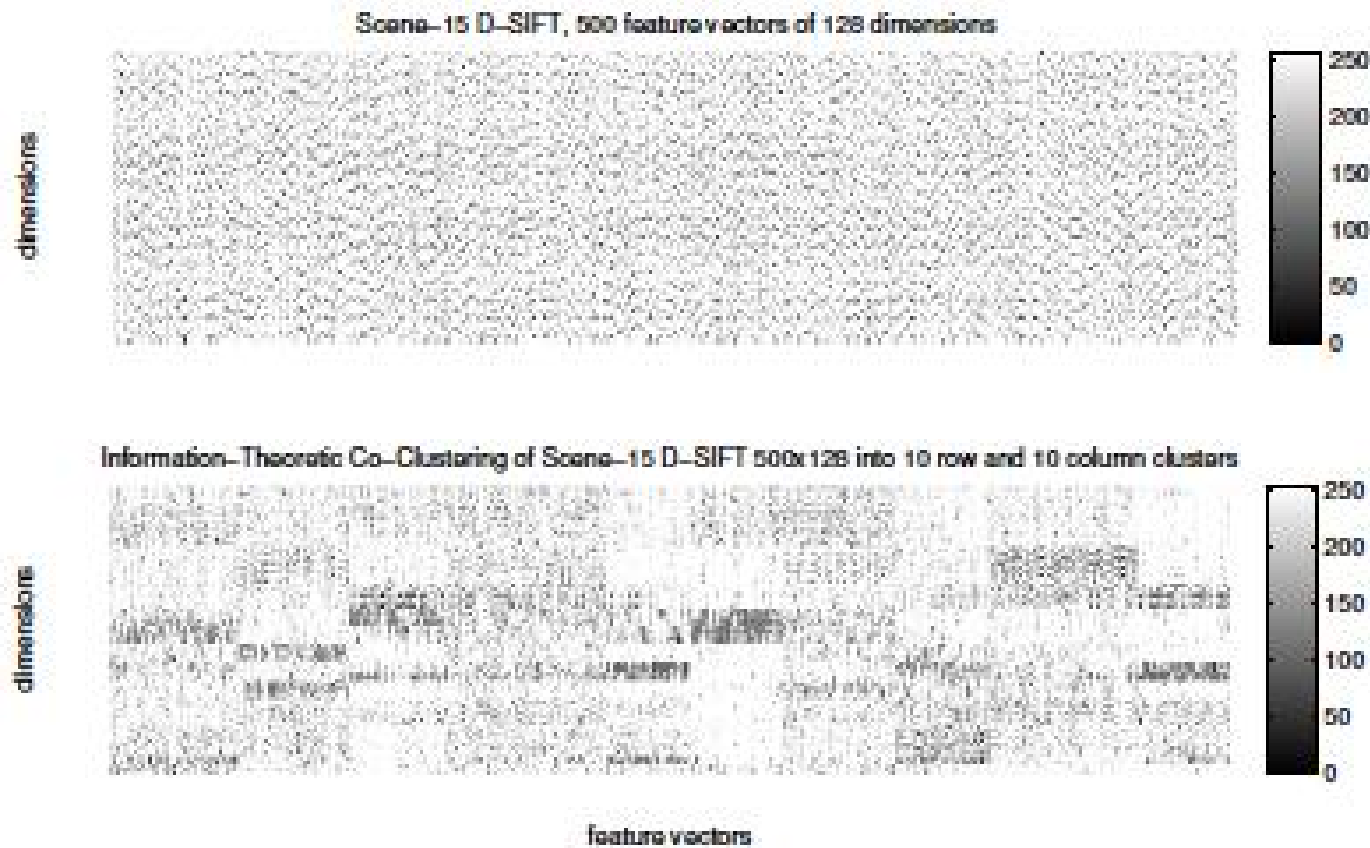


Project semantically equivalent parts



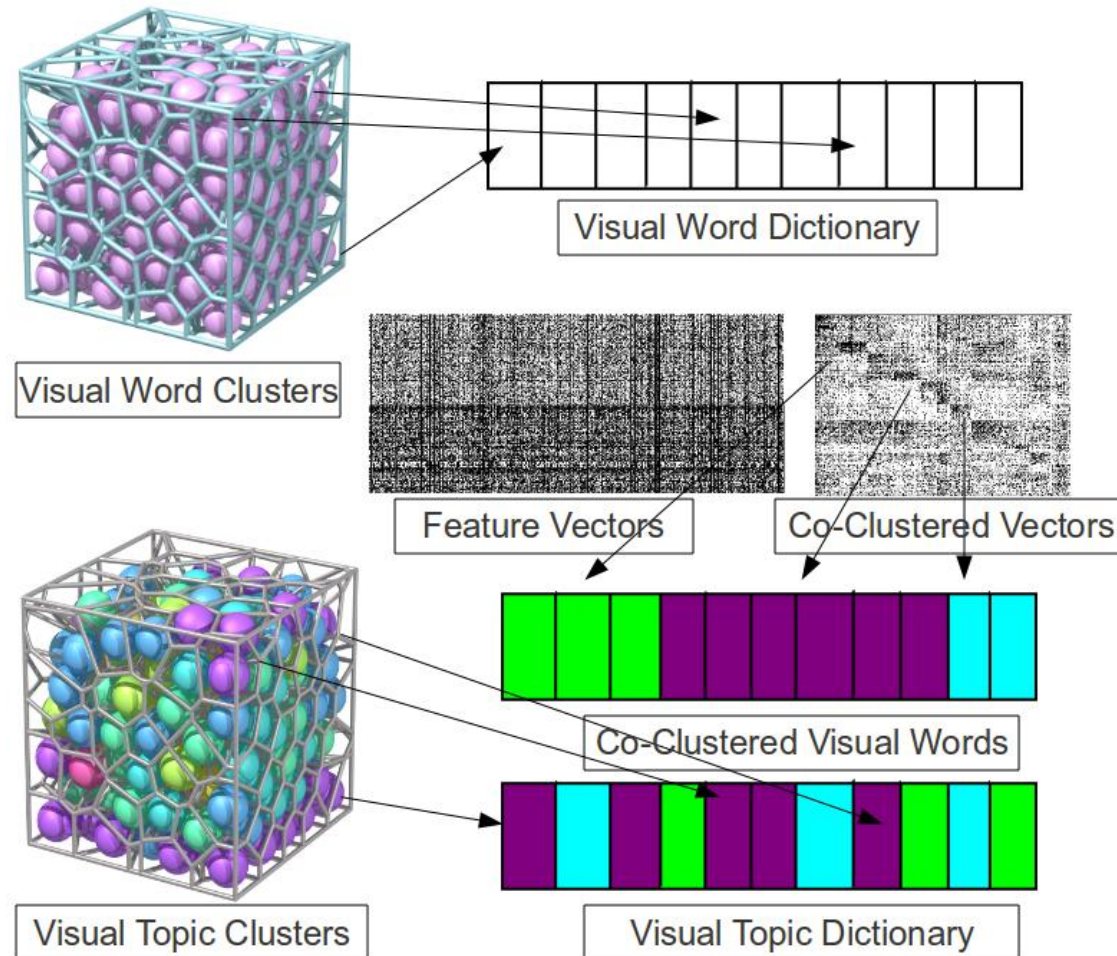


Co-clustering feature descriptors



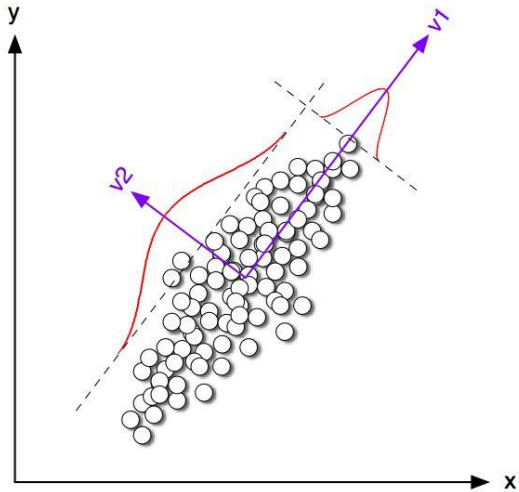


Group semantically equivalent feature space regions

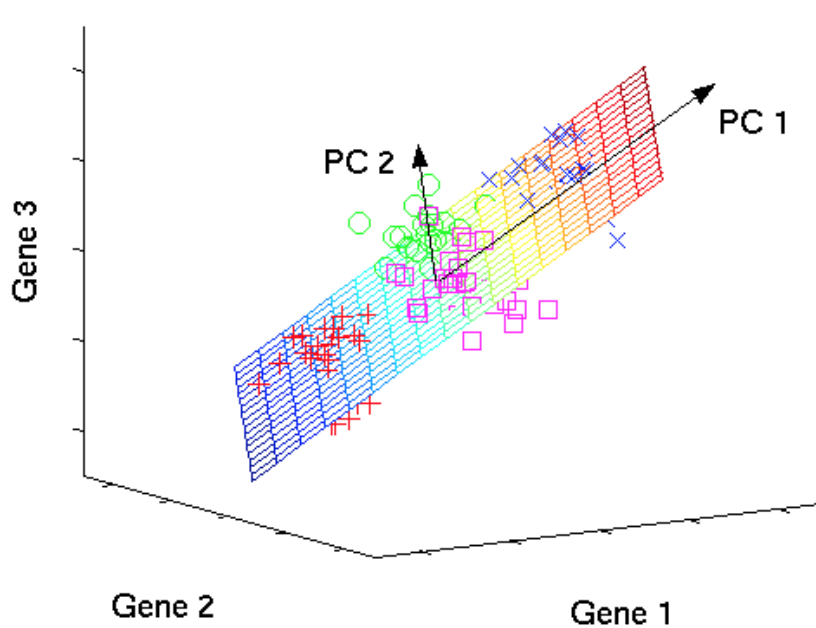




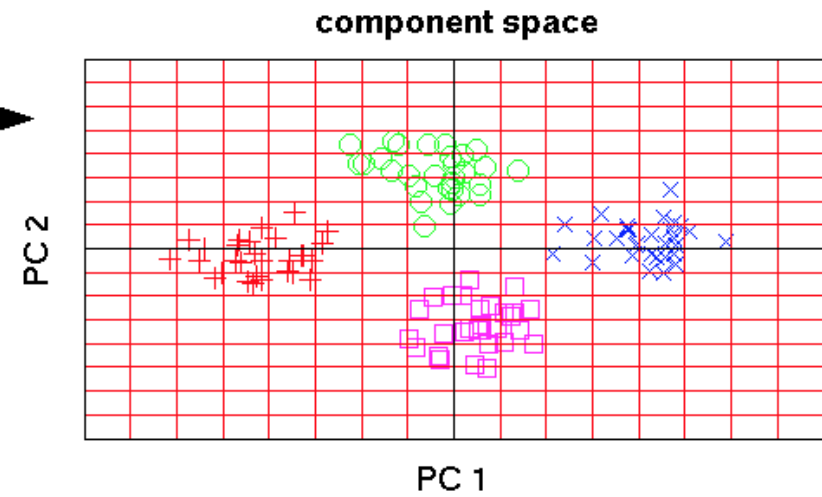
Principal Component Analysis



original data space

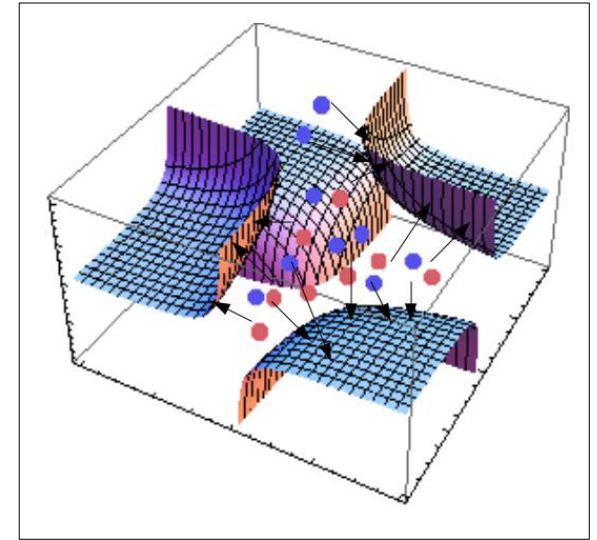
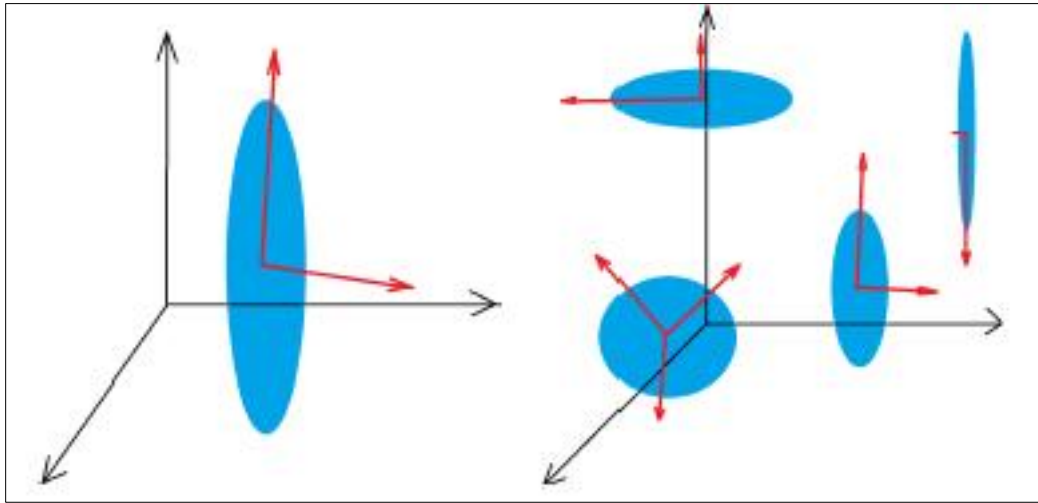


PCA



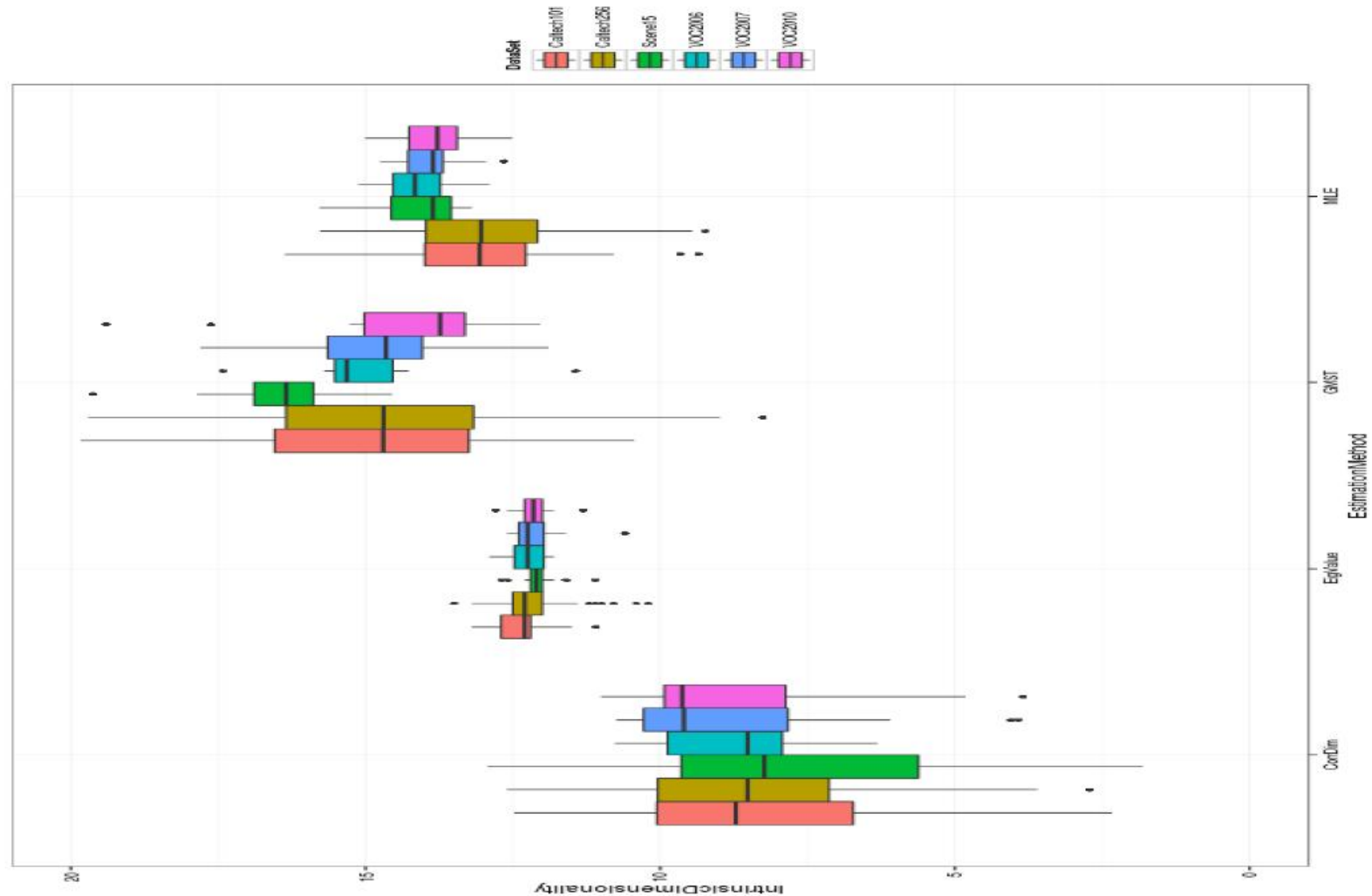


Project onto multiple subspaces





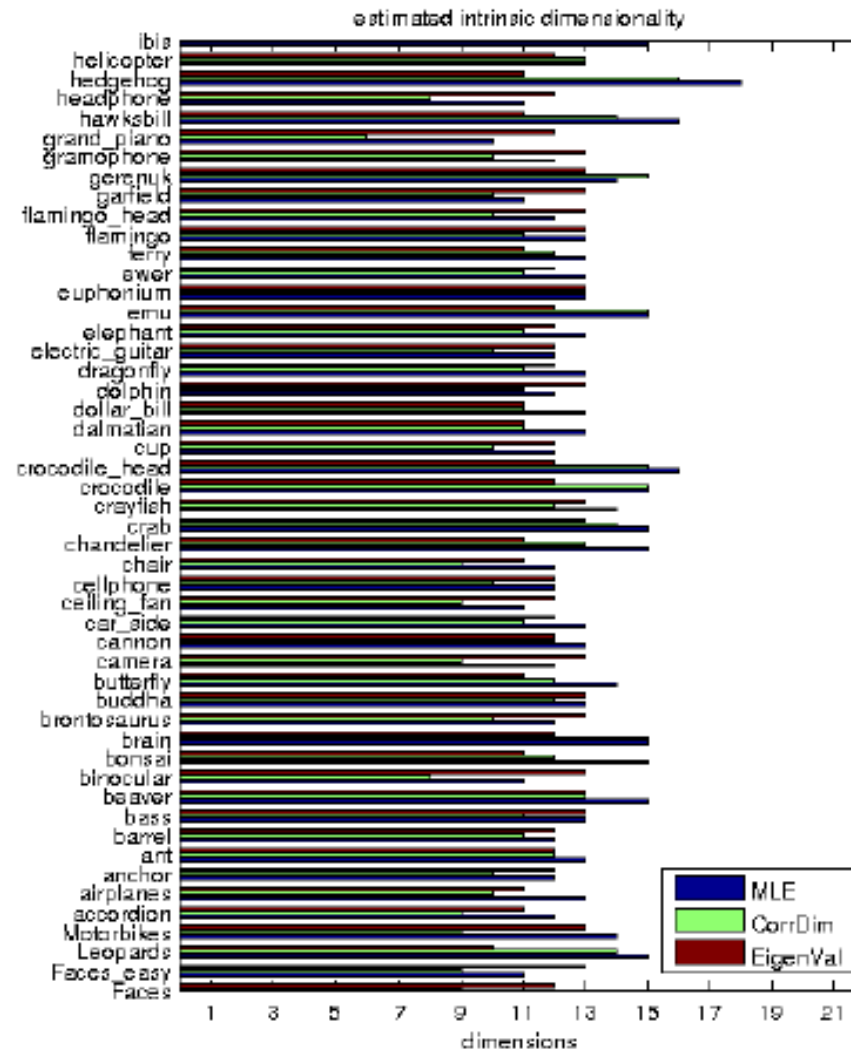
Intrinsic Dimensionality



Intrinsic dimensionality was much lower than typically assumed in literature

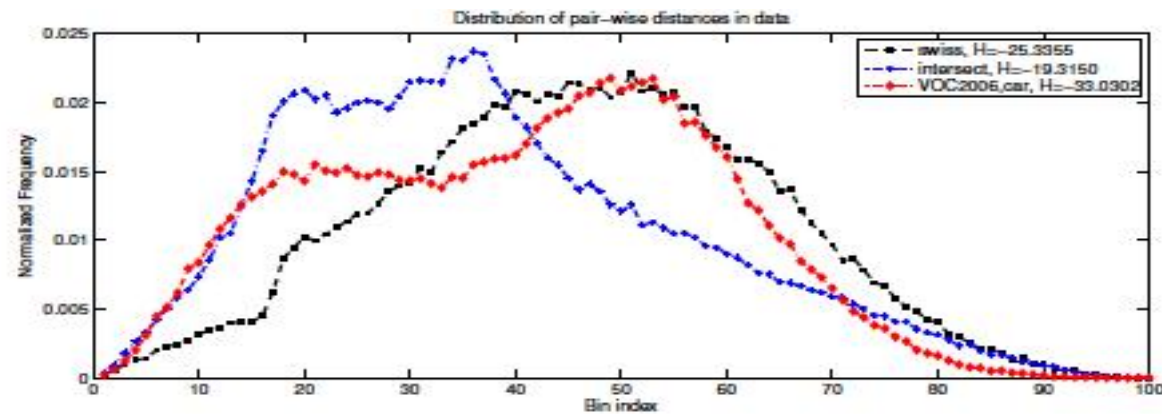
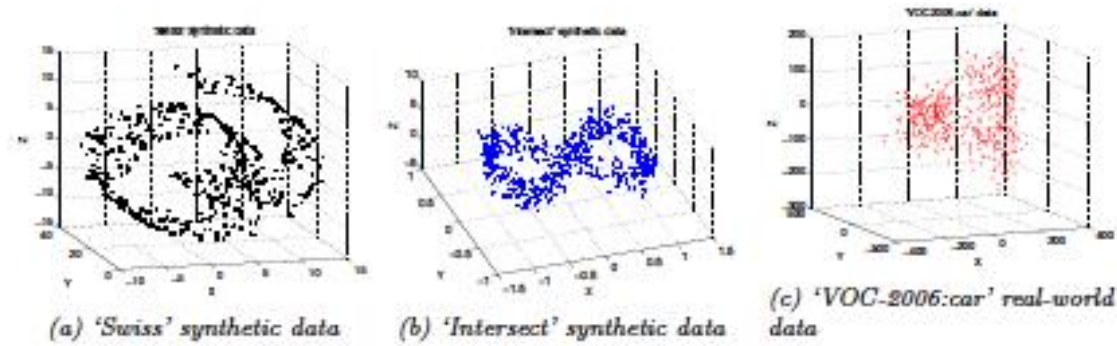


Intrinsic dimensionality of different visual categories





Measure information in embedded space





Sub-space projection methods

Global Methods

- Principal Components
- Multi-Dimensional Scaling
- Stochastic Proximity Embedding
- Isomap
- Diffusion Maps

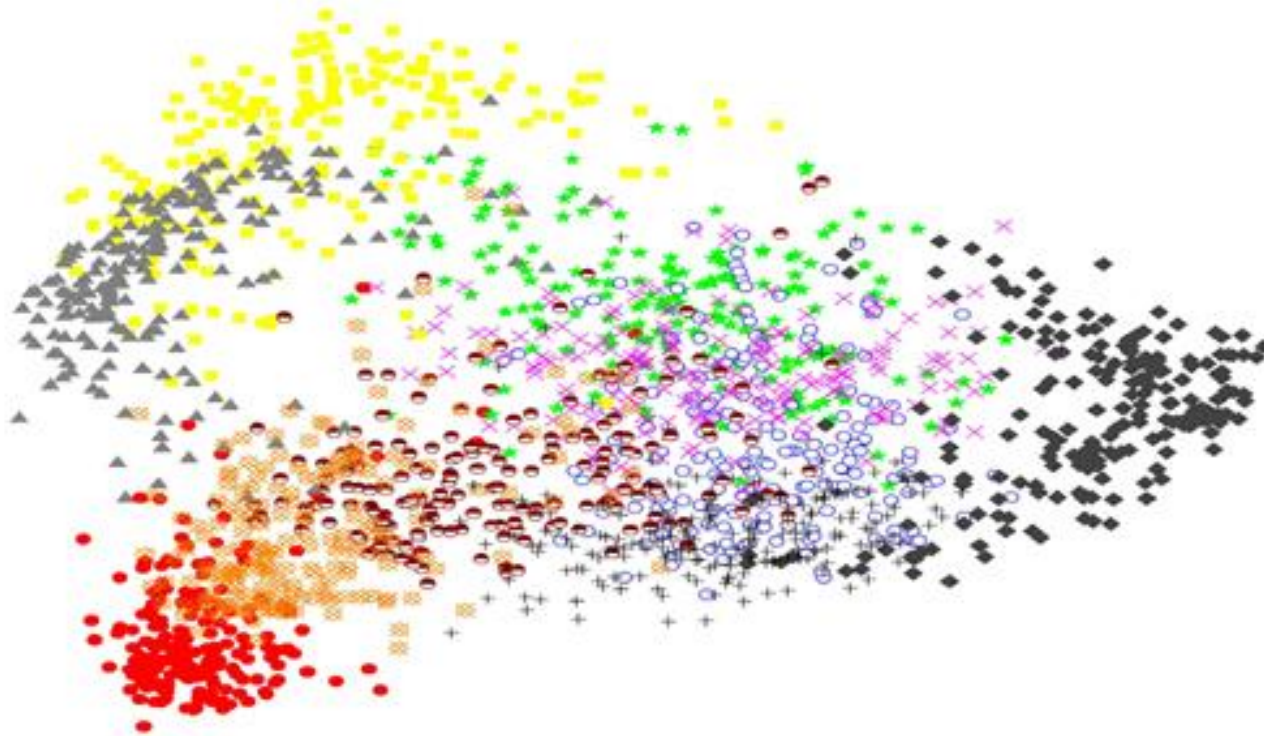
Local Methods

- Locally Linear Embedding
- Locality Preserving Projection
- Neighbourhood Preserving Projection
- Landmark Isomap
- t-Stochastic Neighbourhood Embedding



Visual Comparison 1 of 3

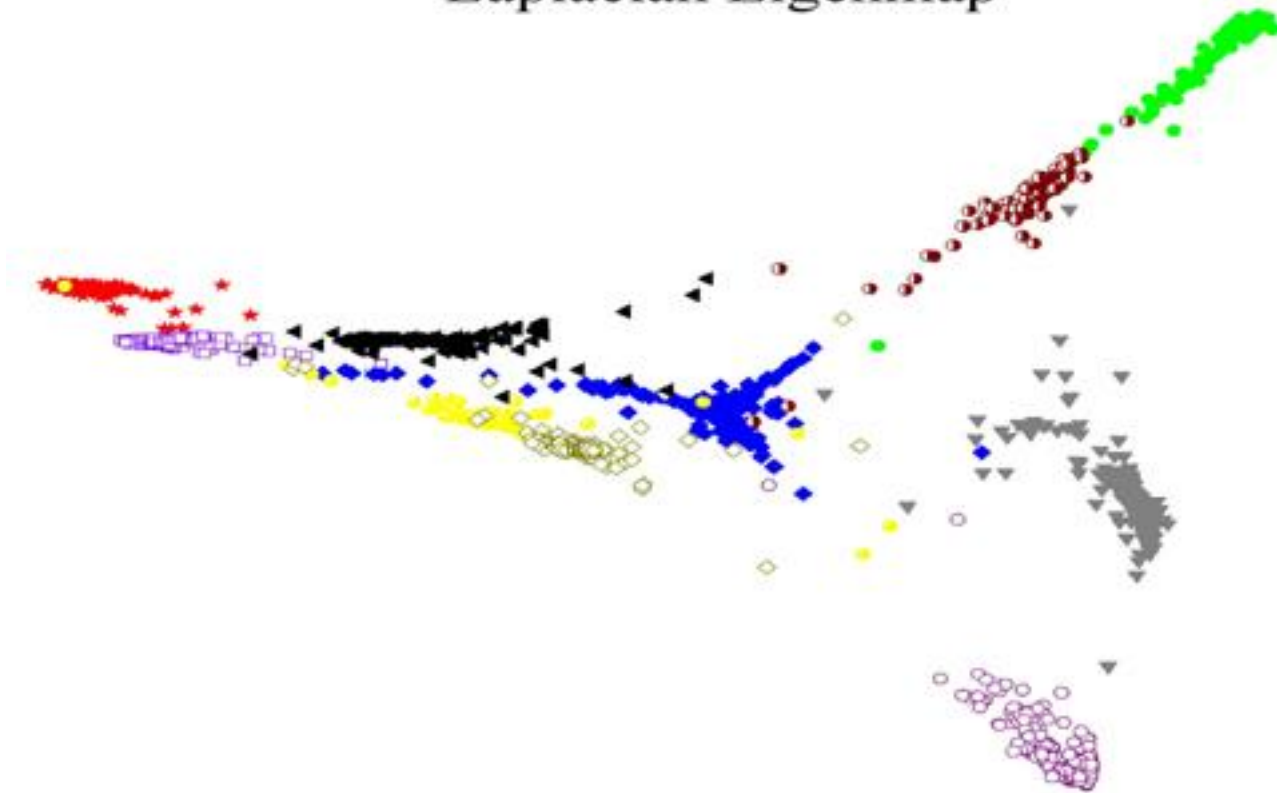
PCA





Visual Comparison 2 of 3

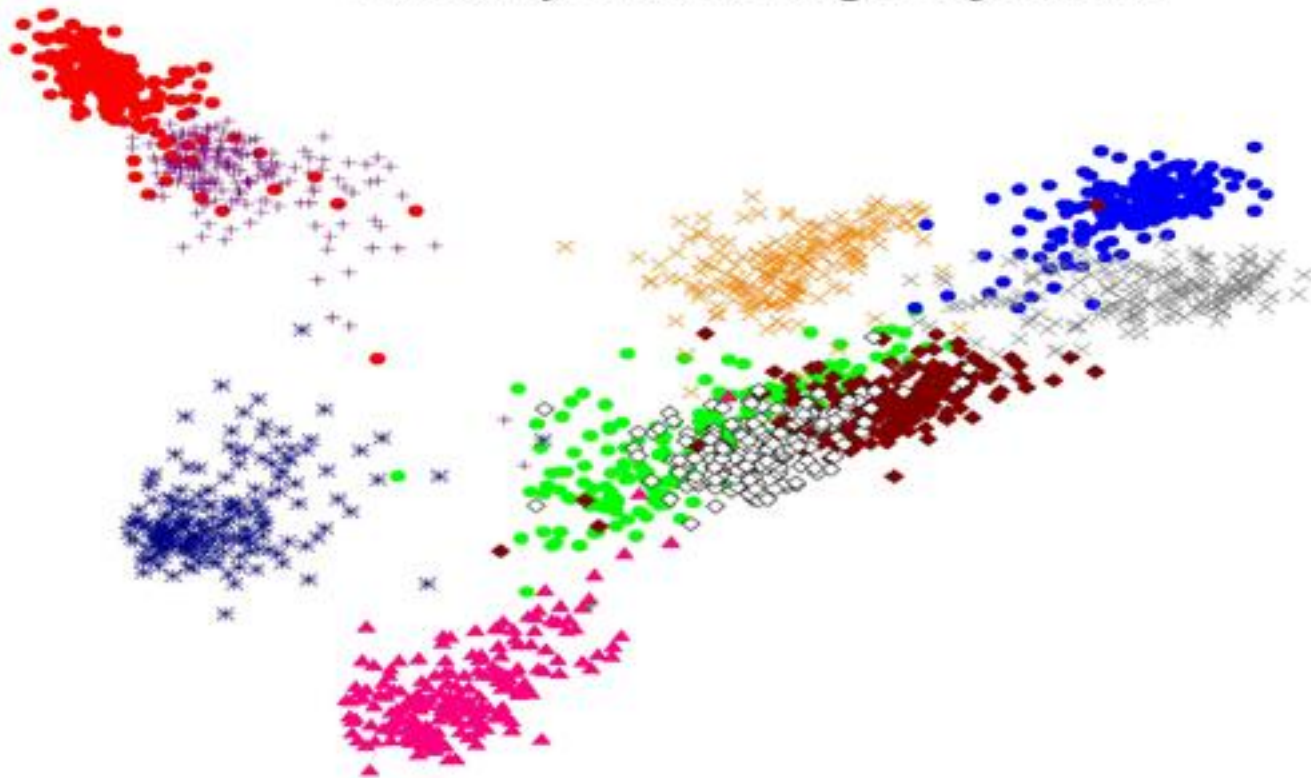
Laplacian Eigenmap





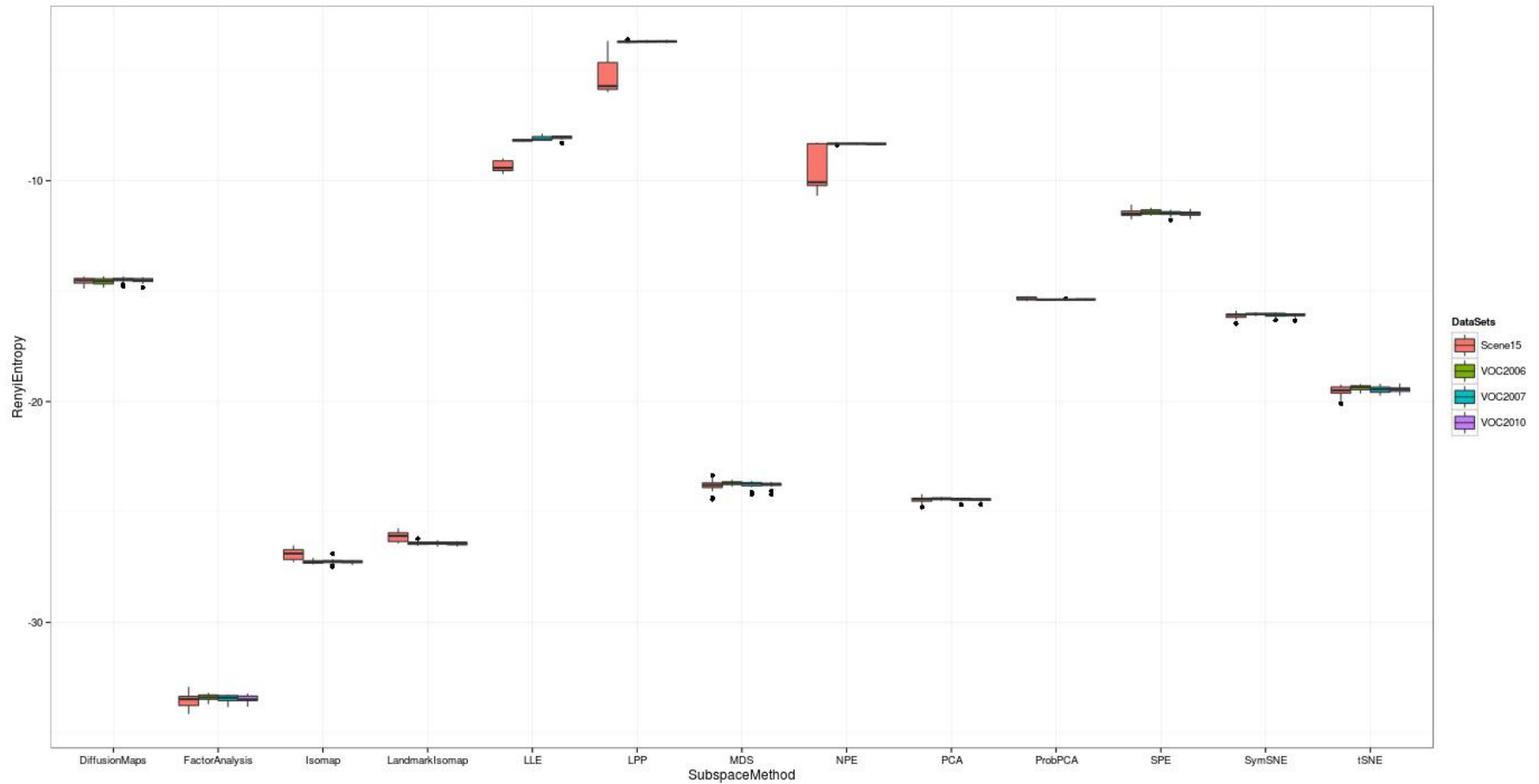
Visual Comparison 3 of 3

Locality Preserving Projection



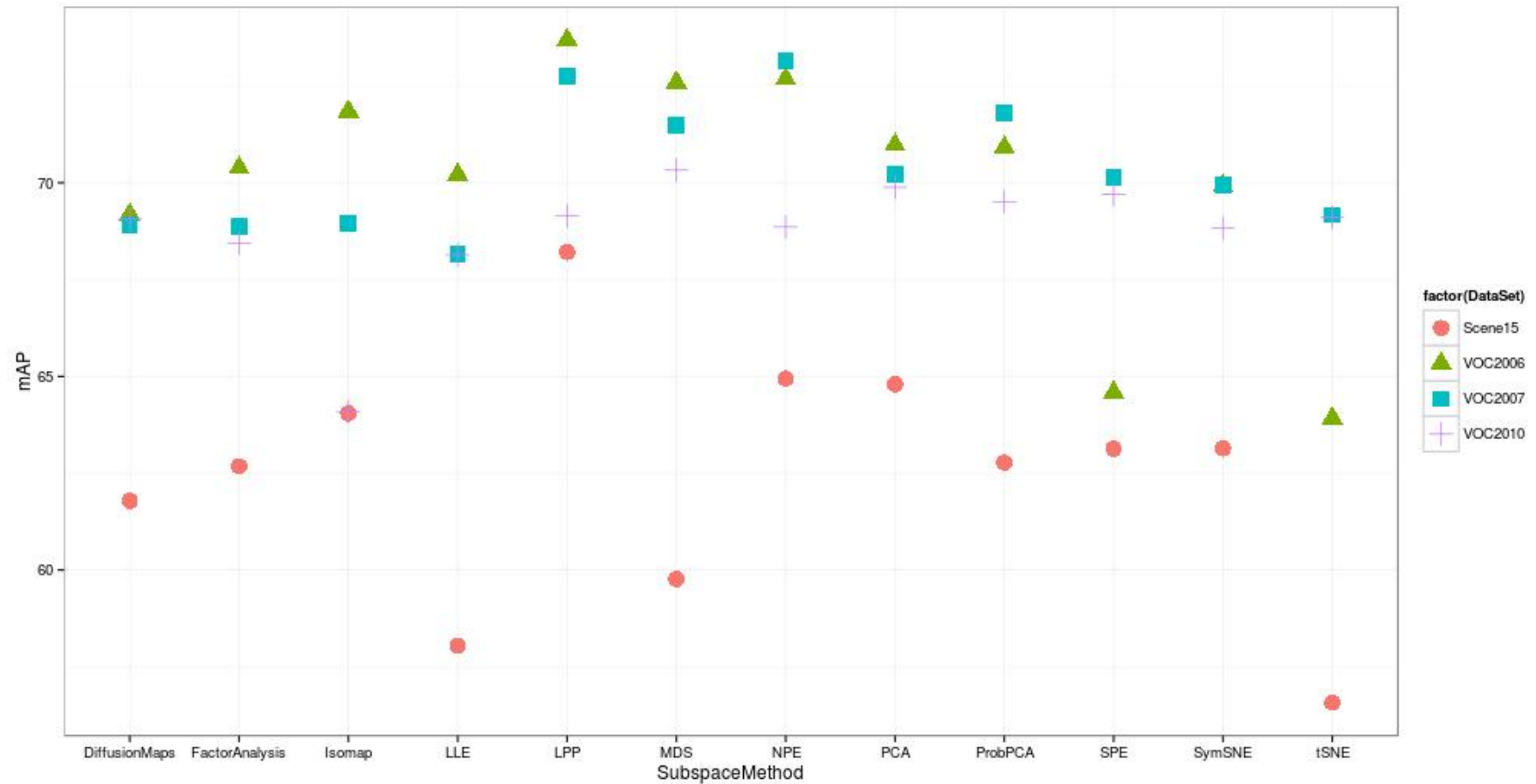


Renyi entropy based comparison



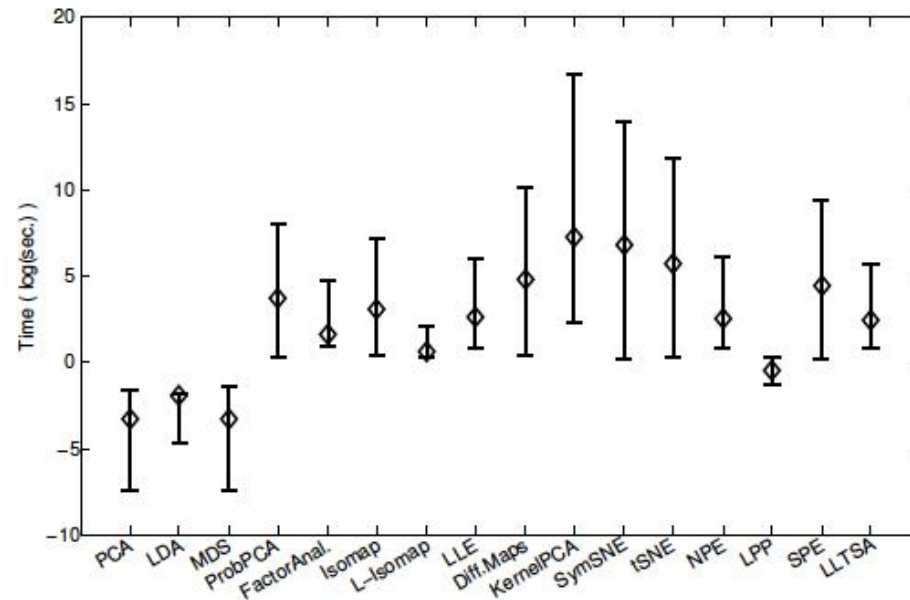
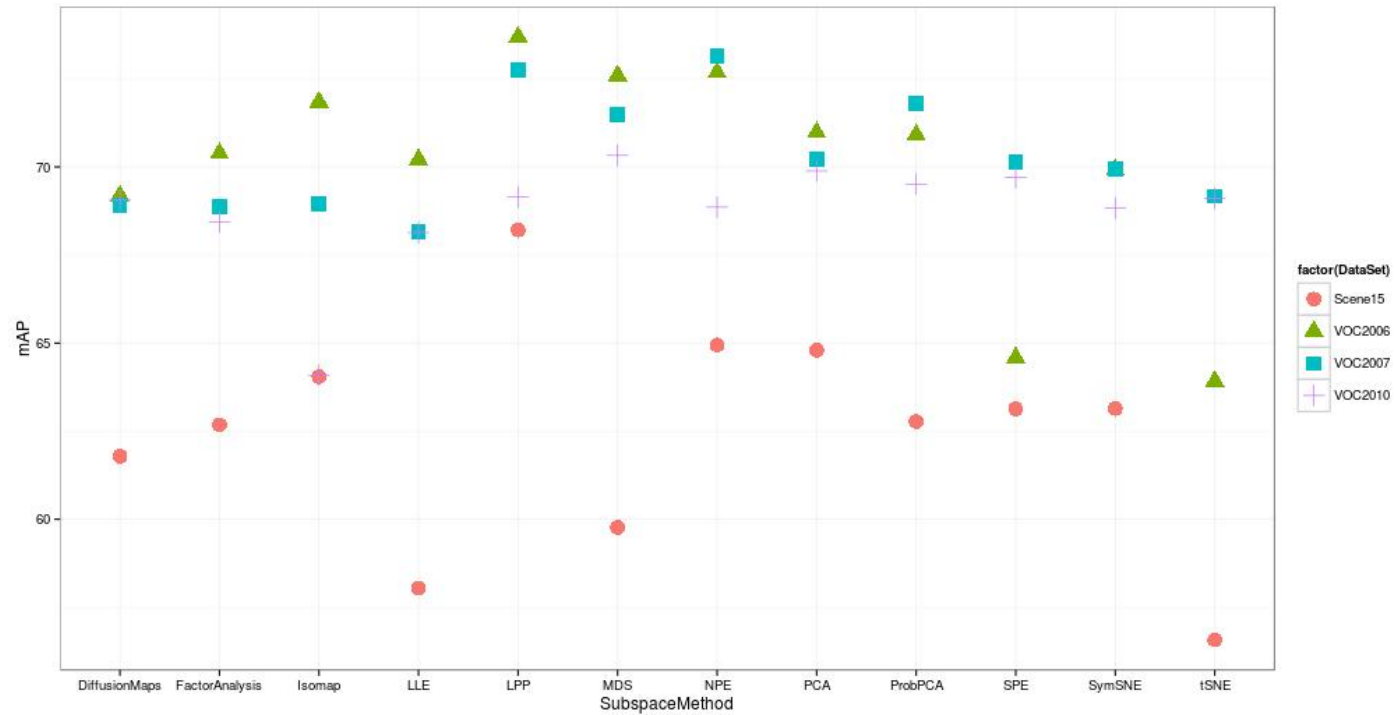


Classification performance based comparison





Processing time based cor





Summary

- Estimated intrinsic dimensionality was in the neighborhood of 14 of the 128-dimensional descriptor.
- The performance of Locality Preserving Projection in comparison to other embedding methods accentuates the importance of modelling structure in local distributions.



Future Work

- Extend work on images to multi-media data
- Implement sub-space projection on Map-Reduce