

# Machine Learning applied to Planetary Sciences

PTYS 595B/495B

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<https://leonpalafox.github.io/MLClass/>

# Manifold Learning

- As we have more dimensions, visualization of the datasets become difficult.
- Manifold learning techniques are used mostly for analyzing the data.
- Nevertheless, these are still important to find interesting relationships in the data.

# T-SNE

t-distributed stochastic neighbor embedding

<http://everynoise.com/engenremap.html>

<http://cs.stanford.edu/people/karpathy/tsnejs/csvdemo.html>

# T-SNE theory

- Given two spaces  $X$  and  $Y$ , where:
  - Dimensionality of  $X \gg$  Dimensionality of  $Y$  (2-3)
- T-SNE calculates probability distributions over  $X$  and  $Y$ , in which similar objects are near.

$$p_{j|i} = \frac{\exp(-\|\mathbf{x}_i - \mathbf{x}_j\|^2 / 2\sigma_i^2)}{\sum_{k \neq i} \exp(-\|\mathbf{x}_i - \mathbf{x}_k\|^2 / 2\sigma_i^2)}, \quad q_{ij} = \frac{(1 + \|\mathbf{y}_i - \mathbf{y}_j\|^2)^{-1}}{\sum_{k \neq i} (1 + \|\mathbf{y}_k - \mathbf{y}_i\|^2)^{-1}}$$

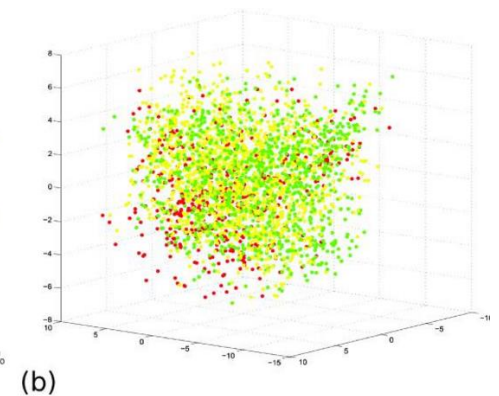
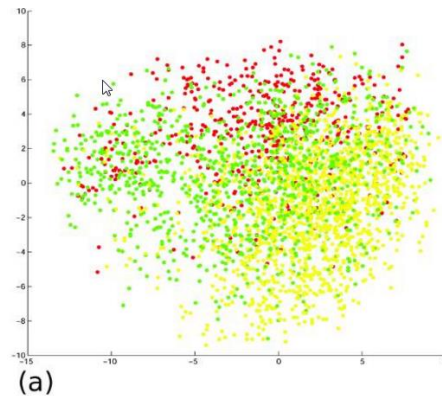
$$p_{ij} = \frac{p_{j|i} + p_{i|j}}{2N}$$

# Applications

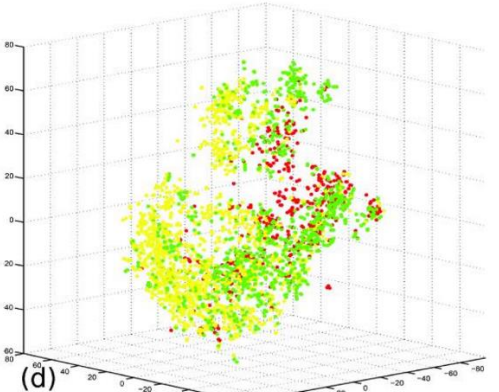
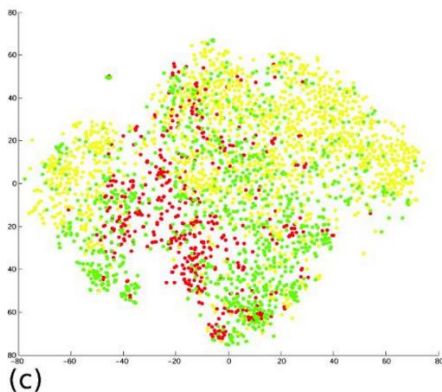
- Cancer Prognosis



PCA

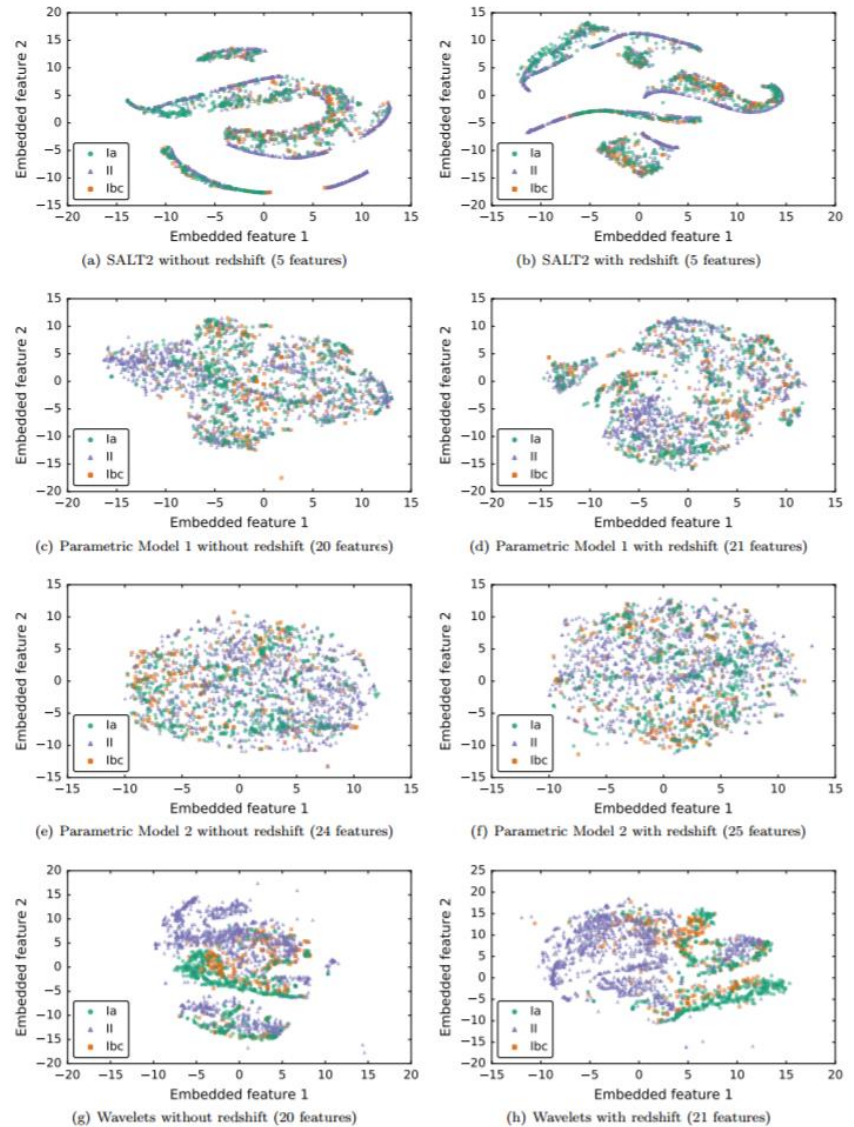


T-SNE



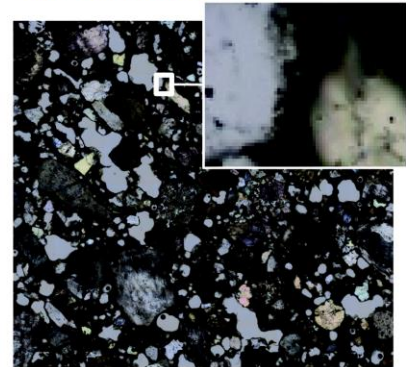
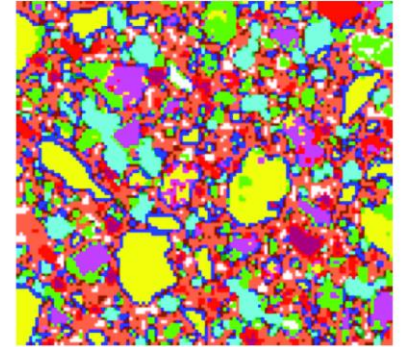
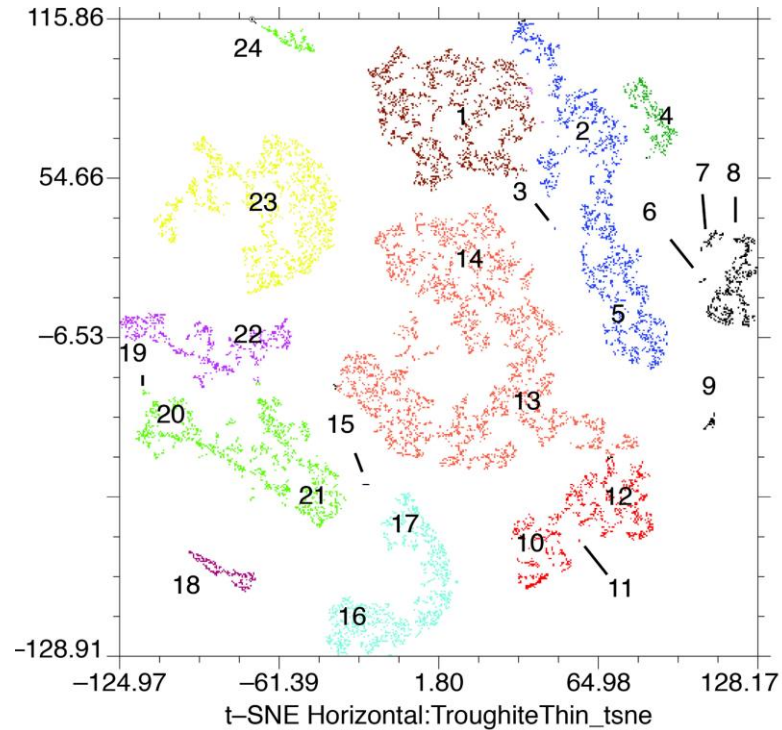
Jamieson, Andrew R., et al. "Exploring nonlinear feature space dimension reduction and data representation in breast CADx with Laplacian eigenmaps and t-SNE." *Medical physics* 37.1 (2010): 339-351.

# Applications



Lochner, Michelle, et al. "Photometric Supernova Classification with Machine Learning." *arXiv preprint arXiv:1603.00882* (2016).

# Applications



Thompson, David R., et al. "Automating X-ray Fluorescence Analysis for Rapid Astrobiology Surveys." *Astrobiology* 15.11 (2015): 961-976.

# Matlab/Python

- `sklearn.manifold.TSNE`
  - `N_components`: 2/3
  - Perplexity (how big is the Gaussian neighborhood)
- <https://lvdmaaten.github.io/drtoolbox/>
  - Has a ton of different tools aside from TSNE