

Spectral Clustering

Machine Learning and Deep Learning - 2021

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November 4th, 2021

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Clustering model based on the spectral graph theory.

- build a graph over examples, representing it with the adjacency matrix A

$$A_{i,j} = e^{-\frac{\sum_{k=1}^d ||x_i^k - x_j^k||^2}{\sigma^2}}$$

- build the degree matrix D of the graph. It is a diagonal matrix holding for each element the sum of the incoming edges.
- compute the normalized laplacian L

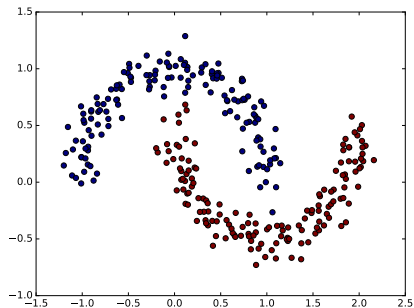
$$L = D - A \tag{1}$$

- Compute the eigenvectors and sort them for increasing eigenvalues
- Choose the eigenvectors from the second to the desired number of clusters
- Those eigenvectors provide a representation of data in a fancy embedding space!

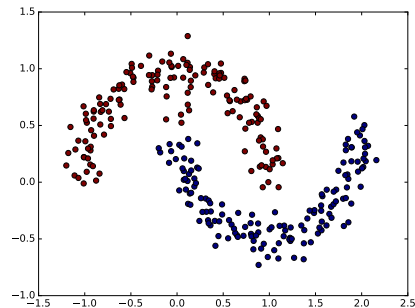
You may implement the solution according to one of the following:

- **Fiedler-Vector Solution:** in the event of a **binary clustering problem**, you may simply consider the eigenvector corresponding to the 2nd smallest eigenvalue (i.e. the Fiedler Vector) and use the positivity of its components as labels;
- **K-Means solution:** in a more general case where you need n **distinct clusters**, you may consider the eigenvectors corresponding to the n smallest eigenvalues and use their components as the representation of old data in a new embedding space. You can apply K-Means to this new data representation.

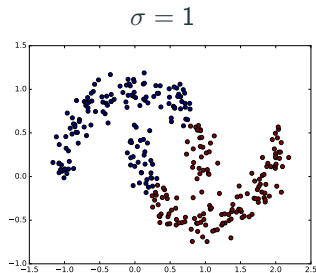
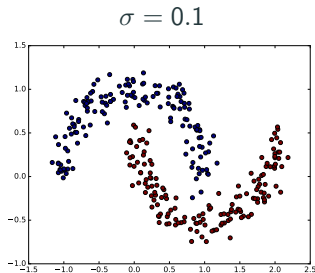
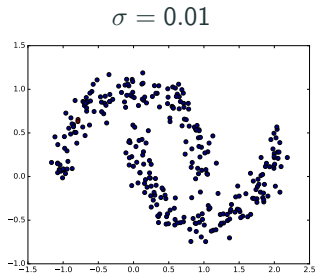
GT



Result



Beware: σ counts (in large amounts!)



- `datasets.gaussian_dataset`

```
data, cl = gaussians_dataset(3, [100,100,70], [[1, 1],[-4, 6],[8, 8]], [[1, 1],[3, 3],[1, 1]])
```

- `datasets.two_moon_dataset`

```
data, cl = two_moon_dataset(n_samples=300, noise=0.1)
```

- `matplotlib.pyplot.plot`
- `matplotlib.pyplot.scatter`
- `scipy.linalg.fractional_matrix_power`
- `numpy.linalg.eig`
- `numpy.argsort`