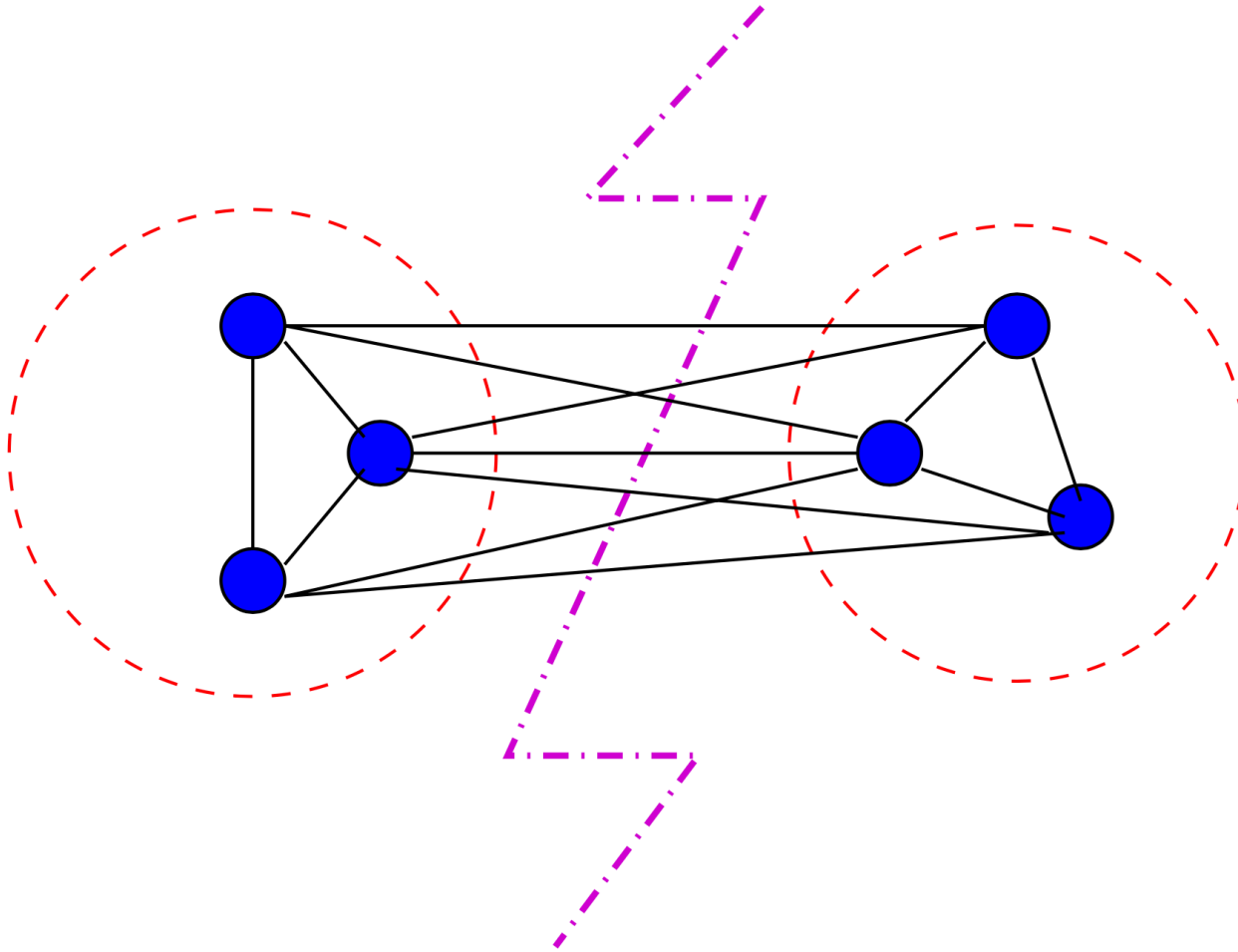


Spectral Clustering

What is Spectral Clustering?

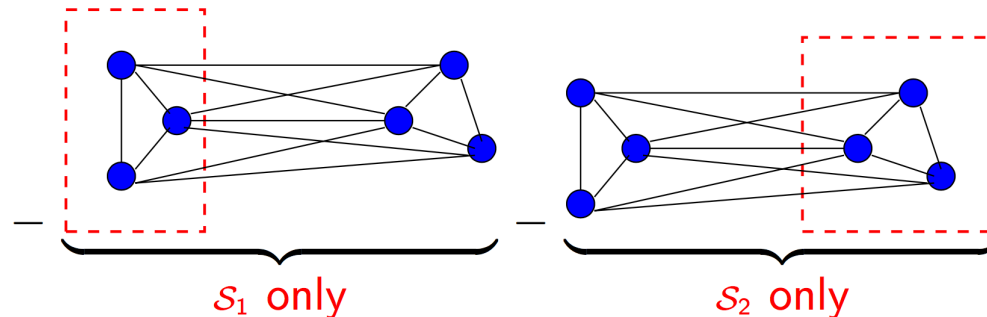
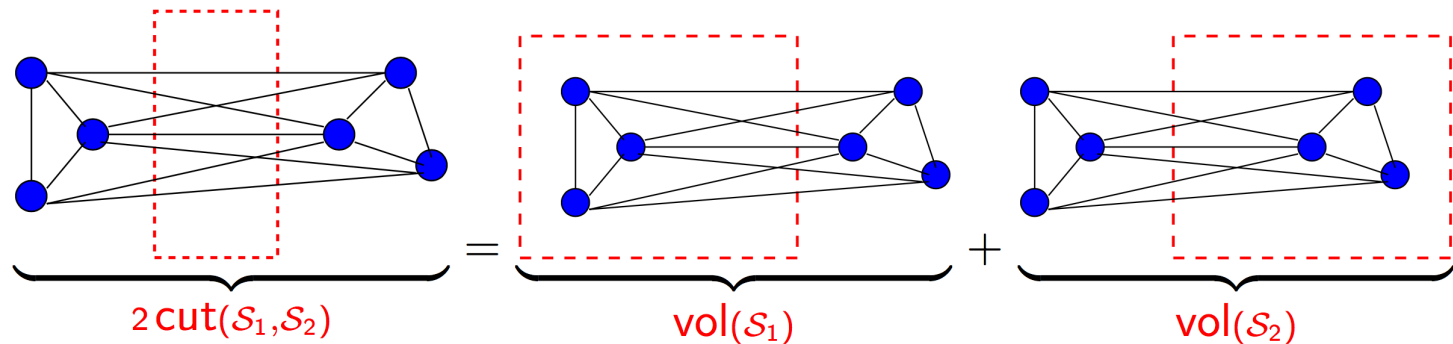
- Easy to implement
- Reasonably fast especially for sparse data
- Spectral clustering treats the data clustering as a graph partitioning problem **without making any assumption on the form of the data clusters using the spectrum of the Laplacian matrix**

Graph Bipartitioning



Graph Bipartitioning

- Affinity matrix $W = [W_{ij}] \in \mathbb{R}^{n \times n}$
- Degree of nodes $d_i = \sum_j W_{ij}$
- Volume $\text{vol}(\mathcal{S}_1) = d_{\mathcal{S}_1} = \sum_{i \in \mathcal{S}_1} d_i$



Graph Bipartitioning

$$\begin{aligned}x^\top Lx &= x^\top Dx - x^\top Wx \\&= \sum_{i=1}^n d_i x_i^2 - \sum_{i=1}^n \sum_{j=1}^n W_{ij} x_i x_j \\&= \frac{1}{2} \left(\sum_i d_i x_i^2 - 2 \sum_i \sum_j W_{ij} x_i x_j + \sum_j d_j x_j^2 \right) \\&= \frac{1}{2} \sum_i \sum_j W_{ij} (x_i - x_j)^2.\end{aligned}$$

Algorithm

- ① Construct affinity matrix

$$W_{ij} = \begin{cases} \exp\{-\beta\|v_i - v_j\|^2\} & \text{if } i \neq j \\ 0 & \text{if } i = j \end{cases}$$

- ② Calculate graph Laplacian $L = D - W$ where
 $D = \text{diag}\{d_1, \dots, d_n\}$ and $d_i = \sum_j W_{ij}$ is the degree of node v_i

- ③ Compute **second smallest eigenvector** of graph Laplacian
(denoted by $u = [u_1 \dots u_n]^\top$, Fiedler vector)

- ④ Partition u_i 's by pre-specified threshold value and assign data point to cluster v_i

