

# **CS 556: Computer Vision**

## **Lecture 19**

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# **Computational Framework of Perceptual Grouping**

# Perceptual Grouping: Image Segmentation

- Find contiguous clusters of pixels, so pixels within each cluster are
  - closer
  - more similar in terms of
    - color
    - texture
- Than pixels from different clusters

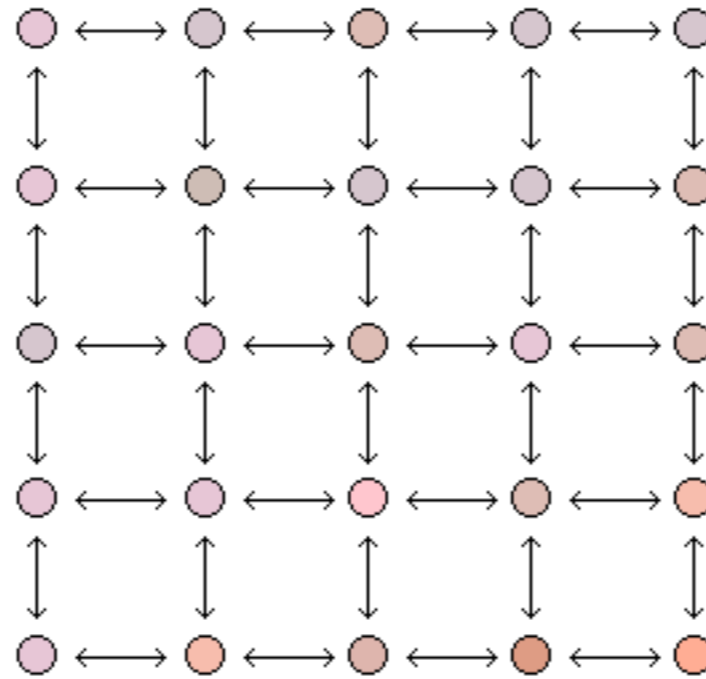
# Example: N-Cuts Segmentation

segments  
or  
regions

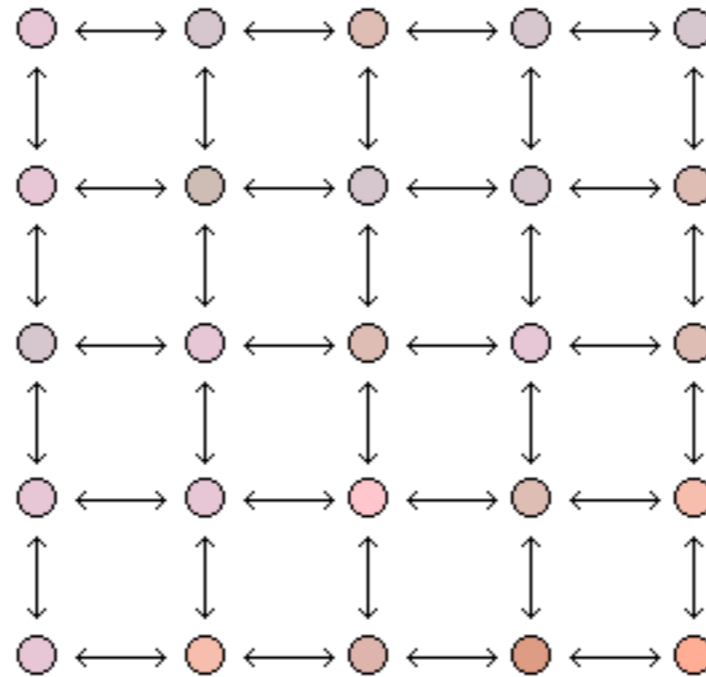


results of the Ncuts algorithm

# Graph-based Clustering

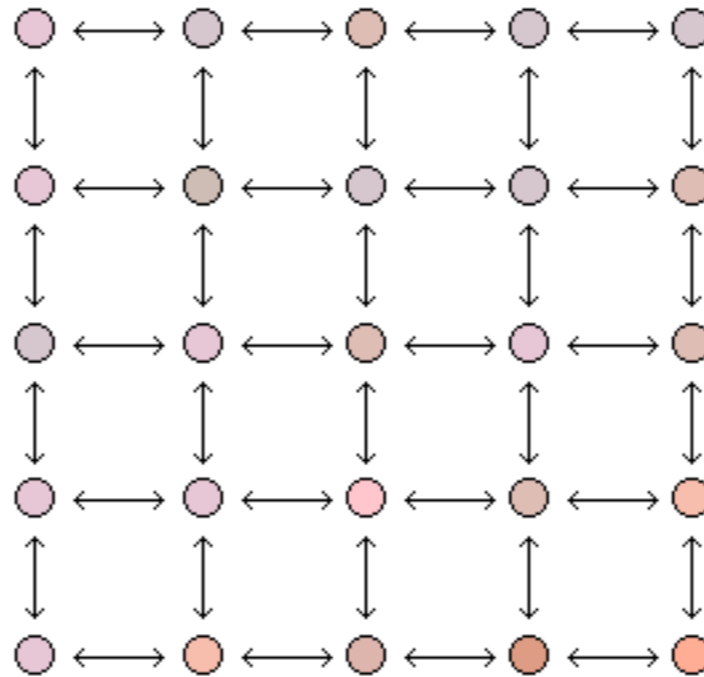


# Graph-based Clustering



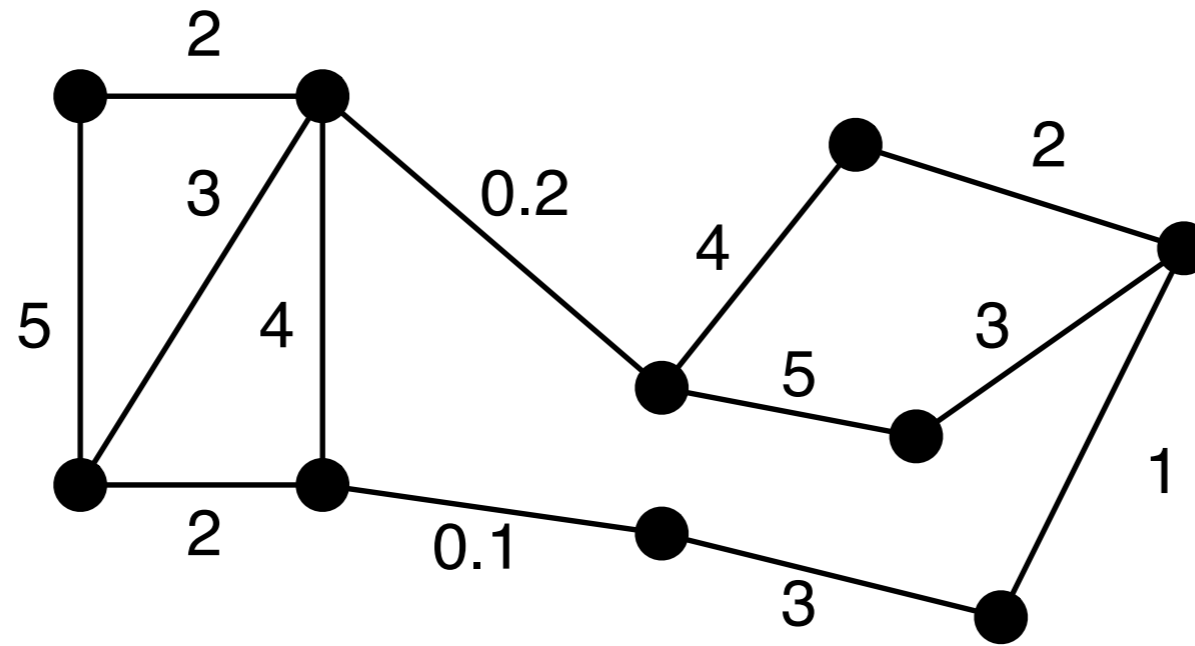
- Image elements are represented by a graph

# Graph-based Clustering



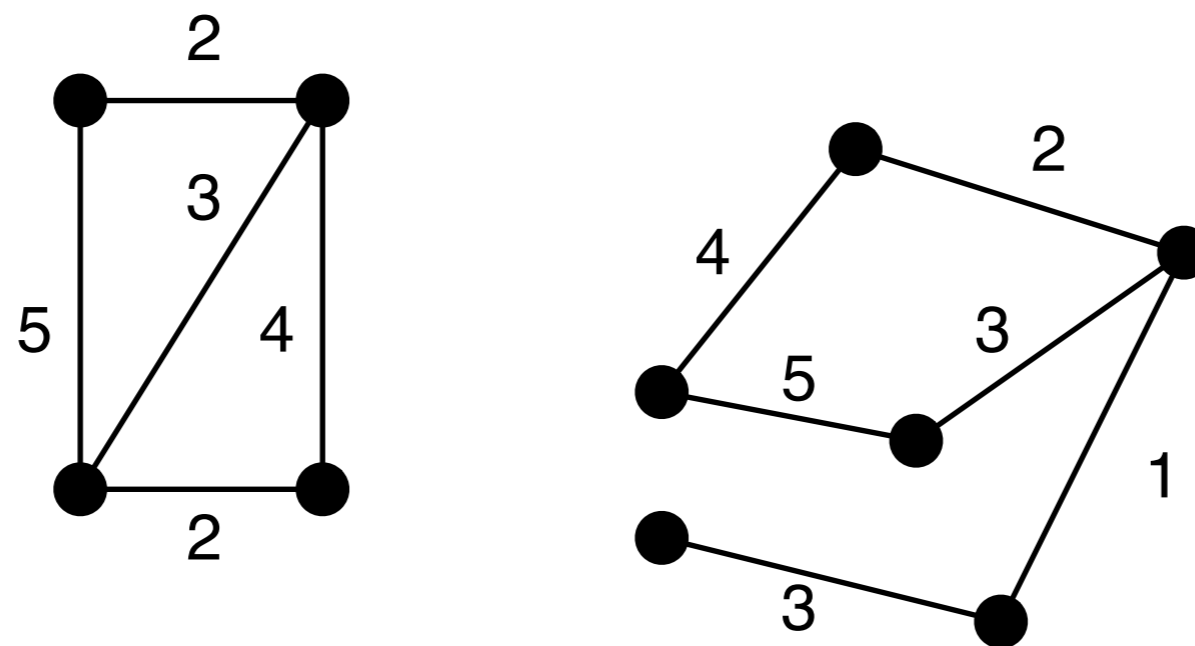
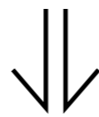
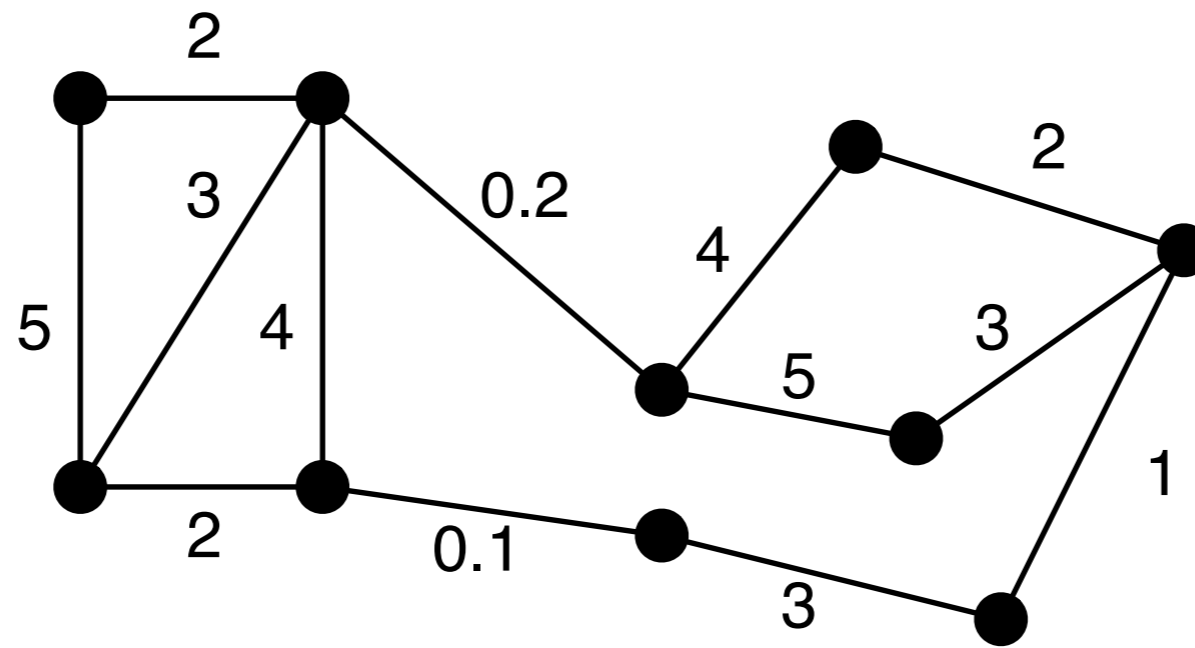
- Image elements are represented by a graph
- Clustering = Graph partitioning into subgraphs

# Major Goal: Graph Partitioning

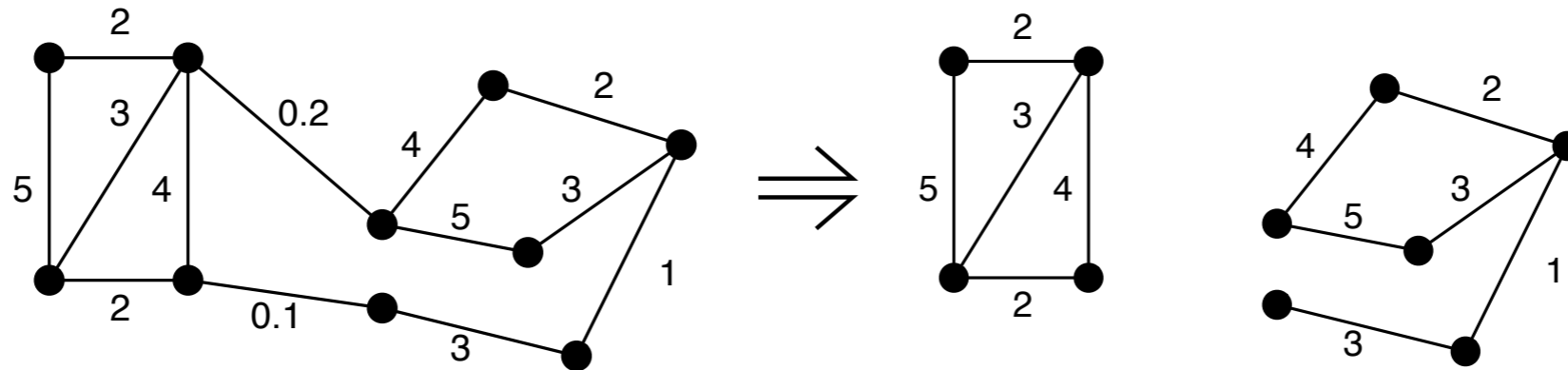




# Major Goal: Graph Partitioning



# Example: Pixel Clustering



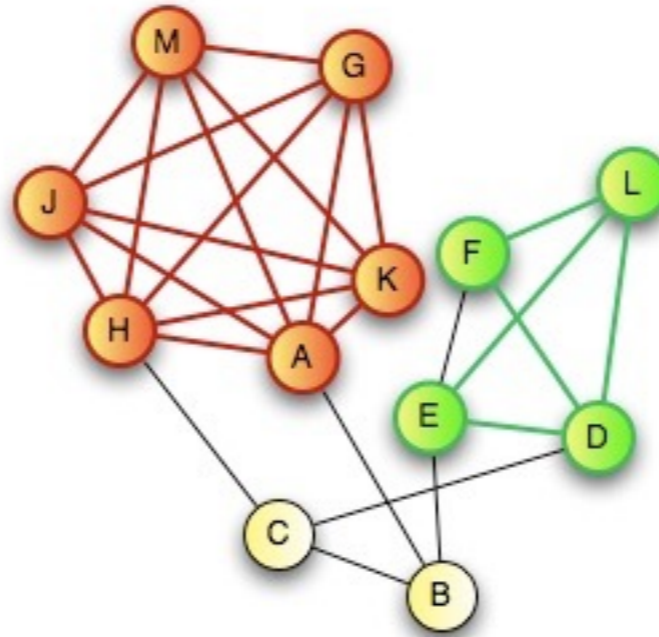
Find a cut, such that in one cluster all pixels are:

- neighbors
- more similar to one another than to pixels in the other cluster

# Graph Terminology

- Clique
- Adjacency matrix
- Node degree
- Volume
- Cut
- Association

# Clique



- A subgraph with all nodes fully connected
- Maximal clique
- Maximum weighted clique



# Adjacency Matrix

$$A = \begin{matrix} & & & j \\ & & & \vdots \\ i & & \dots & a_{ij} \\ & & & \end{matrix} \begin{matrix} \\ \\ \\ \end{matrix}$$

$$a_{ij} = \exp\left(-\frac{1}{2\sigma^2} \|\text{SIFT}_i - \text{SIFT}_j\|^2\right)$$

# Adjacency Matrix

$$A = \begin{matrix} & & & j \\ & & & \vdots \\ & & & \\ i & & \dots & a_{ij} \\ & & & \end{matrix} \left[ \begin{array}{c} \\ \\ \\ \\ \end{array} \right]$$

- Affinities between pairs of nodes (i,j) in the graph

$$a_{ij} = \exp \left( -\frac{1}{2\sigma^2} \|\text{SIFT}_i - \text{SIFT}_j\|^2 \right)$$

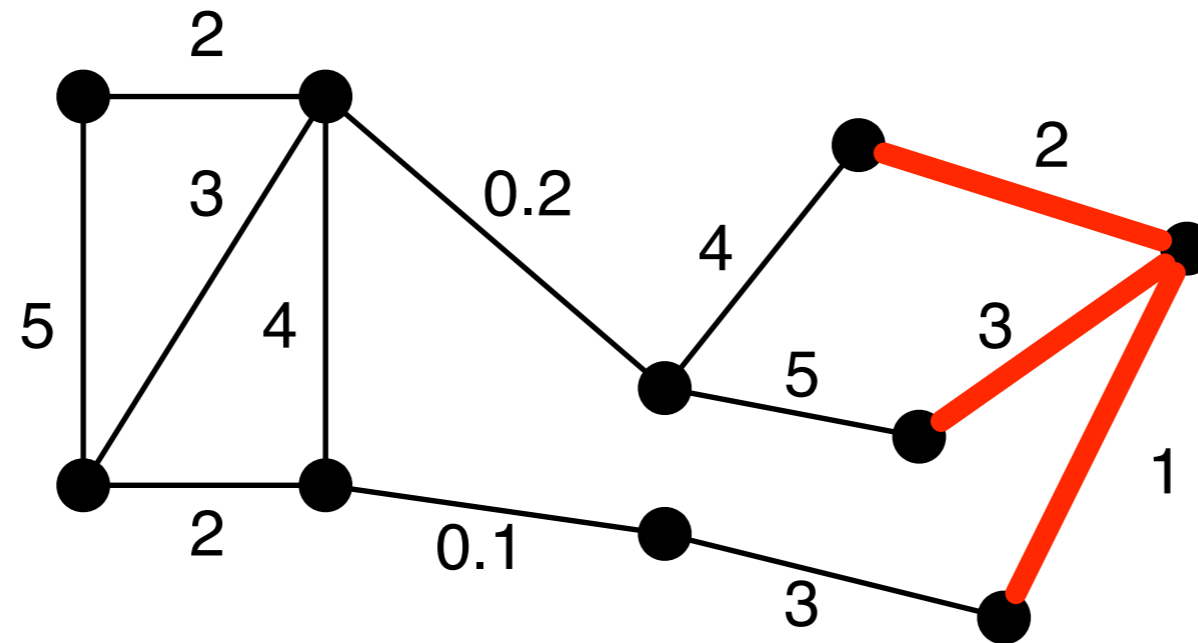
# Adjacency Matrix

$$A = \begin{matrix} & & & j \\ & & & \vdots \\ & & & \\ i & & \dots & a_{ij} \\ & & & \end{matrix} \left[ \begin{array}{c} \\ \\ \\ \end{array} \right]$$

- Affinities between pairs of nodes (i,j) in the graph
- Example: Nodes = Pixels  $\Rightarrow$

$$a_{ij} = \exp \left( -\frac{1}{2\sigma^2} \|\text{SIFT}_i - \text{SIFT}_j\|^2 \right)$$

# Node Degree

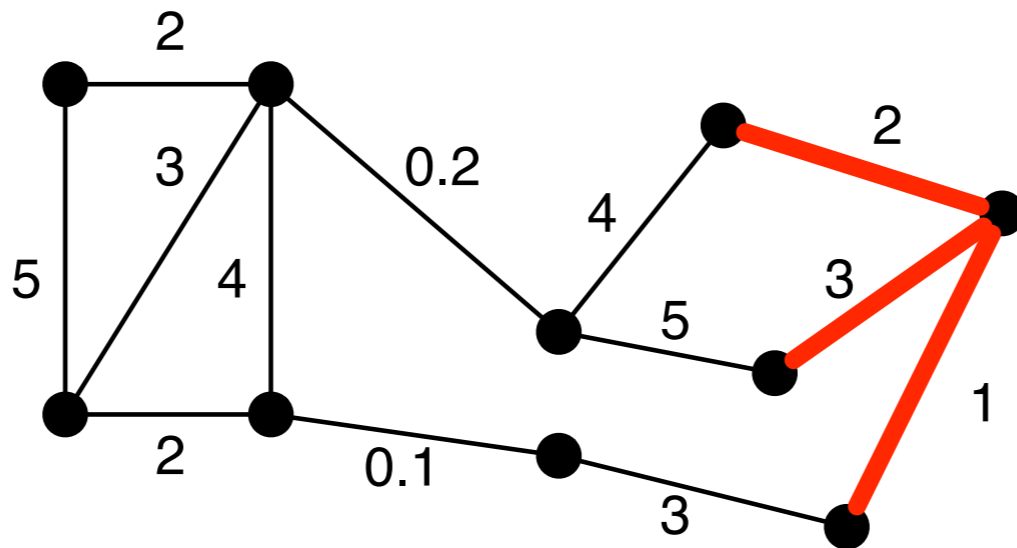


$$d_i = 6$$

$$d_i = \sum_j a_{ij}$$



# Node Degree

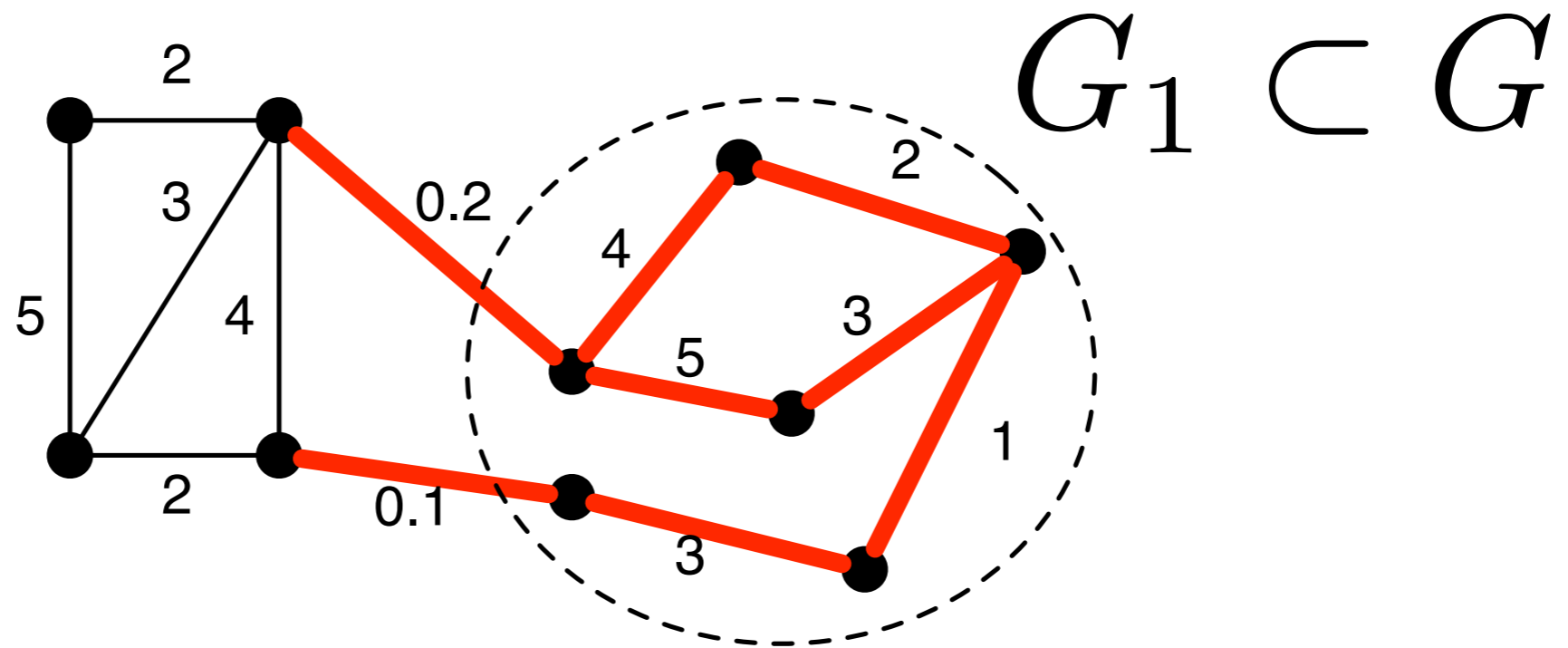


$$D = \begin{bmatrix} d_1 & 0 & 0 & \dots \\ 0 & d_2 & 0 & \\ \vdots & & & \\ \dots & 0 & 0 & d_N \end{bmatrix}$$

$$\mathbf{1}_i = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \leftarrow i$$

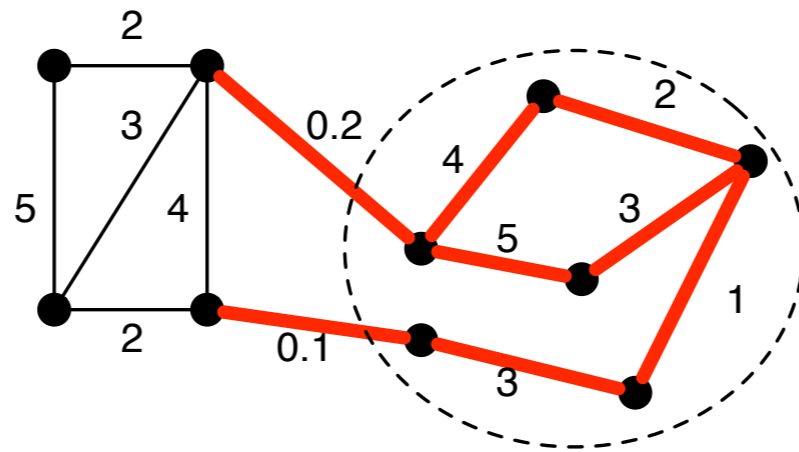
$$\Rightarrow d_i = \mathbf{1}_i^T D \mathbf{1}_i$$

# Volume of a Subgraph



$$\text{vol}(G_1) = \sum_{i \in G_1} d_i$$

# Volume of a Subgraph

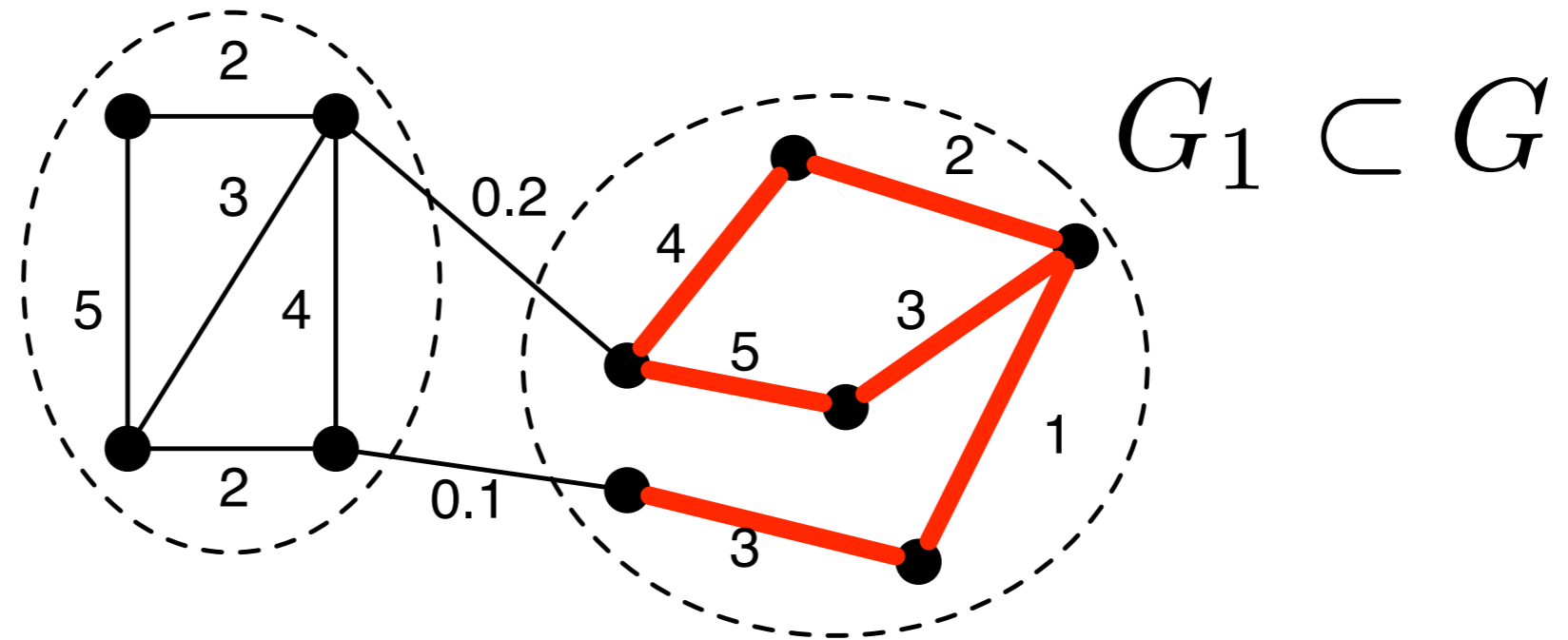


$$D = \begin{bmatrix} d_1 & 0 & 0 & \dots \\ 0 & d_2 & 0 & \\ \vdots & & & \\ \dots & 0 & 0 & d_N \end{bmatrix} \quad \mathbf{1}_{G_1} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}$$

$\swarrow \quad \searrow$   
 $G_1$

$$\Rightarrow \text{vol}(G_1) = \mathbf{1}_{G_1}^T D \mathbf{1}_{G_1}$$

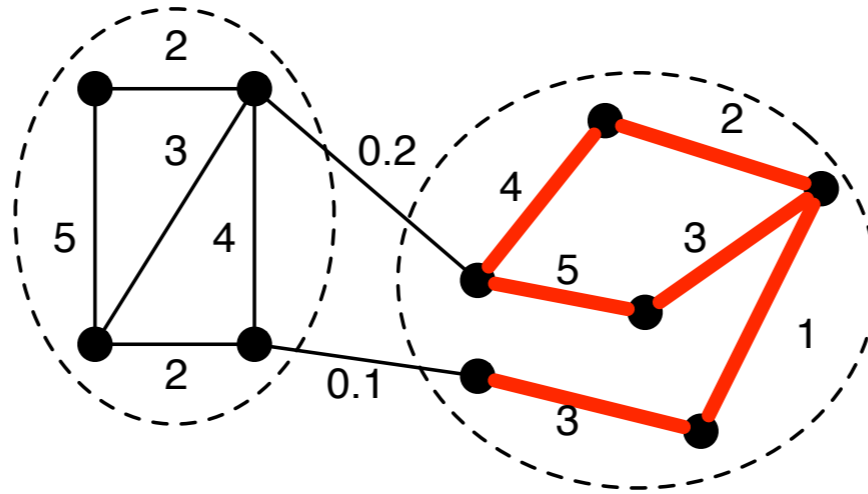
# Association within a Subgraph



$$\text{assoc}(G_1) = \sum_{i,j \in G_1} a_{ij}$$



# Association within a Subgraph

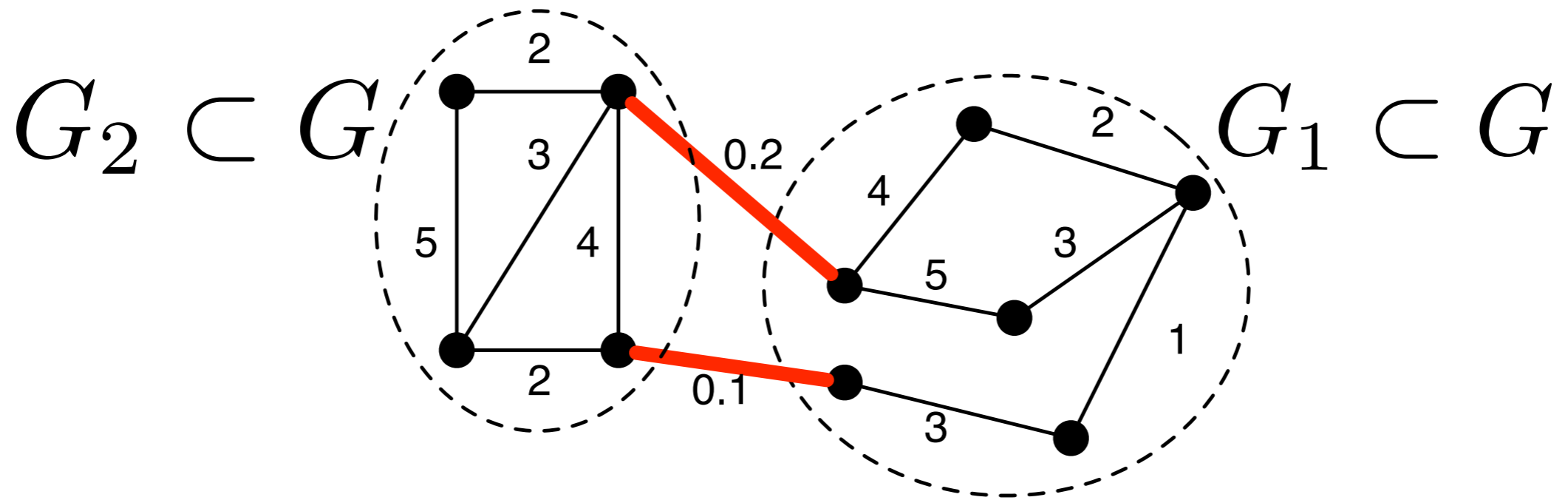


adjacency  
matrix

$$A = \begin{matrix} & & & & j \\ & & & & \vdots \\ & & & & a_{ij} \\ i & \left[ \begin{array}{c} \dots \\ \dots \\ \dots \end{array} \right] \end{matrix}$$

$$\text{assoc}(G_1) = \sum_{i,j \in G_1} a_{ij} = \mathbf{1}_{G_1}^T A \mathbf{1}_{G_1}$$

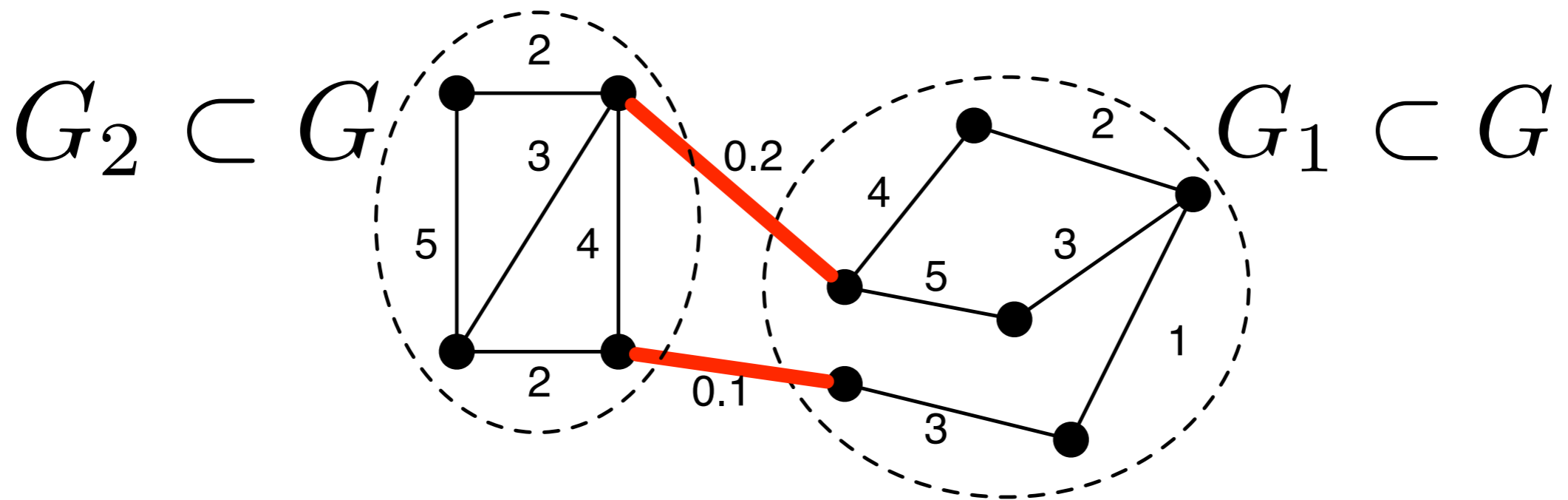
# Cut between Two Graph Partitions



$$G_1 \cap G_2 = \emptyset$$

$$\text{cut}(G_1, G_2) = \sum_{i \in G_1, j \in G_2} a_{ij}$$

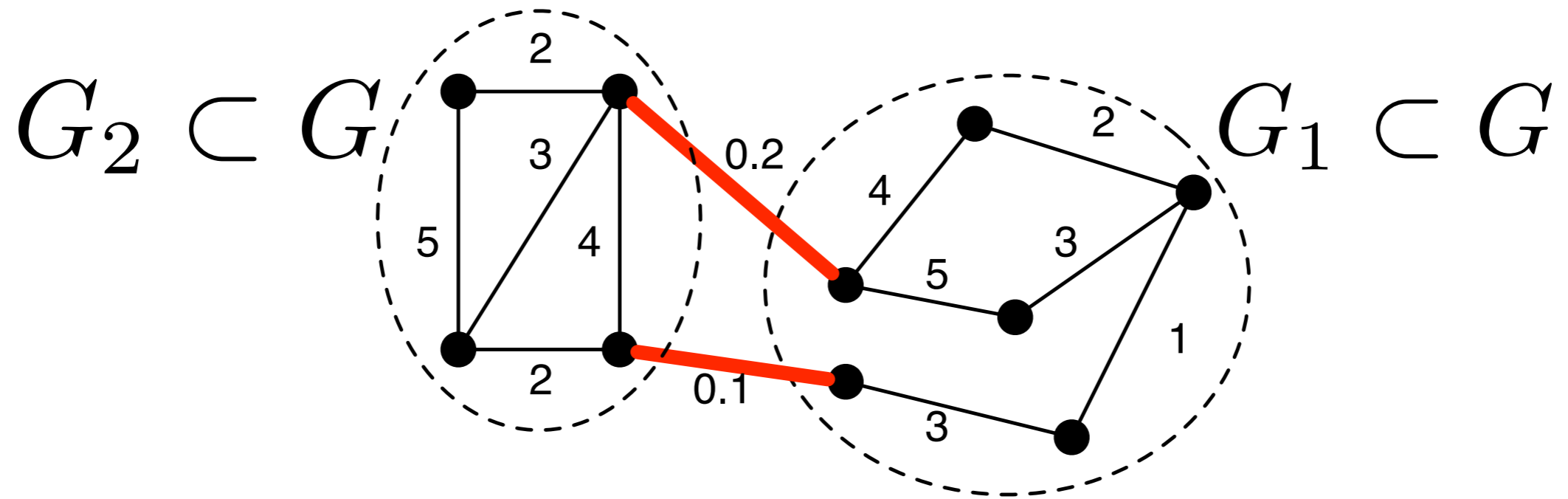
# Cut between Two Graph Partitions



$$G_1 \cap G_2 = \emptyset$$

$$\text{cut}(G_2, G_1) = \sum_{i \in G_2, j \in G_1} a_{ij}$$

# Cut between Two Graph Partitions

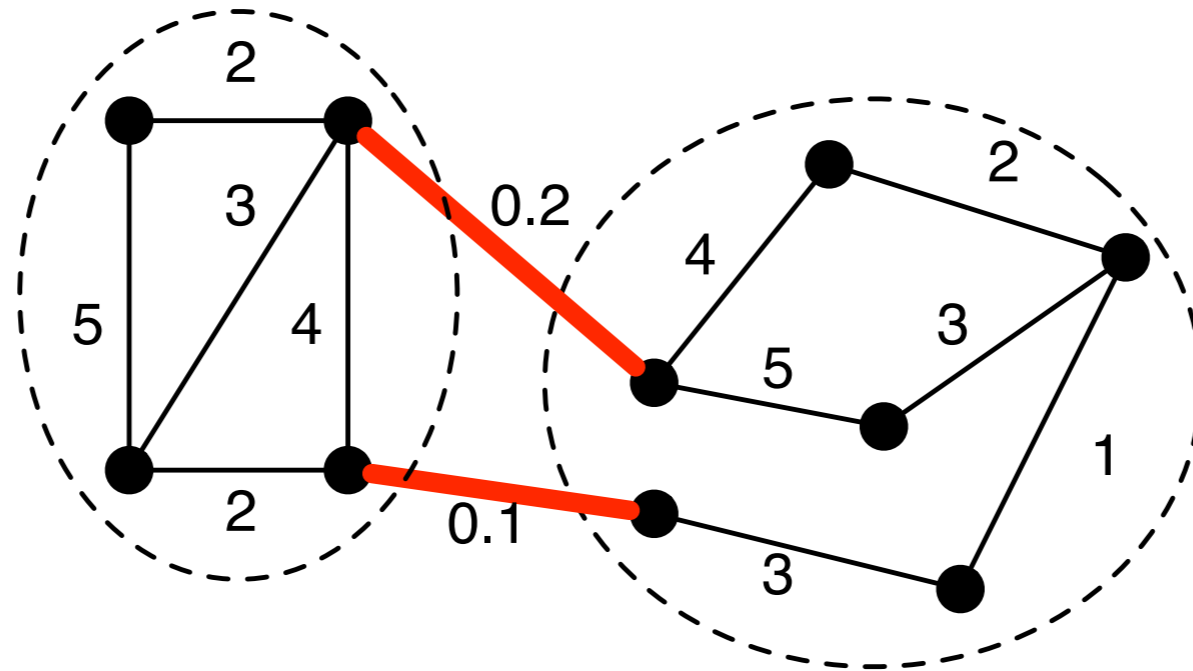


$$G_1 \cap G_2 = \emptyset$$

$$\text{cut}(G_1, G_2) \neq \text{cut}(G_2, G_1)$$

in general

# Cut between Two Graph Partitions

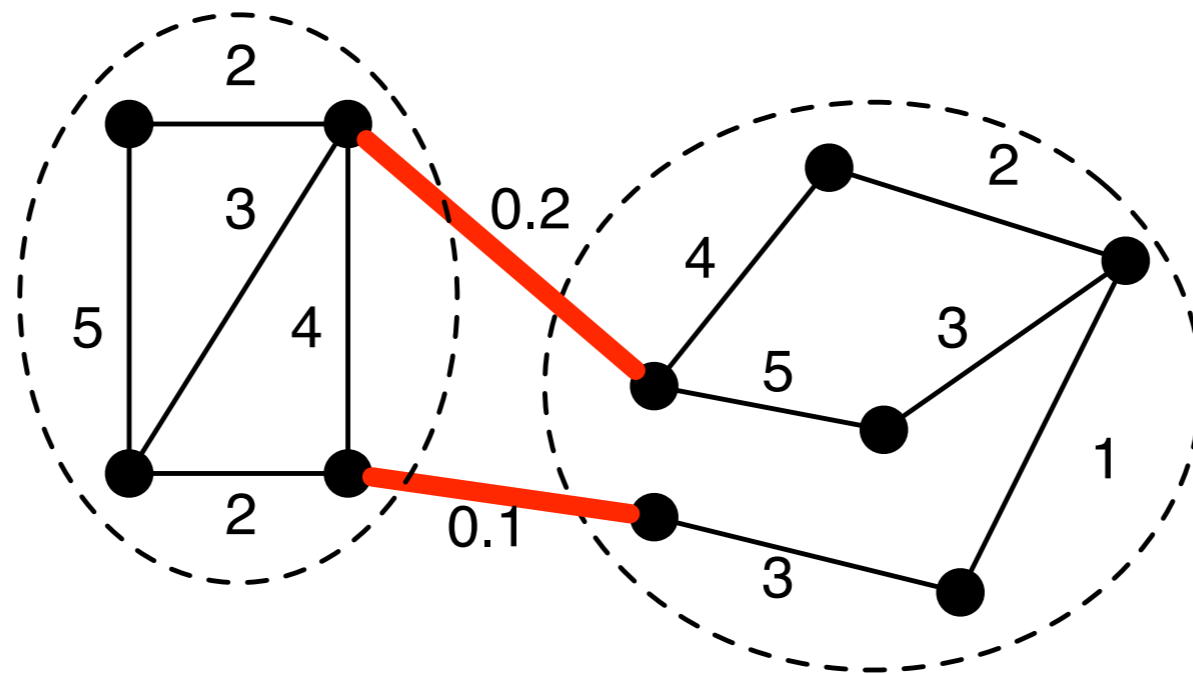


$$\text{cut}(G_1, G_2) = \text{vol}(G_1) - \text{assoc}(G_1)$$

$$= \mathbf{1}_{G_1}^T D \mathbf{1}_{G_1} - \mathbf{1}_{G_1}^T A \mathbf{1}_{G_1}$$

$$= \mathbf{1}_{G_1}^T (D - A) \mathbf{1}_{G_1}$$

# Cut between Two Graph Partitions



$$\text{cut}(G_2, G_1) = \mathbf{1}_{G_2}^T (D - A) \mathbf{1}_{G_2}$$

$$= (\mathbf{1} - \mathbf{1}_{G_1})^T (D - A) (\mathbf{1} - \mathbf{1}_{G_1})$$