

$$1) A = \begin{bmatrix} 0 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}$$

$$D = \begin{bmatrix} 4 & & & & & & & & & \\ & 2 & & & & & & & & \\ & & 4 & & & & & & & \\ & & & 2 & & & & & & \\ & & & & 3 & & & & & \\ & & & & & 2 & & & & \\ & & & & & & 3 & & & \\ & & & & & & & 3 & & \\ & & & & & & & & 3 & \\ & & & & & & & & & 2 \end{bmatrix}$$

$$\Rightarrow L = D - A = \begin{bmatrix} 4 & -1 & -1 & -1 & -1 & & & & & \\ -1 & 2 & -1 & & & & & & & \\ -1 & -1 & 4 & -1 & & & & & & -1 \\ -1 & & -1 & 2 & & & & & & \\ -1 & & & & 3 & -1 & -1 & & & \\ & & & & & -1 & 2 & -1 & & \\ & & & & & & -1 & 1 & 3 & -1 \\ & & & & & & & -1 & 3 & -1 & -1 \\ & & -1 & & & & & & -1 & 3 & -1 \\ & & & & & & & & & -1 & -1 & 2 \end{bmatrix}$$

2) a) U_1 is not the Fiedler vector because it is associated with 0 while the graph is connected: it is not the second smallest eigenvalue, but the smallest.

$$b) L \times U_2 = \begin{bmatrix} -4 - (-\sqrt{2} - 1 - \sqrt{2} + \sqrt{2} - 1) = -2 + \sqrt{2} = -1 \times (2 - \sqrt{2}) \\ -2\sqrt{2} - (-1 - 1) = -2\sqrt{2} + 2 = -\sqrt{2}(2 - \sqrt{2}) \\ -4 - (\sqrt{2} - 1 - \sqrt{2} + \sqrt{2} - 1) = -1 \times (2 - \sqrt{2}) \\ -2\sqrt{2} + 2 = -\sqrt{2}(2 - \sqrt{2}) \\ 3(\sqrt{2} - 1) - (\sqrt{2} + 1) = 3\sqrt{2} - 4 = (\sqrt{2} - 1)(2 - \sqrt{2}) \\ 2 - (\sqrt{2} - 1 + 1) = 1 \times (2 - \sqrt{2}) \\ 3 - (\sqrt{2} - 1 + 1 + 1) = 1 \times (2 - \sqrt{2}) \\ 3 - (1 + \sqrt{2} - 1 + 1) = 1 \times (2 - \sqrt{2}) \\ 3(\sqrt{2} - 1) - (\sqrt{2} + 1 + 1) = (\sqrt{2} - 1)(2 - \sqrt{2}) \\ 2 - (\sqrt{2} + \sqrt{2} - 1) = 1 \times (2 - \sqrt{2}) \end{bmatrix} = (2 - \sqrt{2}) \times U_2$$

Thus v_2 is an eigenvector of L associated with $2 - \sqrt{2} (\approx 0.59)$

c): $L v_3 = \begin{cases} 4 - (6/5 + 1 + 6/5 + 4/5) = 3 - (12/5 + 4/5) = 3 - 16/5 = \frac{15-16}{5} = -1/5 \\ 2 \times \frac{6}{5} - 2 = \frac{12}{5} - 2 = \frac{12-10}{5} = \frac{2}{5} = \frac{2}{5} \times \frac{3}{3} = \frac{6}{5} \times \frac{1}{3} \\ 2 \times \frac{6}{5} - 2 = \frac{12}{5} - 2 = \frac{2}{5} = \frac{6}{5} \times \frac{1}{3} \\ 4 - (2 \times \frac{6}{5} + 1 - 4/5) = 3 - \frac{32}{5} \neq 1 \times \frac{1}{3} \\ \vdots \end{cases}$

Hence, v_3 is not an eigenvector of L (and can thus not be the Fiedler vector).

d): $L v_4 = \begin{cases} 6 \times 4 - (-4 + 6 - 4 - 4) = 6 \times 4 + 6 = 5 \times 6 \\ \vdots \end{cases}$

If v_4 is an eigenvector, it is associated with $5 > 2 - \sqrt{2}$ and this is thus not the Fiedler vector.

⇒ The Fiedler vector is then v_2 .

3) Isoperimetric Ratio: our graph is $G = (V, E)$, $m = |V|$

Let $S \subset V$, thus $\phi(S) = \frac{\text{cut}(S)}{\min(|S|, m - |S|)}$

4) $v_2 = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \\ x_9 \\ x_{10} \end{pmatrix}$

$(x_2, x_4, x_1, x_3, x_5, x_9, x_6, x_7, x_8, x_{10})$ is the list of v_2 's coord. in \uparrow order.

S	{2}	{2, 4}	{2, 4, 1}	{2, 4, 1, 3}	{2, 4, 1, 3, 5}	{2, 4, 1, 3, 5, 9}	{2, 4, 1, 3, 5, 9, 6}
$\phi(S)$	2	2/2 = 2	4/3 = 1.33	2/4 = 1/2	3/5 = 0.6	4/4 = 1	4/3
S		{2, 4, 1, 3, 5, 9, 6, 7}	V = {10}				
$\phi(S)$		3/2	2				

The minimum value of experimental ratio obtained via the Sweep Cut Method is $\frac{1}{2}$, with $S = \{1, 2, 3, 4\}$.

The resulting partitioning is:

$$G = \{ \{1, 2, 3, 4\}, \{5, 6, 7, 8, 9, 10\} \}.$$

5) A partitioning S^* obtained via the Sweep Cut Method always verifies the Cheeger inequalities:

$$\frac{\lambda_2}{2} \leq \varphi(S^*) \leq \sqrt{2\lambda_2 d_{\max}}$$

here: $\varphi(S^*) = \frac{1}{2}$; $\frac{\lambda_2}{2} = \frac{2-\sqrt{2}}{2} = 1 - \frac{\sqrt{2}}{2} \approx 0.414 < \frac{1}{2} = \varphi(S^*)$

and $\sqrt{2\lambda_2 d_{\max}} = \sqrt{2 \times (2-\sqrt{2}) \times 4} = 2 \times \sqrt{2 \times (2-\sqrt{2})} > 2 > \varphi(S^*)$.

6)

P & C:

Agreement / Disagreement

	12 (111)	9 (110)	(21)
	0 (101)	24 (100)	(24)
	(12)	(33)	

Confusion / Native

4	0
0	3
0	3

S, SC

4	1	(5)
8	32	(40)
(12)	(33)	

2	2				4
		2	1		3
			1	2	3
2	2	2	2	2	

7)

$$ARI(P, G) = \frac{2 \times (12 \times 24 - 0)}{24 \times 12 + 21 \times 33} \approx 0.5872$$

$$ARI(P, K) = \frac{2 \times (4 \times 32 - 8)}{5 \times 33 + 12 \times 40} \approx 0.3721$$

$$\Pi I(P, G) = \frac{4}{10} \times \log_2 \left(\frac{10 \times 4}{16} \right) + \frac{3}{10} \times \log_2 \left(\frac{10 \times 3}{18} \right) \times 2$$

$$\approx 0.971 \quad (0.673 \text{ if } \log \text{ instead of } \log_2)$$

$$MI(P, K) = \frac{2}{10} \times \log_2 \left(\frac{10 \times 2}{8} \right) \times 2 + \frac{2}{10} \times \log_2 \left(\frac{10 \times 2}{6} \right) \times 2 + \frac{1}{10} \times \log_2 \left(\frac{10 \times 1}{5} \right) \times 2$$

$$\approx 1.3710 \quad (0.9503 \text{ if } \log \text{ instead of } \log_2)$$

Thus, using ARI, G is considered closer to P than K while MI metrics, on the opposite, considers K as closer to P than G.

This shows the inconsistency that may exist btw these measures.

$$8) \quad \mathcal{Q}(P, G) = \frac{1}{14} \times \left(\left[5 - \frac{12^2}{56} \right] + 2 \left[3 - \frac{8^2}{56} \right] \right) \approx 0.4888$$

$$\mathcal{Q}(G, G) = \frac{1}{14} \times \left(\left[5 - \frac{12^2}{56} \right] + 2 \cdot \left[7 - \frac{16^2}{56} \right] \right) \approx 0.347$$

$$\mathcal{Q}(K, G) = \frac{1}{14} \times \left(2 \cdot \left[1 - \frac{6^2}{56} \right] + 2 \cdot \left[1 - \frac{5^2}{56} \right] + \left[1 - \frac{6^2}{56} \right] \right) \approx 0.1556$$

$$\phi(G, \rho) = \frac{2}{12} + \frac{2}{8} + \frac{2}{8} = \frac{4+12}{24} = \frac{2}{3}$$

$$\phi(G, \epsilon) = \frac{2}{12} + \frac{2}{16} = \frac{8+6}{48} = \frac{14}{48} = \frac{7}{24}$$

$$\phi(G, \kappa) = \frac{4}{6} + \frac{4}{6} + \frac{3}{5} + \frac{3}{5} + \frac{4}{6} = \frac{4}{2} + \frac{6}{5} = \frac{16}{5}$$

Modularity is more consistent than normalised cuts here.

$$\pi(\rho, \delta) = \frac{2}{12} + \frac{2}{12} + \frac{2}{12} = \frac{6}{12} = \frac{1}{2}$$

(faint text)

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(faint text)

The choice of normalisation that most suits the data is the one that gives the most consistent results. In this case, the normalisation that gives the most consistent results is the one that gives the most consistent results.

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