

OVERVIEW

- semi-supervised learning - graphs.

↳ spectral clustering.

- Clustering:

- k-means

- GMM.

- spectral clustering.

- evaluation.

- Matrix Factorisation

- SVD

- low-rank approximation

- PCA.

- Robust PCA.

ROBUST PCA

Q:

$X \in \mathbb{R}^{m \times n}$, $D_\tau(X)$ - singular value threshold

Prove! (a) $\|D_\tau(X)\|_* \leq \|X\|_*$

(b) $\text{rank}(D_\tau(X)) \leq \text{rank}(X)$

A:

$$X = U \cdot \Sigma \cdot V^T$$

$$D_\tau(X) = U \cdot S_\tau(\Sigma) \cdot V^T$$

$$S_\tau(x) = \begin{cases} x - \tau & x > \tau \\ 0 & |x| \leq \tau \\ x + \tau & x < -\tau. \end{cases}$$

$$\|X\|_* = \sum_{i=1}^r \sigma_i \quad \leftarrow \text{singular values}$$

(a) $\sigma_1, \dots, \sigma_r$ - are singular values of X .

$\Rightarrow S_\tau(\sigma_1), \dots, S_\tau(\sigma_r)$ - " " " of $D_\tau(X)$

$$S_\tau(\sigma_i) \leq \sigma_i \quad (\sigma_i \geq 0)$$

$$\hookrightarrow \begin{pmatrix} \sigma_i - \tau & \sigma_i > \tau \\ 0 & \sigma_i < \tau \end{pmatrix} \leq \sigma_i$$

$$\Rightarrow \|D_\tau(X)\|_* = \sum_{i=1}^r S_\tau(\sigma_i) \leq \sum_{i=1}^r \sigma_i = \|X\|_*$$

(b) If $\sigma_i > 0 \Rightarrow S_\tau(\sigma_i) > 0$
 $\quad \quad \quad = 0$

If $\sigma_i = 0 \Rightarrow S_\tau(\sigma_i) = 0$

\Rightarrow total number of non-zero singular values
 can only go down

$$\Rightarrow \underbrace{\text{rank}(D_\tau(X))}_{\substack{\# \text{ non-zero} \\ \text{Sing. values}}} \leq \text{rank}(X)$$

Q:

When is: $\|D_\tau(X)\|_* = \|X\|_*$?

$\text{rank}(D_\tau(X)) = \text{rank}(X)$?

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$$

$$\begin{array}{ccccccc} \sigma_1 & \geq & \sigma_2 & \geq & \dots & \geq & \sigma_r > 0 \\ \downarrow & & \downarrow & & & & \downarrow \\ S_\tau(\sigma_1) & \geq & S_\tau(\sigma_2) & \geq & \dots & \geq & S_\tau(\sigma_r) \stackrel{?}{>} 0 \end{array}$$

$\text{rank}(D_\tau(X)) = \text{rank}(X)$ iff $S_\tau(\sigma_r) > 0$

\updownarrow

$$\sigma_r > \tau$$

\updownarrow

$$\sigma_1, \dots, \sigma_r > \tau$$

$$S_{-\tau}(\sigma_i) \stackrel{?}{=} \sigma_i \quad (\tau > 0)$$

$$\checkmark 0 \quad \sigma_i - \tau X$$

$$0 = \sigma_i \rightarrow \text{for all } i=1, \dots, r$$

$$X = 0$$

Q:

$$X = \begin{pmatrix} 1 & -2 & 0 & 1 \\ 2 & 3 & 1 & 0 \\ 3 & 1 & 72 & 1 \\ 85 & 4 & 0 & -2 \\ -1 & -5 & 77 & 66 \end{pmatrix}$$

Find $X = L + E$ s.t. $\text{rank}(L) \leq 2$
 $\|E\|_0 \leq 5$. (25%)

$$E = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 71 & 0 \\ 84 & 0 & 0 & 0 \\ 0 & 0 & 78 & 65 \end{pmatrix}$$

$$\|E\|_0 = 4 \checkmark$$

Take:

$$L = \begin{pmatrix} 1 & -2 & 0 & 1 \\ 2 & 3 & 1 & 0 \\ 3 & 1 & 1 & 1 \\ -2 & 4 & 0 & -2 \\ -1 & -5 & -1 & 1 \end{pmatrix}$$

$\text{rank}(L) = 2 \checkmark$

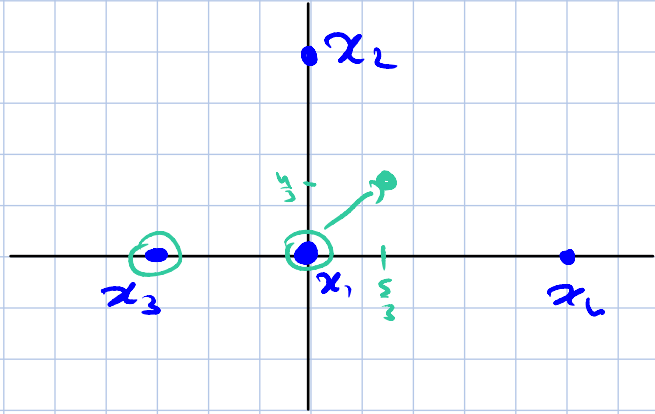
$$\rightarrow R_3 = R_1 + R_2$$

$$\rightarrow R_4 = 2 \cdot R_1$$

$$\rightarrow R_5 = R_1 - R_2$$

k-means

$$x_1 = (0,0) \quad x_2 = (0,4), \quad x_3 = (-3,0) \\ x_n = (5,0)$$



run k-means with $k=2$

$$\mu_1^{(0)} = (0,0), \quad \mu_2^{(0)} = (-3,0)$$

Solution: Step 1: $C_1 = \{x_1, x_2, x_n\}$

$$C_2 = \{x_3\}$$

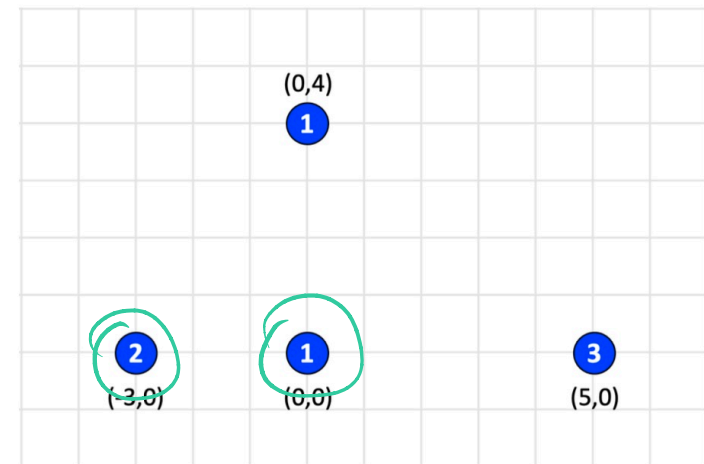
$$\mu_1^{(1)} = \left(\frac{5}{3}, \frac{4}{3}\right), \quad \mu_2^{(1)} = (-3,0)$$

Step 2: $C_1 = \{x_1, x_2, x_n\}$

$$C_2 = \{x_3\}$$

no changes

stop



$$x_1 = (0,0), \quad x_2 = (0,4) \quad x_3 = x_n = (-3,0)$$

$$x_5 = x_6 = x_7 = (5,0)$$

$$\mu_1^{(0)} = (0,0), \quad \mu_2^{(0)} = (-3,0)$$

Solution:

Step 1: $C_1 = \{x_1, x_2, x_5, x_6, x_7\}$

$$C_2 = \{x_3, x_4\}$$

$$M_1^{(1)} = \left(3, \frac{4}{5}\right), \quad M_2^{(1)} = (-3, 0)$$

Step 2: $C_1 = \{x_2, x_5, x_6, x_7\}$

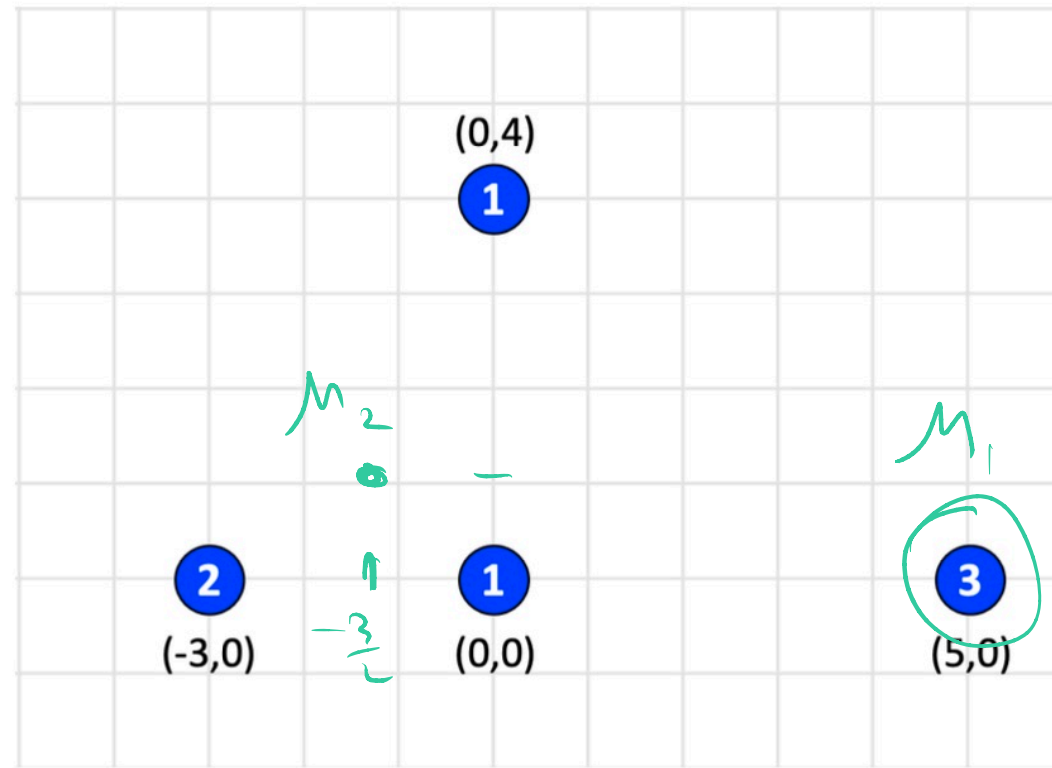
$$C_2 = \{x_1, x_3, x_4\}$$

$$M_1^{(2)} = \left(\frac{15}{4}, 1\right), \quad M_2^{(2)} = (-2, 0)$$

Step 3: $C_1 = \{x_5, x_6, x_7\}$

$$C_2 = \{x_1, x_2, x_3, x_4\}$$

$$M_1^{(3)} = (5, 0), \quad M_2^{(3)} = \left(-\frac{3}{2}, 1\right)$$

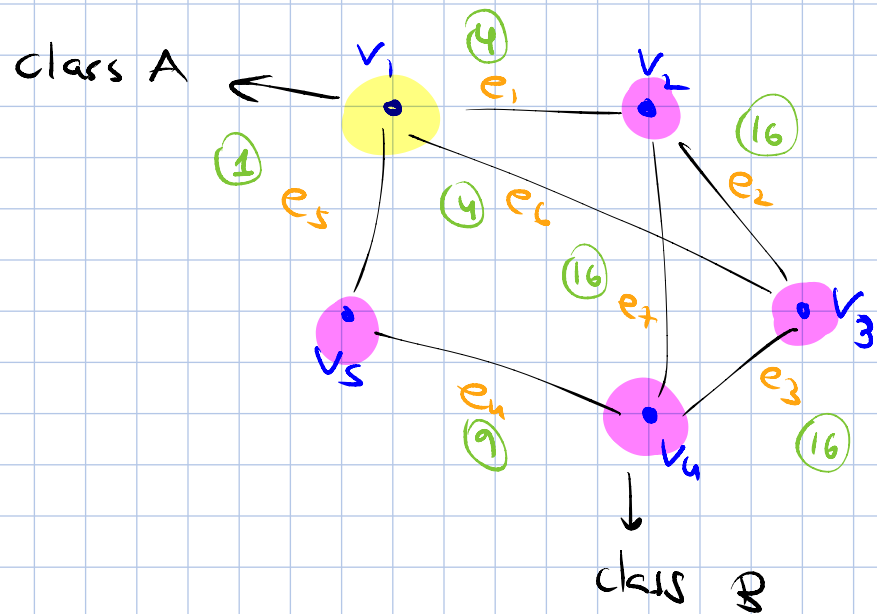


Step 4: $C_1 = \{x_5, x_6, x_7\}$

$$C_2 = \{x_1, x_2, x_3, x_4\}$$

↓
no change.

Semi-supervised Learning



Labelled: $V_L = \{1, 4\}$

$V_U = \{2, 3, 5\}$

$$L = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 9 & -4 & -4 & 0 & -1 \\ -4 & 36 & -16 & -16 & 0 \\ -4 & -16 & 36 & -16 & 0 \\ 0 & -16 & -16 & 9 & -9 \\ -1 & 0 & 0 & -9 & 10 \end{pmatrix} \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix}$$

$$P_L = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} \quad P_U = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

known labels: $y = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

Solve: $Ag = b$

$$A = P_U L P_U^T = \begin{pmatrix} 36 & -16 & 0 \\ -16 & 36 & 0 \\ 0 & 0 & 10 \end{pmatrix}$$

$$b = -(P_U L P_U^T) \cdot y = \begin{pmatrix} -4 & -16 \\ -4 & -16 \\ -1 & -9 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -16 \\ -16 \\ -9 \end{pmatrix}$$

$$g^* = \begin{pmatrix} 0.8 \\ 0.8 \\ 0.9 \end{pmatrix} \rightarrow f^* = \begin{pmatrix} 0 \\ 0.8 \\ 0.8 \\ 1 \\ 0.9 \end{pmatrix} \rightarrow \text{Labels: } \begin{pmatrix} 0 \\ - \\ - \\ - \\ - \end{pmatrix}$$

Suppose - no labels \rightarrow Spectral clustering!

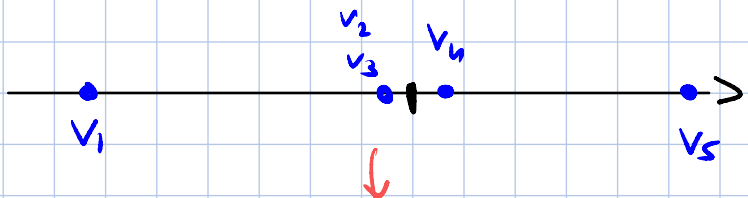
$$L = \begin{pmatrix} 9 & -4 & -4 & 0 & -1 \\ -4 & 36 & -16 & -16 & 0 \\ -4 & -16 & 36 & -16 & 0 \\ 0 & -16 & -16 & 41 & -9 \\ -1 & 0 & 0 & -9 & 10 \end{pmatrix}$$

$$\lambda_1 = 0$$

$$\lambda_2 \geq \lambda_1$$

\Leftrightarrow

$$\lambda_2 = 9.276$$

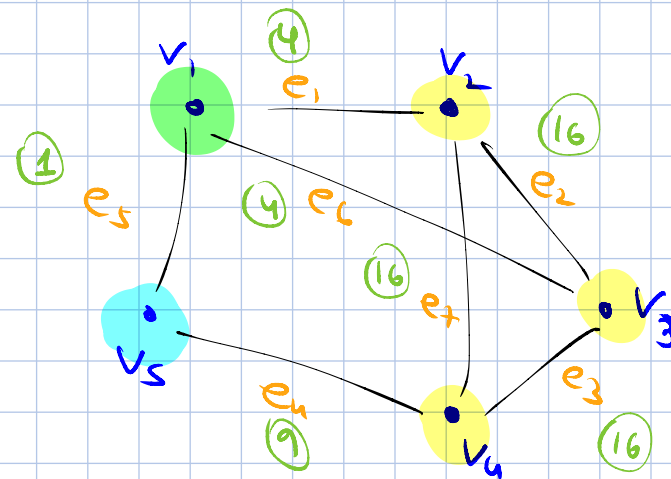


3 clusters!

$$u_2 = \begin{pmatrix} -0.7 \\ -0.06 \\ -0.06 \\ 0.41 \\ 0.69 \end{pmatrix} \begin{matrix} - \\ - \\ - \\ + \\ + \end{matrix}$$

Labels:

$$\begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 1 \end{pmatrix}$$



Spectral clustering for $k=3$!

$$\lambda_3 = 13.83$$

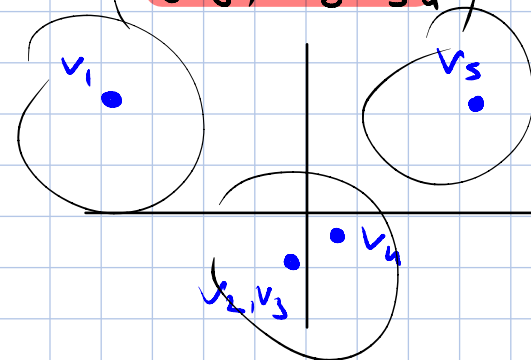
$$u_3 = \begin{pmatrix} 0.55 \\ -0.4 \\ -0.4 \\ -0.29 \\ 0.54 \end{pmatrix}$$

$$\rightarrow U = \begin{matrix} u_2 & u_3 \\ \hline -0.7 & 0.55 \\ -0.06 & -0.4 \\ -0.06 & -0.4 \\ 0.41 & -0.29 \\ 0.69 & 0.54 \end{matrix}$$

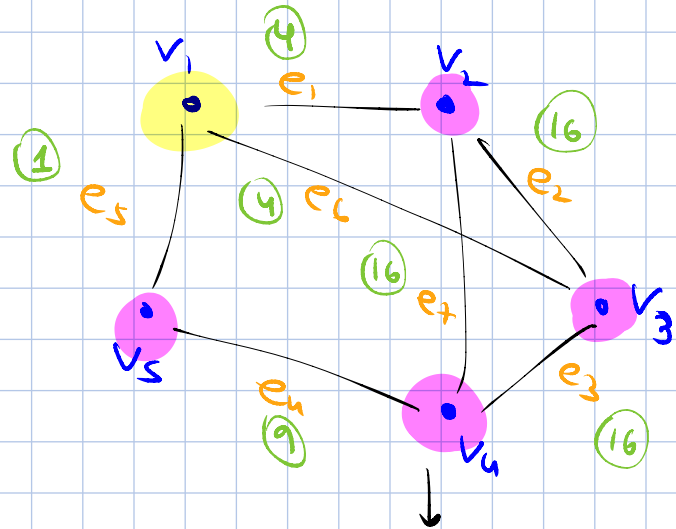
$$P(v_1) = (-0.7, 0.55)$$

$$P(v_2) = (-0.06, -0.4)$$

\vdots



Evaluation of clustering

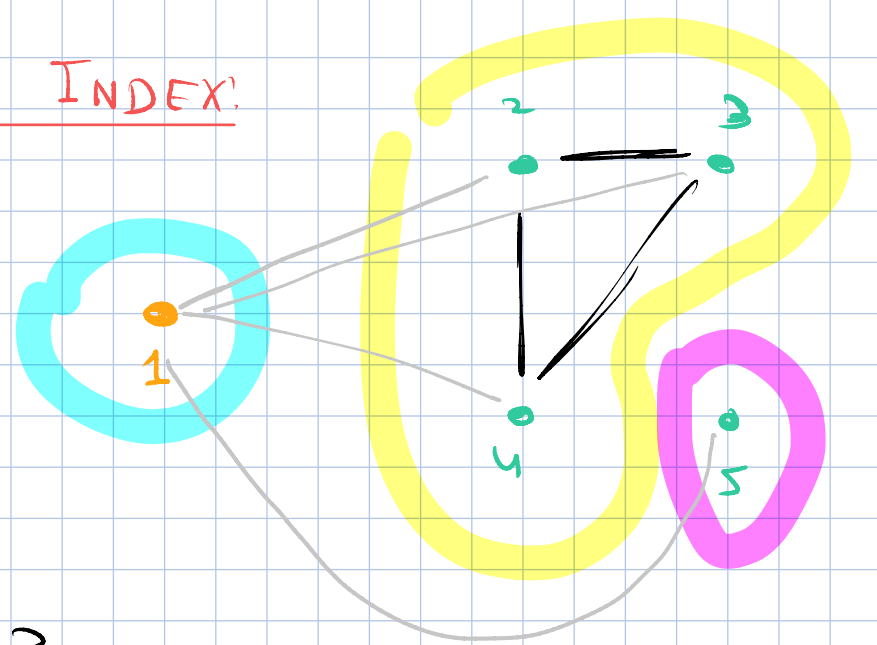


External evaluation:

Out clusters: $C_1 = \{1\}$, $C_2 = \{2, 3, 4, 5\}$

Ground truth: $C_1^* = \{1\}$, $C_2^* = \{2, 3, 4\}$
 $C_3^* = \{5\}$

RAND INDEX:



$$TP = 3$$

$$TN = 4$$

$$FP = 3$$

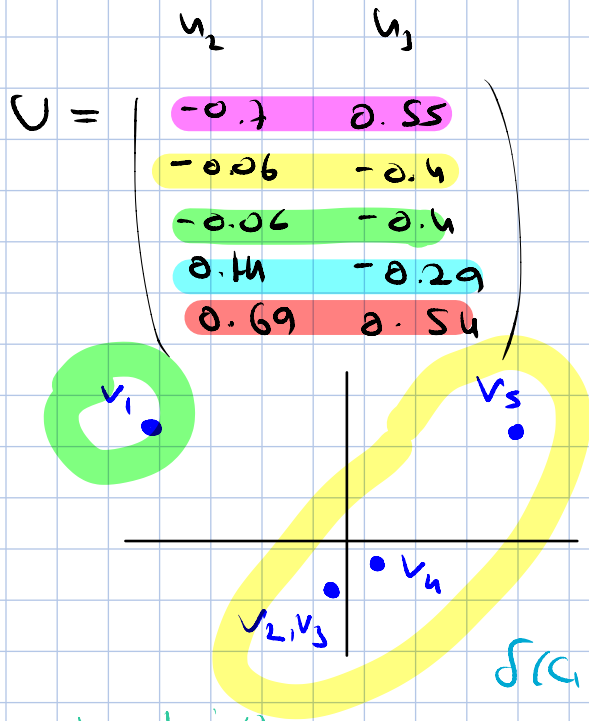
$$FN = 0$$

$$\rightarrow RI = \frac{TP + TN}{TP + TN + FP + FN} = 0.7$$

$$\binom{5}{2} \Rightarrow \binom{5}{2} = 10$$

Internal evaluation:

⊗ need distances → use spectral representation



Distance matrix:

0	1.1452	1.1452	1.1863	1.3959
1.1452	0	0.0000	0.2241	1.2070
1.1452	0.0000	0	0.2241	1.2070
1.1863	0.2241	0.2241	0	1.0034
1.3959	1.2070	1.2070	1.0034	0

→ $\Delta(c_2)$
= max(-)

Dunn INDEX:

Inter-cluster distance: $d(c_1, c_2) = 1.1452$

Intra-cluster distance: $\Delta(c_1) = 0$

$\Delta(c_2) = 1.207$

$$DI = \frac{\min_{i,j} d(c_i, c_j)}{\max_i \Delta(c_i)} = \frac{1.1452}{1.207} = 0.948$$

single-linkage
bimetric