



선형대수학 in 통계

오정현

1 | 6

내적과 공분산

● ● ● ● ● ●

$$x \cdot y = \|x\| \|y\| \cos \theta$$

$$x \cdot y = \sum_{i=1}^n x_i y_i$$

$$\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) \quad \longrightarrow \quad \text{Cov}(x, y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

$$\cos \theta = \frac{x \cdot y}{\|x\| \|y\|} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum (x_i - \bar{x})^2} \sqrt{\sum (y_i - \bar{y})^2}} = \text{Cor}(X, Y)$$

2 | 6

직교 부분공간

• • • • •

$$SSTO = SSE + SSR$$

$$Y^T(I - H_0)Y = Y^T(I - H)Y + Y^T(H - H_0)Y$$

$$\Rightarrow (I - H)(H - H_0) = 0$$

$$\underbrace{(I - H)}_A \perp \underbrace{(H - H_0)}_B$$

$$\left\{ \begin{array}{l} Col(A) = Row(A) \\ Row(A) \perp Null(A) \\ Col(A) \perp Col(B) \end{array} \right. \Rightarrow \begin{array}{l} Null(A) = Col(B) \\ Null(I - H) = Col(H - H_0) \end{array}$$

3 | 6

회귀 계수와 Gram Schmidt

• • • • •

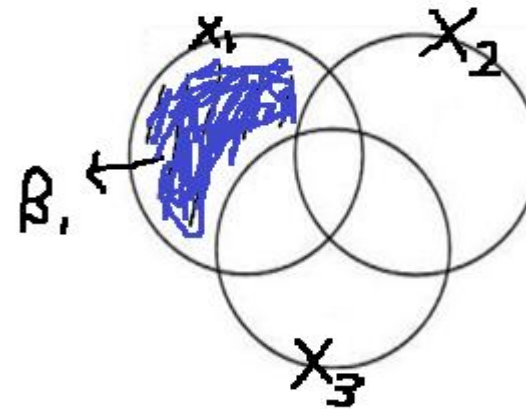
$\beta : ?$

Ex) $e(x_1|1, x_2, x_3)$ is orthogonal to $\text{span}\{1, x_2, x_3\}$.

$$\text{span}\{1, x_1, x_2, x_3\} = \text{span}\{1, x_2, x_3\} \oplus \text{span}\{e(x_1|1, x_2, x_3)\}$$

$$\Rightarrow \hat{y}(1, x_1, x_2, x_3) = \hat{y}(1, x_2, x_3) + \hat{y}(e(x_1|1, x_2, x_3)).$$

$$= \hat{y}(1, x_2, x_3) + \hat{\beta}_1 e(x_1|1, x_2, x_3)$$



4 | 5

고유값과 마르코브 체인



시간에 따른 상태의 변화

미래의 상태는 오직 현재에만 의존

Ex)

$$A = \begin{bmatrix} 0.8 & 0.3 \\ 0.2 & 0.7 \end{bmatrix}$$

$$v = c_1 x_1 + c_2 x_2$$

$$A^k v = c_1 (1)^k x_1 + c_2 \left(\frac{1}{2}\right)^k x_2$$

As k increases, $A^k v$ approaches $c_1 x_1 = \text{steady state}$

5 | 6

Spectral clustering



1. Calculate Laplacian Matrix ($L = D - A$)

2. Spectral decompose L

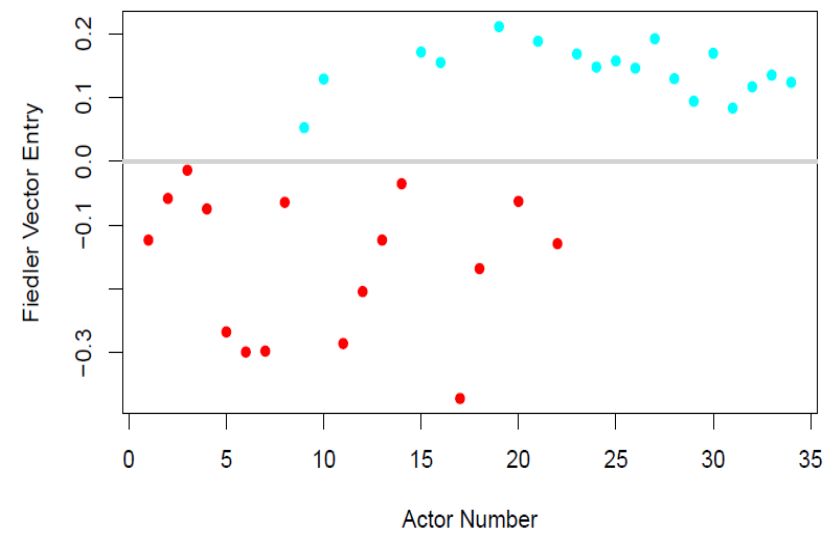
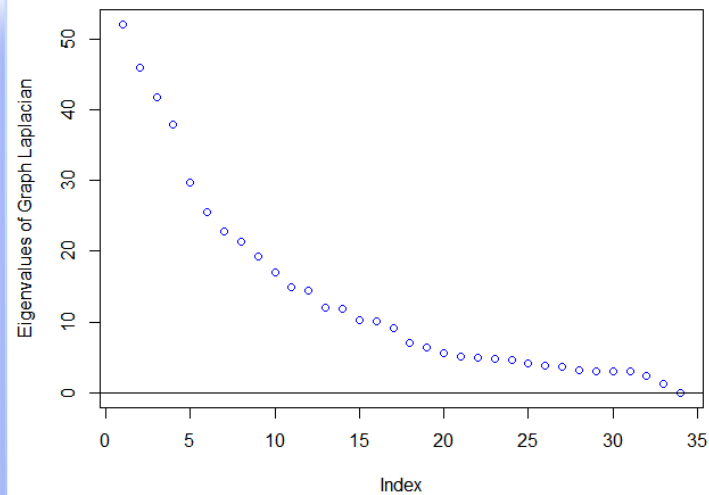
$$(q_1 \dots q_n) \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} \begin{pmatrix} q_1^T \\ \vdots \\ q_n^T \end{pmatrix} (\lambda_1 \geq \dots \geq \lambda_n)$$
$$\lambda_n = 0 \quad \lambda_{n-m} \approx \dots \approx \lambda_{n-1} \approx 0$$

3. Select eigenvectors whose eigenvalue are close to zero

4. Using eigenvectors apply k-means clustering

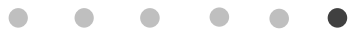
5 | 6

Spectral clustering



6 | 6

판별분석



$$R(x) = \frac{x^T S x}{x^T x} \quad S : \text{symmetric}$$

Maximum of $R(x)$: Largest eigenvalue of S

At eigenvector q_1

$$\rightarrow \frac{x^T S x}{x^T M x} \quad S \ \& \ M : \text{symmetric}$$

Fisher LDA

목표 : 분류를 최대로!!

Maximizes the ratio $\frac{x^T B x}{x^T W x}$

B : between group of sum of square

W : within group of sum of square

—
THANK
YOU
—