PCA

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Biến đổi lại công thức từ trên lớp

$$\mathbf{x} = \left(\begin{array}{c} \dots x_1^T \dots \\ \dots x_2^T \dots \\ \dots x_3^T \dots \\ \dots \\ \dots \\ \dots x_N^T \dots \end{array} \right) \in R^{N.D}$$

$$\mathbf{x} = \begin{pmatrix} \dots x_1^T \dots \\ \dots x_2^T \dots \\ \dots x_3^T \dots \\ \dots \dots \dots \dots \dots \end{pmatrix} \in R^{N.D}$$
 Ma trận hệ số $\mathbf{B} = \begin{pmatrix} b_1 & b_2 & \dots & b_M \\ b_1 & b_2 & \dots & b_M \\ \dots & \dots & \dots & \dots \\ b_1 & b_2 & \dots & b_M \end{pmatrix} \in R^{D.M}$

 $\Rightarrow maxvariance(\mathbf{x}_1^T b_1, x_2^T b_1, ..., x_N^T b_1)$

Asymption mean = 0

$$\mu_z = \frac{x_1^T b_1 + x_2^T b_1 + \dots + x_N^T b_1}{N}$$

$$= \frac{b_1^T \sum_{n=1}^N x_i}{N}$$

$$= b_1^T . \mu_x$$

$$Var_{z} = \frac{\sum_{n=1}^{N} (x_{i}^{T}b_{1} - \mu_{z})^{2}}{N}$$

$$= \frac{\sum_{n=1}^{N} (x_{i}^{T}b_{1})^{2}}{N}$$

$$= \frac{\sum_{n=1}^{N} b_{1}^{T} x_{i} x_{i}^{T} b_{1}}{N}$$

$$= b_{1}^{T} (\frac{\sum_{n=1}^{N} x_{i} x_{i}^{T}}{N}) b_{1}$$

$$= b_{1}^{T} Sb_{1}$$

 $\Rightarrow Maximizeb_1^TSb_1$

$$L(b_1, \lambda) = b_1^T S b_1 - \lambda (b_1^T b_1 - 1)$$

$$\frac{\delta L}{\delta b_1} = 0$$

$$\Leftrightarrow 2Sb_1 - 2\lambda b_1 = 0$$

$$\Leftrightarrow Sb_1 = \lambda b_1$$

$$\frac{\delta L}{\delta \lambda} = 0$$

$$\Leftrightarrow \mathbf{b}_1^T b_1 = 1$$

$$Var_z = b_1^T S b_1$$

$$= b_1^T \lambda b_1$$

$$= \lambda b_1^T b_1$$

$$= \lambda$$