



Suppose that:

$$X = \{x_1, x_2, \dots, x_n\} \quad (x_i \in \mathbb{R}^D)$$

$$X = \begin{bmatrix} \text{---} & x_1^T & \text{---} \\ \text{---} & x_2^T & \text{---} \\ \text{---} & \vdots & \text{---} \\ \text{---} & x_n^T & \text{---} \end{bmatrix} \in \mathbb{R}^{N \times D}$$

weight \leftarrow

$$B = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_N \end{bmatrix} \in \mathbb{R}^{D \times M}$$

$$Z^{N \times M} = X \cdot B$$

$$\Leftrightarrow \begin{bmatrix} z_1^T \\ z_2^T \\ \vdots \\ z_N^T \end{bmatrix} = \begin{bmatrix} x_1^T \\ x_2^T \\ \vdots \\ x_n^T \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_N \end{bmatrix}$$

$$= \begin{bmatrix} x_1^T b_1 & x_1^T b_2 & \dots & x_1^T b_M \\ x_2^T b_1 & x_2^T b_2 & \dots & x_2^T b_M \\ \vdots & \vdots & \ddots & \vdots \\ x_N^T b_1 & x_N^T b_2 & \dots & x_N^T b_M \end{bmatrix}$$

Maximize variance:

$$V_1 = \frac{1}{N} \sum_{n=1}^N z_{1n}^2 \quad (z_{1n}^2 = \lambda_1^T b_{1n})$$

$$V_1 = \frac{1}{N} \sum_{n=1}^N (\lambda_1^T b_{1n})^2 = \frac{1}{N} \sum_{n=1}^N b_{1n}^T \lambda_1 \lambda_1^T b_{1n}$$

$$= b_1^T \left(\frac{1}{N} \sum_{n=1}^N \lambda_{1n} \lambda_{1n}^T \right) b_1 = b_1^T S b_1$$

→ maximize $b_1^T S b_1$ subject to $\|b_1\|^2 = 1$

We have:

$$L = b_1^T S b_1 + \alpha (1 - b_1^T b_1)$$

$$\frac{\partial L}{\partial b_1} = 0 \Leftrightarrow 2Sb_1 - 2\alpha b_1 = 0 \Leftrightarrow Sb_1 = \alpha b_1$$

So we have, α and b_1 : eigenvector
 S : eigenvalue.

$$\frac{\partial L}{\partial \alpha} = 0 \Leftrightarrow b_1^T b_1 = 1$$

$$\Rightarrow \text{Var} = b_1^T S b_1 = b_1^T \alpha b_1 = \alpha b_1^T b_1 = \alpha$$