

对 \mathbf{x} 中的各个成分作线性组合, 得到 $y = \mathbf{w}^T \mathbf{x}$, 这样 n 个样本 $\mathbf{x}_1, \dots, \mathbf{x}_m$ 就产生了 n 个投影结果 y_1, \dots, y_n , 相应的属于集合 Y_0 和 Y_1 , 即 $Y_i = \mathbf{w}^T X_i$ ($i = 0, 1$). 如果 μ_i 为 d 维样本均值为

$$\mu_i = \frac{1}{n_i} \sum_{\mathbf{x} \in X_i} \mathbf{x}, \quad i = 0, 1,$$

定义类间散度矩阵 $S_b = (\mu_0 - \mu_1)(\mu_0 - \mu_1)^T$, 可证明其为对称半正定的. 投影后两类样本均值之差展开为

$$|\tilde{\mu}_0 - \tilde{\mu}_1|^2 = \mathbf{w}^T (\mu_0 - \mu_1)(\mu_0 - \mu_1)^T \mathbf{w} = \mathbf{w}^T S_b \mathbf{w}.$$

原样本空间 X_1 和 X_2 的协方差矩阵

$$C_1 = \frac{1}{n_1} \sum_{\mathbf{x} \in X_0} (\mathbf{x} - \mu_0)(\mathbf{x} - \mu_0)^T$$

$$C_2 = \frac{1}{n_2} \sum_{\mathbf{x} \in X_1} (\mathbf{x} - \mu_1)(\mathbf{x} - \mu_1)^T$$

定义原样本空间 X_1 和 X_2 中的类内散度矩阵

$$\Sigma_0 = \sum_{\mathbf{x} \in X_0} (\mathbf{x} - \mu_0)(\mathbf{x} - \mu_0)^T,$$

$$\Sigma_1 = \sum_{\mathbf{x} \in X_1} (\mathbf{x} - \mu_1)(\mathbf{x} - \mu_1)^T.$$

则总类内散度矩阵 $S_w = \Sigma_0 + \Sigma_1$, 可证明其是对称半正定的.

投影后第 i 内的类内散度为:

$$\begin{aligned} \tilde{s}_i^2 &= \sum_{y \in Y_i} (y - \tilde{\mu}_i)^2 \\ &= \sum_{y \in Y_i} (\mathbf{w}^T \mathbf{x} - \mathbf{w}^T \mu_i)^2 \\ &= \sum_{\mathbf{x} \in X_i} \mathbf{w}^T (\mathbf{x} - \mu_i)(\mathbf{x} - \mu_i)^T \mathbf{w} = \mathbf{w}^T \Sigma_i \mathbf{w}. \end{aligned}$$

故散度矩阵总和可写为 $\tilde{s}_0^2 + \tilde{s}_1^2 = \mathbf{w}^T S_w \mathbf{w}$. 所以

$$J(\mathbf{w}) = \frac{|\tilde{\mu}_0 - \tilde{\mu}_1|^2}{\tilde{s}_0^2 + \tilde{s}_1^2} = \frac{\mathbf{w}^T S_b \mathbf{w}}{\mathbf{w}^T S_w \mathbf{w}}.$$

日期: /

□ m 维空间到一维空间投影轴的最佳方向

$$w^* = S_w^{-1}(\mu_0 - \mu_1) \quad (\text{因 } w \text{ 与大小无关, 只与方向有关})$$

□ $J(w)$ 最大值

$$(\mu_0 - \mu_1)^T S_w^{-1} (\mu_0 - \mu_1)$$

□ 最佳投影变换为

$$y = (\mu_0 - \mu_1)^T S_w^{-1} x$$

由两类二维数据计算线性判别分析(LDA)投影向量

第一类采样数据 ω_1 : $\mathbf{X}_1 = (x_1, x_2) = \{(4,2), (2,4), (2,3), (3,6), (4,4)\}$ (红色点)

第二类采样数据 ω_2 : $\mathbf{X}_2 = (x_1, x_2) = \{(9,10), (6,8), (9,5), (8,7), (10,8)\}$ (蓝色点)

$$\mu_1 = \frac{1}{N_1} \sum_{x \in \omega_1} x = \frac{1}{5} \left[\begin{pmatrix} 4 \\ 2 \end{pmatrix} + \begin{pmatrix} 2 \\ 4 \end{pmatrix} + \begin{pmatrix} 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 3 \\ 6 \end{pmatrix} + \begin{pmatrix} 4 \\ 4 \end{pmatrix} \right] = \begin{pmatrix} 3 \\ 3.8 \end{pmatrix} \quad \text{平均值}$$

$$\mu_2 = \frac{1}{N_2} \sum_{x \in \omega_2} x = \frac{1}{5} \left[\begin{pmatrix} 9 \\ 10 \end{pmatrix} + \begin{pmatrix} 6 \\ 8 \end{pmatrix} + \begin{pmatrix} 9 \\ 5 \end{pmatrix} + \begin{pmatrix} 8 \\ 7 \end{pmatrix} + \begin{pmatrix} 10 \\ 8 \end{pmatrix} \right] = \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix}$$

第一类样本的类内散度矩阵为:

$$S_1 = \sum_{x \in \omega_1} (x - \mu_1)(x - \mu_1)^T = \left[\begin{pmatrix} 4 \\ 2 \end{pmatrix} - \begin{pmatrix} 3 \\ 3.8 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 2 \\ 4 \end{pmatrix} - \begin{pmatrix} 3 \\ 3.8 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 3 \\ 3.8 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 3 \\ 6 \end{pmatrix} - \begin{pmatrix} 3 \\ 3.8 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 4 \\ 4 \end{pmatrix} - \begin{pmatrix} 3 \\ 3.8 \end{pmatrix} \right]^2$$

$\left(\frac{1}{n-1} \right)$

$$= \begin{pmatrix} 1 & -0.25 \\ -0.25 & 2.2 \end{pmatrix}$$

日期:

/

第二类样本的类内散度矩阵为:

$$\begin{aligned} S_2 &= \sum_{x \in \mathcal{O}_2} (x - \mu_2)(x - \mu_2)^T = \left[\begin{pmatrix} 9 \\ 10 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 6 \\ 8 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]^2 \\ &\quad + \left[\begin{pmatrix} 9 \\ 5 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 8 \\ 7 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]^2 + \left[\begin{pmatrix} 10 \\ 8 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]^2 \\ &= \begin{pmatrix} 2.3 & -0.05 \\ -0.05 & 3.3 \end{pmatrix} \rightarrow \begin{matrix} 4.2 & 0.2 \\ -0.2 & 13.2 \end{matrix} \end{aligned}$$

总类内散度矩阵为:

$$\begin{aligned} S_w &= S_1 + S_2 = \begin{pmatrix} 1 & -0.25 \\ -0.25 & 2.2 \end{pmatrix} + \begin{pmatrix} 2.3 & -0.05 \\ -0.05 & 3.3 \end{pmatrix} \\ &= \begin{pmatrix} 3.3 & -0.3 \\ -0.3 & 5.5 \end{pmatrix} \quad \begin{matrix} 13.2 & 1.2 \\ -1.2 & 22 \end{matrix} \end{aligned}$$

类间散度矩阵为:

$$\begin{aligned} S_B &= (\mu_1 - \mu_2)(\mu_1 - \mu_2)^T \\ &= \left[\begin{pmatrix} 3 \\ 3.8 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right] \left[\begin{pmatrix} 3 \\ 3.8 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]^T \\ &= \begin{pmatrix} -5.4 \\ -3.8 \end{pmatrix} \begin{pmatrix} -5.4 & -3.8 \end{pmatrix} \\ &= \begin{pmatrix} 29.16 & 20.52 \\ 20.52 & 14.44 \end{pmatrix} \end{aligned}$$

日期: /

直接计算 w :

$$w^* = S_W^{-1}(\mu_1 - \mu_2) = \begin{pmatrix} 3.3 & -0.3 \\ -0.3 & 5.5 \end{pmatrix}^{-1} \left[\begin{pmatrix} 3 \\ 3.8 \end{pmatrix} - \begin{pmatrix} 8.4 \\ 7.6 \end{pmatrix} \right]$$

$$\begin{pmatrix} 13.2 & 1.2 \\ -1.2 & 22 \end{pmatrix}^{-1} = \begin{pmatrix} 0.3045 & 0.0166 \\ 0.0166 & 0.1827 \end{pmatrix} \begin{pmatrix} -5.4 \\ -3.8 \end{pmatrix} = \begin{pmatrix} 0.9088 \\ 0.4173 \end{pmatrix}$$

$$= \begin{pmatrix} 22 & 1.2 \\ -1.2 & 13.2 \end{pmatrix} \cdot \frac{1}{13.2 \times 22 + 1.2 \times 1.2} \begin{pmatrix} -5.4 \\ -3.8 \end{pmatrix}$$
$$= \begin{pmatrix} -0.4226 \\ -0.1496 \end{pmatrix} \quad \frac{1}{291.84}$$