

特征选择与变换习题

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1.

(1)

$$He = 0:$$

$$He = (I - ee^T)e = Ie - e(e^Te) = e - e = 0$$

$$HH = H:$$

$$HH = (I - ee^T)(I - ee^T) = I - ee^T - ee^T + e(e^Te)e^T = I - ee^T = H$$

$$(x_1 - \bar{x}, \dots, x_n - \bar{x})^T = HX:$$

$$(\bar{x}, \bar{x}, \dots, \bar{x})^T = [\bar{x}(1, 1, \dots, 1)]^T = (X^T ee^T)^T = ee^T X$$

$$(x_1 - \bar{x}, \dots, x_n - \bar{x})^T = X - ee^T X = (I - ee^T)X = HX$$

$$\Sigma = \frac{1}{n} X^T HX:$$

$$\Sigma = \frac{1}{n} (HX)^T (HX) = \frac{1}{n} X^T (H^T H) X = \frac{1}{n} X^T (HH) X = \frac{1}{n} X^T HX$$

(2)

v_i 为 $X^T HX$ 的特征值:

$$X^T HX v_i = \lambda_i v_i$$

$$HXX^T HX v_i = \lambda_i HX v_i$$

$$HXX^T H(HX v_i) = \lambda_i (HX v_i)$$

u_i 为 $HXX^T H$ 的特征值:

$$\begin{aligned}
\mathbf{H}\mathbf{X}\mathbf{X}^T\mathbf{H}\mathbf{u}_i &= l_i\mathbf{u}_i \\
\mathbf{X}^T\mathbf{H}\mathbf{H}\mathbf{X}\mathbf{X}^T\mathbf{H}\mathbf{u}_i &= l_i\mathbf{X}^T\mathbf{H}\mathbf{u}_i \\
\mathbf{X}^T\mathbf{H}\mathbf{X}(\mathbf{X}^T\mathbf{H}\mathbf{u}_i) &= l_i(\mathbf{X}^T\mathbf{H}\mathbf{u}_i)
\end{aligned}$$

又由于：

$$\begin{aligned}
\mathbf{v}_i^T\mathbf{X}^T\mathbf{H}\mathbf{X}\mathbf{v}_i &= (\mathbf{H}\mathbf{X}\mathbf{v}_i)^T(\mathbf{H}\mathbf{X}\mathbf{v}_i) = \lambda_i, & \mathbf{u}_i^T\mathbf{u}_i &= 1 \\
\mathbf{u}_i^T\mathbf{H}\mathbf{X}\mathbf{X}^T\mathbf{H}\mathbf{u}_i &= (\mathbf{X}^T\mathbf{H}\mathbf{u}_i)^T(\mathbf{X}^T\mathbf{H}\mathbf{u}_i) = l_i, & \mathbf{v}_i^T\mathbf{v}_i &= 1
\end{aligned}$$

可以知道对应的比例为 $\sqrt{\lambda_i}$ 和 $\sqrt{l_i}$ ，即 $\mathbf{v}_i = \frac{\mathbf{X}^T\mathbf{H}\mathbf{u}_i}{\sqrt{l_i}}$ ， $\mathbf{u}_i = \frac{\mathbf{H}\mathbf{X}\mathbf{v}_i}{\sqrt{\lambda_i}}$ 。

2.

(1)

由于 \mathbf{u} 为 \mathbf{B} 的特征向量：

$$\begin{aligned}
l_i\mathbf{u}_i &= \mathbf{H}\mathbf{X}\mathbf{X}^T\mathbf{H}\mathbf{u}_i = \mathbf{H}\mathbf{H}\mathbf{X}\mathbf{X}^T\mathbf{H}\mathbf{u}_i = l_i\mathbf{H}\mathbf{u}_i \\
&\Rightarrow \mathbf{u}_i = \mathbf{H}\mathbf{u}_i
\end{aligned}$$

于是有：

$$\mathbf{H}\hat{\mathbf{X}} = \mathbf{H}(\sqrt{l_1}\mathbf{u}_1, \dots, \sqrt{l_n}\mathbf{u}_n) = (\sqrt{l_1}\mathbf{H}\mathbf{u}_1, \dots, \sqrt{l_n}\mathbf{H}\mathbf{u}_n) = (\sqrt{l_1}\mathbf{u}_1, \dots, \sqrt{l_n}\mathbf{u}_n) = \hat{\mathbf{X}}$$

(2)

$$\hat{\mathbf{B}} = \mathbf{H}\hat{\mathbf{X}}\hat{\mathbf{X}}^T\mathbf{H} = \mathbf{H}\hat{\mathbf{X}}\hat{\mathbf{X}}^T\mathbf{H}^T = \mathbf{H}\hat{\mathbf{X}}(\mathbf{H}\hat{\mathbf{X}})^T = \hat{\mathbf{X}}\hat{\mathbf{X}}^T = \mathbf{B}$$

(3)

$$\begin{aligned}
s_{ij} &= (\mathbf{x}_i - \mathbf{x}_j)^T(\mathbf{x}_i - \mathbf{x}_j) \\
&= ((\mathbf{x}_i - \bar{\mathbf{x}}) - (\mathbf{x}_j - \bar{\mathbf{x}}))^T((\mathbf{x}_i - \bar{\mathbf{x}}) - (\mathbf{x}_j - \bar{\mathbf{x}})) \\
&= (\mathbf{x}_i - \bar{\mathbf{x}})^T(\mathbf{x}_i - \bar{\mathbf{x}}) + (\mathbf{x}_j - \bar{\mathbf{x}})^T(\mathbf{x}_j - \bar{\mathbf{x}}) - 2(\mathbf{x}_i - \bar{\mathbf{x}})^T(\mathbf{x}_j - \bar{\mathbf{x}}) \\
&= b_{ii} + b_{jj} - 2b_{ij}
\end{aligned}$$

所以由于 $\hat{\mathbf{B}} = \mathbf{B}$ ，即可以直接得到 $\hat{\mathbf{S}} = \mathbf{S}$ 。

3.

$$\begin{aligned}\Sigma_{\varphi} \boldsymbol{v} &= \lambda \boldsymbol{v} \\ \Rightarrow \frac{1}{n} \sum_{i=1}^n \varphi(\boldsymbol{x}_i) \varphi(\boldsymbol{x}_i)^T \boldsymbol{v} &= \lambda \boldsymbol{v}\end{aligned}$$

当 $\lambda \neq 0$ 时:

$$\boldsymbol{v} = \frac{1}{n\lambda} \sum_{i=1}^n \varphi(\boldsymbol{x}_i) [\varphi(\boldsymbol{x}_i)^T \boldsymbol{v}] = \frac{1}{n\lambda} \sum_{i=1}^n [\varphi(\boldsymbol{x}_i)^T \boldsymbol{v}] \varphi(\boldsymbol{x}_i)$$

其中 $[\varphi(\boldsymbol{x}_i)^T \boldsymbol{v}]$ 为标量, 故可以表示为 $\boldsymbol{v} = \sum_{i=1}^n \alpha_i \varphi(\boldsymbol{x}_i)$ 的形式。