

**Theorem 11.32** For a fixed zero-mean Gaussian random vector  $\mathbf{X}^*$ , let

$$\mathbf{Y} = \mathbf{X}^* + \mathbf{Z},$$

where the joint pdf of  $\mathbf{Z}$  exists and  $\mathbf{Z}$  is independent of  $\mathbf{X}^*$ . Under the constraint that the correlation matrix of  $\mathbf{Z}$  is equal to  $K$ , where  $K$  is any symmetric positive definite matrix,  $I(\mathbf{X}^*; \mathbf{Y})$  is minimized if and only if  $\mathbf{Z} = \mathbf{Z}^* \sim \mathcal{N}(0, K)$ .

**Proof**

1. Since  $E\mathbf{Z}^* = 0$ ,  $\tilde{K}_{\mathbf{Z}^*} = K_{\mathbf{Z}^*} = K$ . Therefore,  $\mathbf{Z}^*$  and  $\mathbf{Z}$  have the same correlation matrix.

2. By noting that  $\mathbf{X}^*$  has zero mean, we apply Lemma 11.33 to see that  $\mathbf{Y}^*$  and  $\mathbf{Y}$  have the same correlation matrix.

3. The theorem is proved by considering

$$\begin{aligned} I(\mathbf{X}^*; \mathbf{Y}^*) - I(\mathbf{X}^*; \mathbf{Y}) &= h(\mathbf{Y}^*) - h(\mathbf{Z}^*) - h(\mathbf{Y}) + h(\mathbf{Z}) \\ &= - \int f_{\mathbf{Y}^*}(\mathbf{y}) \log f_{\mathbf{Y}^*}(\mathbf{y}) d\mathbf{y} + \int f_{\mathbf{Z}^*}(\mathbf{z}) \log f_{\mathbf{Z}^*}(\mathbf{z}) d\mathbf{z} \\ &\quad + \int f_{\mathbf{Y}}(\mathbf{y}) \log f_{\mathbf{Y}}(\mathbf{y}) d\mathbf{y} - \int_{\mathcal{S}_{\mathbf{Z}}} f_{\mathbf{Z}}(\mathbf{z}) \log f_{\mathbf{Z}}(\mathbf{z}) d\mathbf{z} \\ &= - \int f_{\mathbf{Y}}(\mathbf{y}) \log f_{\mathbf{Y}^*}(\mathbf{y}) d\mathbf{y} + \int_{\mathcal{S}_{\mathbf{Z}}} f_{\mathbf{Z}}(\mathbf{z}) \log f_{\mathbf{Z}^*}(\mathbf{z}) d\mathbf{z} \\ &\quad + \int f_{\mathbf{Y}}(\mathbf{y}) \log f_{\mathbf{Y}}(\mathbf{y}) d\mathbf{y} - \int_{\mathcal{S}_{\mathbf{Z}}} f_{\mathbf{Z}}(\mathbf{z}) \log f_{\mathbf{Z}}(\mathbf{z}) d\mathbf{z} \\ &= \int \log \left( \frac{f_{\mathbf{Y}}(\mathbf{y})}{f_{\mathbf{Y}^*}(\mathbf{y})} \right) f_{\mathbf{Y}}(\mathbf{y}) d\mathbf{y} - \int_{\mathcal{S}_{\mathbf{Z}}} \log \left( \frac{f_{\mathbf{Z}}(\mathbf{z})}{f_{\mathbf{Z}^*}(\mathbf{z})} \right) f_{\mathbf{Z}}(\mathbf{z}) d\mathbf{z} \\ &= D(f_{\mathbf{Y}} \| f_{\mathbf{Y}^*}) - D(f_{\mathbf{Z}} \| f_{\mathbf{Z}^*}) \\ &\leq 0. \end{aligned}$$