

Problem 10.1

$$\begin{aligned} \text{r.v.s: } & \left. \begin{aligned} x, x \in \mathbb{R}, x &\sim \mathcal{N}(\mu_x, \sigma_x^2) \\ y, y \in \mathbb{R}, y &\sim \mathcal{N}(\mu_y, \sigma_y^2) \end{aligned} \right\} \Rightarrow x \text{ \& } y = \text{ind.} \\ & z = x + y, z \sim \mathcal{N}(\mu_z, \sigma_z^2) \end{aligned}$$

$$\Rightarrow \mu_z = E[z] = E[x+y] = E[x] + E[y] = \mu_x + \mu_y$$

$$\begin{aligned} \Rightarrow \sigma_z^2 &= E[(z - \mu_z)^2] = E[(x - \mu_x) + (y - \mu_y)]^2 \\ &= E[(x - \mu_x)^2] + 2E[(x - \mu_x)(y - \mu_y)] + E[(y - \mu_y)^2] \\ &= \sigma_x^2 + \underbrace{2(E[x] - \mu_x)(E[y] - \mu_y)}_{\text{b/c } x \text{ \& } y \text{ are ind.}} + \sigma_y^2 \end{aligned}$$

Problem 10.2

→ continuous r.v.s:  $\underline{x} \in \mathbb{R}^n$  and  $\underline{y} \in \mathbb{R}^n$  w/ joint Gaussian Distribution.

$$\begin{bmatrix} \underline{x} \\ \underline{y} \end{bmatrix} \sim \mathcal{N}\left(\begin{bmatrix} \underline{\mu}_x \\ \underline{\mu}_y \end{bmatrix}, \begin{bmatrix} \underline{C}_x & \underline{C}_{xy} \\ \underline{C}_{xy}^T & \underline{C}_y \end{bmatrix}\right)$$

$$\rightarrow \text{let } \underline{c} := \begin{bmatrix} \underline{x} \\ \underline{y} \end{bmatrix}$$

$$\Rightarrow \underline{z} = \underline{x} + \underline{y} = \underline{M} \begin{bmatrix} \underline{x} \\ \underline{y} \end{bmatrix} = \underline{M} \underline{c}, \quad \underline{M} = \begin{bmatrix} \underline{I} & \underline{I} \end{bmatrix}, \quad \underline{I} \in \mathbb{R}^{n \times n} = \text{identity matrix.}$$

$$\Rightarrow \underline{\mu}_z = E[\underline{M} \underline{c}] = \underline{M} E[\underline{c}] = \underline{M} \underline{\mu}_c = \underline{M} \begin{bmatrix} \underline{\mu}_x \\ \underline{\mu}_y \end{bmatrix} = \underline{\mu}_x + \underline{\mu}_y$$

$$\begin{aligned} \Rightarrow \underline{C}_z &= E[(\underline{z} - \underline{\mu}_z)(\underline{z} - \underline{\mu}_z)^T] = E[(\underline{M} \underline{c} - \underline{M} \underline{\mu}_c)(\underline{M} \underline{c} - \underline{M} \underline{\mu}_c)^T] \\ &= E[\underline{M}(\underline{c} - \underline{\mu}_c)(\underline{c} - \underline{\mu}_c)^T \underline{M}^T] = \underline{M} E[(\underline{c} - \underline{\mu}_c)(\underline{c} - \underline{\mu}_c)^T] \underline{M}^T \\ &= \underline{M} \underline{C}_c \underline{M}^T = \underline{M} \begin{bmatrix} \underline{C}_x & \underline{C}_{xy} \\ \underline{C}_{xy}^T & \underline{C}_y \end{bmatrix} \underline{M}^T = \underline{C}_x + \underline{C}_{xy} + \underline{C}_{xy}^T + \underline{C}_y \end{aligned}$$