

• Mean

$$E[x_1] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1 p(x_1, x_2) dx_1 dx_2$$

$$E[x_1] = \int_{-\infty}^{\infty} x_1 \frac{1}{\sqrt{2\pi}\sigma_1} \exp\left(-\frac{(x_1 - \mu_1)^2}{2\sigma_1^2}\right) dx_1$$

$$E[x_1] = \mu_1, \quad E[x_2] = \mu_2$$

$$\mu = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}$$

• Covariance Matrix

$$\Sigma = \begin{bmatrix} \text{Var}(x_1) & \text{Cov}(x_1, x_2) \\ \text{Cov}(x_1, x_2) & \text{Var}(x_2) \end{bmatrix}$$

$$\text{Var}(x_1) = E[(x_1 - \mu_1)^2]$$

$$= \int_{-\infty}^{\infty} (x_1 - \mu_1)^2 p(x_1) dx_1$$

$$= \sigma_1^2$$

$$\text{Var}(x_2) = \sigma_2^2$$

$$\text{Cov}(x_1, x_2) = E[(x_1 - \mu_1)(x_2 - \mu_2)]$$

x_1 and x_2 are independent

$$\text{Cov}(x_1, x_2) = 0$$

$$\Sigma = \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & \sigma_2^2 \end{bmatrix}$$