Proof +

The consider the Birariate Graussian with x_1 and x_2 Handom variables with mean = $\begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}$ and covariance matrix = $\begin{bmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{21} & \sigma_{2}^2 \end{bmatrix}$

with cofficient of correlation blw XI and XZ as $P = \frac{\sigma_{12}}{\sigma_{1}\sigma_{2}} \cdot \text{Here } \sigma_{12} = \sigma_{21}$

I The PDF function of the gaussian in terms XI ander and p

$$f(x_1, x_2) = \frac{-1}{1-p^2} \left[\frac{(x_1 - \mu_1)^2}{\sigma_1^2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} - \frac{2p(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} \right] / 2\pi\sigma_1\sigma_2 \sqrt{1-p^2}$$

Finding the conditional distribution of variable xz when $x_1 = x_1$ $\frac{f(x_1|x_1 = x_1) = \frac{f(x_1,x_2)}{f(x_1)}}{f(x_1)}$

$$f(x_1) = \frac{1}{\sigma_1 \sqrt{2\pi}} \exp \left\{ -\frac{1}{2} \frac{(x_1 - M_1)^2}{\sigma_1^2} \right\}$$

$$f(x_2) = \frac{1}{\sigma_2 \sqrt{2\pi}} \exp \left\{ -\frac{1}{2} \frac{(x_2 - M_2)^2}{\sigma_2^2} \right\}$$
Considering the marginalised pdf as given above

$$\frac{1}{f(x_{2} | x_{1} = x_{1})} = \frac{f(x_{1} | x_{1})}{f(x_{1})}$$

$$= \exp \left\{ \frac{-1}{2(1-p^{2})} \left[\frac{(x_{1} - M_{1})^{2}}{\sigma_{1}^{2}} + \frac{(x_{1} - M_{1})^{2}}{\sigma_{2}^{2}} - \frac{2p(x_{1} - M_{1})(x_{2} - M_{1})}{\sigma_{1}\sigma_{2}} \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^{2})} \left[\frac{(x_{1} - M_{1})^{2}}{\sigma_{1}^{2}} + \frac{(x_{1} - M_{1})^{2}}{\sigma_{2}^{2}} \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^{2})} \left[\frac{(x_{1} - M_{1})^{2}}{\sigma_{1}^{2}} + \frac{(x_{1} - M_{1})^{2}}{\sigma_{2}^{2}} - \frac{2p(x_{1} - M_{1})(x_{1} - M_{2})}{\sigma_{1}\sigma_{2}} \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2} \frac{(x_{1} - M_{1})^{2}}{\sigma_{1}^{2}} + \frac{(x_{1} - M_{1})^{2}}{\sigma_{2}^{2}} \right\}$$

$$= \exp \left\{ \frac{-1}{2} \frac{(x_{1} - M_{1})^{2}}{\sigma_{1}^{2}} \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)} \left[\frac{(\chi_1 - \mu_1)^2}{\sigma_1^2} + \frac{(\chi_2 - \mu_1)^2}{\sigma_1^2} - \frac{2p(\chi_1 - \mu_1)(\chi_2 - \mu_2)}{\sigma_1\sigma_2} \right] + \frac{1}{2} \frac{(\chi_1 - \mu_1)^2}{\sigma_1^2} \right]$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_2^2} \left[\frac{\sigma_2^2}{\sigma_1^2} (\chi_1 - \mu_1)^2 + (\chi_2 - \mu_2)^2 - \frac{2p(\chi_1 - \mu_1)(\chi_2 - \mu_2)\sigma_2}{\sigma_1} - \frac{\sigma_2^2}{\sigma_1^2} (\mu_1 - \mu_1)^2 \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_2^2} \left[p^2 \frac{\sigma_2^2}{\sigma_1^2} (\chi_1 - \mu_1)^2 + (\chi_2 - \mu_2)^2 - 2p \frac{\sigma_2}{\sigma_1} (\chi_1 - \mu_1)(\chi_2 - \mu_2) \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_2^2} \left[p^2 \frac{\sigma_2^2}{\sigma_1^2} (\chi_1 - \mu_1)^2 + (\chi_2 - \mu_2)^2 - 2p \frac{\sigma_2}{\sigma_1} (\chi_1 - \mu_1)(\chi_2 - \mu_2) \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_2^2} \left[p^2 \frac{\sigma_2^2}{\sigma_1^2} (\chi_1 - \mu_1)^2 + \chi^2 - 2\mu_1\chi_2 + \mu_2^2 - 2\chi_2 \left(\frac{p\sigma_2}{\sigma_1} (\chi_1 - \mu_1) \right) + 2\mu_2 \left(\frac{p\sigma_2}{\sigma_1} (\chi_2 - \mu_2) \right) \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_2^2} \left[p^2 \frac{\sigma_2^2}{\sigma_1^2} (\chi_1 - \mu_1)^2 + \chi^2 - 2\mu_1\chi_2 + \mu_2^2 - 2\chi_2 \left(\frac{p\sigma_2}{\sigma_1} (\chi_1 - \mu_1) \right) + 2\mu_2 \left(\frac{p\sigma_2}{\sigma_1} (\chi_2 - \mu_2) \right) \right\} \right\}$$

Using the identity,
$$a^{2}-2ab+b^{2}-2ac+2bc+c^{2}=(a-b-c)^{2}$$

$$=\exp\left\{\frac{-l}{2(1-p^{2})\sigma_{2}^{2}}\left[\begin{array}{ccc}x^{2}-\mu_{2}-p\frac{\sigma_{2}}{\sigma_{1}}(x_{1}-\mu_{1})\end{array}\right]^{2}\right\}$$

$$=\frac{\exp\left\{\frac{-l}{2(1-p^{2})\sigma_{2}^{2}}\left[\begin{array}{ccc}x^{2}-\mu_{2}-p\frac{\sigma_{2}}{\sigma_{1}}(x_{1}-\mu_{1})\end{array}\right]^{2}\right\}}{\sqrt{2\pi}\sigma_{2}}\left[\begin{array}{ccc}xp\left\{\frac{-l}{2\sigma_{2}^{2}}\left[\begin{array}{c}x^{2}-\mu_{2}\end{array}\right]^{2}\right\}\right]}$$

$$\to \text{This resembles the Univariate gaussian pdf with}$$

$$\mu_{x}=\mu_{1}+p\left(\frac{\sigma_{2}}{\sigma_{1}}\right)(x_{1}-\mu_{1})$$
and
$$\sigma_{x}^{2}=(l-p^{2})\sigma_{2}$$

$$f(x \mid (x \mid 2 = x \mid 2)) = \frac{f(x \mid x \mid 2)}{f(x \mid 2)}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)} \left[\frac{(x \mid -\mu \mid)^2}{\sigma_1^2} + \frac{(x \mid 2 - \mu \mid)^2}{\sigma_2^2} - \frac{2p(x \mid -\mu \mid x \mid x \mid 2 - \mu \mid 2)}{\sigma_1 \sigma_2} \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)} \left[\frac{(x \mid -\mu \mid)^2}{\sigma_1^2} + \frac{(x \mid 2 - \mu \mid)^2}{\sigma_2^2} \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_1^2} \left[\frac{\sigma_1^2}{\sigma_2^2} (x \mid 2 - \mu \mid 2)^2 + (x \mid 2 - \mu \mid 2)^2 - \frac{\sigma_1}{\sigma_2^2} (x \mid 2 - \mu \mid 2) \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_1^2} \left[\frac{\sigma_1^2}{\sigma_2^2} (x \mid 2 - \mu \mid 2)^2 + (x \mid 2 - \mu \mid 2)^2 - \frac{\sigma_1}{\sigma_2^2} (x \mid 2 - \mu \mid 2) \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_1^2} \left[\frac{\sigma_1^2}{\sigma_2^2} (x \mid 2 - \mu \mid 2)^2 + (x \mid 2 - \mu \mid 2)^2 - \frac{\sigma_1}{\sigma_2^2} (x \mid 2 - \mu \mid 2) \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^2)\sigma_1^2} \left[\frac{\sigma_1^2}{\sigma_2^2} (x \mid 2 - \mu \mid 2)^2 + (x \mid 2 - \mu \mid 2) \right] \right\}$$

$$= \exp \left\{ \frac{-1}{2(1-p^{2})\sigma_{1}^{2}} \left[\chi_{1} - \mu_{1} - p \frac{\sigma_{1}}{\sigma_{2}} (\chi_{2} - \mu_{2}) \right]^{2} \right.$$

$$- \sqrt{2\pi} \sigma_{1} \sqrt{1-p^{2}}$$

This resembles the univariable garrium with random variable x1 and mean $\mu^* = \mu_1 + \rho \frac{\sigma_1}{\sigma_2} (x_2 - \mu_2)$

and variance = $\sigma_1^2 (1-p^2)$