

Conditional Distribution.

$$\bar{x}, Y. \quad \bar{x} \in \mathbb{R}^n, Y \in \mathbb{R}^m. \quad (\bar{x}, Y) \sim N(\mu, \Sigma).$$

$$\mu = (\mu_1, \mu_2)^T, \mu_1 = E(\bar{x}), \mu_2 = E(Y). \quad \mu \in \mathbb{R}^{n+m}.$$

$$\Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix} \in \mathbb{R}^{(n+m) \times (n+m)} \quad \Sigma_{11} \in \mathbb{R}^{n \times n} = E(\bar{x} - \mu_1)(\bar{x} - \mu_1)^T$$

$$\Sigma_{22} \in \mathbb{R}^{m \times m} = E(Y - \mu_2)(Y - \mu_2)^T$$

$$P_{Y|\bar{x}}(y|x) = \frac{P_{\bar{x}, Y}(x, y)}{P_{\bar{x}}(x)}.$$

$$P_{\bar{x}, Y}(x, y) = C \exp\left(-\frac{1}{2} (x - \mu_1, y - \mu_2)^T \Sigma^{-1} \begin{pmatrix} x - \mu_1 \\ y - \mu_2 \end{pmatrix}\right).$$

$$P_{\bar{x}}(x) = C_1 \exp\left(-\frac{1}{2} (x - \mu_1)^T \Sigma_{11}^{-1} (x - \mu_1)\right).$$

$$\begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}^{-1} \quad \text{Matrix Inversion Formula. "Hole"}$$

$$AH + B = 0 \Rightarrow H = -A^{-1}B, \quad \begin{pmatrix} A & B \\ C & D \end{pmatrix} \begin{pmatrix} I & H \\ 0 & I \end{pmatrix} = \begin{pmatrix} A & AH + B \\ C & CH + D \end{pmatrix}$$

$$\begin{pmatrix} I & 0 \\ H' & I \end{pmatrix} \begin{pmatrix} A & 0 \\ C & D - CA^{-1}B \end{pmatrix} = \begin{pmatrix} A & 0 \\ H'A + C & D - CA^{-1}B \end{pmatrix} = \begin{pmatrix} A & 0 \\ 0 & D - CA^{-1}B \end{pmatrix}$$

$$H'A + C = 0 \Rightarrow H' = -CA^{-1}$$

$$\begin{pmatrix} I & 0 \\ -CA^{-1} & I \end{pmatrix} \begin{pmatrix} A & B \\ C & D \end{pmatrix} \begin{pmatrix} I & -A^{-1}B \\ 0 & I \end{pmatrix} = \begin{pmatrix} A & 0 \\ 0 & D - CA^{-1}B \end{pmatrix} \quad (H_1, H_2, H_3)^{-1}$$

$$= H_3^{-1} H_2^{-1} H_1^{-1}$$

$$\begin{pmatrix} A & B \\ C & D \end{pmatrix}^{-1} = \begin{pmatrix} I & -A^{-1}B \\ 0 & I \end{pmatrix} \begin{pmatrix} A^{-1} & 0 \\ 0 & (D - CA^{-1}B)^{-1} \end{pmatrix} \begin{pmatrix} I & 0 \\ -CA^{-1} & I \end{pmatrix}$$

$$= \begin{pmatrix} A^{-1} & -A^{-1}B(D - CA^{-1}B)^{-1} \\ 0 & (D - CA^{-1}B)^{-1} \end{pmatrix} \begin{pmatrix} I & 0 \\ -CA^{-1} & I \end{pmatrix}$$

$$= \begin{pmatrix} A^{-1} + A^{-1}B(D - CA^{-1}B)^{-1}CA^{-1} & -A^{-1}B(D - CA^{-1}B)^{-1} \\ -(D - CA^{-1}B)^{-1}CA^{-1} & (D - CA^{-1}B)^{-1} \end{pmatrix}$$

$$\begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}^{-1} = \begin{pmatrix} I & -\Sigma_{11}^{-1}\Sigma_{12} \\ 0 & I \end{pmatrix} \begin{pmatrix} \Sigma_{11}^{-1} & 0 \\ 0 & (\Sigma_{22} - \Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12})^{-1} \end{pmatrix} \begin{pmatrix} I & 0 \\ -\Sigma_{21}\Sigma_{11}^{-1} & I \end{pmatrix}$$

$$(x-\mu_1, y-\mu_2)^T \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}^{-1} \begin{pmatrix} x-\mu_1 \\ y-\mu_2 \end{pmatrix}$$

$$= ((x-\mu_1)^T, (y-\mu_2)^T) \begin{pmatrix} I & -\Sigma_{11}^{-1}\Sigma_{12} \\ 0 & I \end{pmatrix} \begin{pmatrix} \Sigma_{11}^{-1} & 0 \\ 0 & (\Sigma_{22}-\Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12}) \end{pmatrix}^{-1} \begin{pmatrix} I & 0 \\ \Sigma_{21}\Sigma_{11}^{-1} & I \end{pmatrix} \begin{pmatrix} x-\mu_1 \\ y-\mu_2 \end{pmatrix}$$

$$= (x-\mu_1)^T \Sigma_{11}^{-1} (x-\mu_1) + \underbrace{(y-\mu_2 - \Sigma_{21}\Sigma_{11}^{-1}(x-\mu_1))^T (\Sigma_{22}-\Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12})^{-1}}$$

$$(y-\mu_{Y|X})^T \Sigma_{Y|X}^{-1} (y-\mu_{Y|X}) = \underbrace{(y-\mu_2 - \Sigma_{21}\Sigma_{11}^{-1}(x-\mu_1))}$$

$$\mu_{Y|X} = \mu_2 + \begin{pmatrix} \sigma_{21} \\ \sigma_{11} \end{pmatrix} (x-\mu_1)$$

$$X \sim N(\mu, \sigma_{11}^2)$$

$$\frac{X-\mu}{\sigma_{11}} \sim N(0, 1)$$

$$E(X) = E(Y) = 0 \quad \alpha X \rightarrow Y \quad \text{Linear Representation}$$

$$\min_{\alpha} E(Y - \alpha X)^2 \Rightarrow \alpha = \frac{E(XY)}{E(X^2)} \quad \text{Projection!}$$

$$\mu_{Y|X} = E(Y|X) = \arg \min_g E(Y - g(X))^2$$

∇g.

$$\begin{aligned} E(Y - g(X))^2 &= E(\underbrace{Y - E(Y|X)}_{\text{VI}} + \underbrace{E(Y|X) - g(X)})^2 \\ &= E(Y - E(Y|X))^2 + E(E(Y|X) - g(X))^2 \\ &\quad + E((Y - E(Y|X))(E(Y|X) - g(X))) \end{aligned}$$

$$\begin{aligned} E(G(X, Y)) &= E(E(G(X, Y)|X)) \\ &= E(h(X)Y|X) \\ &= h(X)E(Y|X) \end{aligned}$$

$$G(X, Y) = (Y - E(Y|X))(E(Y|X) - g(X))$$

$$\mu_{Y|X} = \boxed{\mu_2} + \boxed{\Sigma_{21} \Sigma_{11}^{-1}} (X - \mu_1)$$

$$\Sigma_{Y|X} = \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12}$$

$$\mu_{Y|X} = \mu_2 + \begin{pmatrix} \sigma_{21} \\ \sigma_{11} \end{pmatrix} (X - \mu_1)$$

$$E(X) = E(Y) = 0$$

$$\alpha X \rightarrow Y$$

Linear Representation

$$\min_{\alpha} E(Y - \alpha X)^2 \Rightarrow \alpha_{\text{opt}} = E(XY) / E(X^2) \quad \text{Projection!}$$

∇ Prior

$$Y|X \sim N(\mu_{Y|X}, \Sigma_{Y|X})$$

$$X \sim N(\mu, \sigma_{11}^2)$$

$$\frac{X - \mu}{\sigma_{11}} \sim N(0, 1)$$

$$\begin{pmatrix} I & 0 \\ H' & I \end{pmatrix} \begin{pmatrix} A & 0 \\ C & D - CA^{-1}B \end{pmatrix} = \begin{pmatrix} A & 0 \\ H'A + C & D - CA^{-1}B \end{pmatrix} = \begin{pmatrix} A & 0 \\ 0 & D - CA^{-1}B \end{pmatrix}$$

$$H'A + C = 0 \Rightarrow H' = -CA^{-1}$$

$$\begin{pmatrix} I & 0 \\ -CA^{-1} & I \end{pmatrix} \begin{pmatrix} A & B \\ C & D \end{pmatrix} \begin{pmatrix} I & -A^{-1}B \\ 0 & I \end{pmatrix} = \begin{pmatrix} A & 0 \\ 0 & D - CA^{-1}B \end{pmatrix} \quad (H_1, H_2, H_3)^{-1}$$

$$= H_3^{-1} H_2^{-1} H_1^{-1}$$

$$\begin{pmatrix} A & B \\ C & D \end{pmatrix}^{-1} = \begin{pmatrix} I & -A^{-1}B \\ 0 & I \end{pmatrix} \begin{pmatrix} A^{-1} & 0 \\ 0 & (D - CA^{-1}B)^{-1} \end{pmatrix} \begin{pmatrix} I & 0 \\ -CA^{-1} & I \end{pmatrix}$$

$$E((Y - E(Y|\mathcal{X}))(E(Y|\mathcal{X}) - g(\mathcal{X})))$$

$$= E\left(\frac{E((Y - E(Y|\mathcal{X}))(E(Y|\mathcal{X}) - g(\mathcal{X}))|\mathcal{X}))}{\parallel}\right)$$

$$(E(Y|\mathcal{X}) - g(\mathcal{X})) \frac{E(Y - E(Y|\mathcal{X})|\mathcal{X})}{\parallel}$$

$$E(Y|\mathcal{X}) - E(Y|\mathcal{X})E(1|\mathcal{X}) = 0$$

$$\bar{x} \sim f_{\bar{x}}(x) \quad E(\bar{x}) = \int_{-\infty}^{+\infty} x f_{\bar{x}}(x) dx. \quad Y = G(\bar{x}) \quad \text{"Magic"}$$

$$E(Y) = E(G(\bar{x})) = \int_{-\infty}^{+\infty} y f_Y(y) dy = \int_{-\infty}^{+\infty} G(x) f_{\bar{x}}(x) dx.$$

$$F_Y(y) = P(Y \leq y) = P(G(\bar{x}) \leq y) = P(\bar{x} \leq G^{-1}(y))?$$

$$\begin{aligned} E(\bar{x}) &= \int_{-\infty}^{+\infty} x f_{\bar{x}}(x) dx = \int_{-\infty}^0 F_{\bar{x}}(x) dx + \int_0^{+\infty} (1 - F_{\bar{x}}(x)) dx \\ &= \int_{\Omega} \bar{x}(\omega) P(d\omega) = \sum_k \bar{x}(\omega_k) P(\{\omega_k\}). \end{aligned}$$

$$\begin{aligned} \mu_{Y|\bar{x}} &= \boxed{\mu_2} + \boxed{\bar{\Sigma}_{21} \bar{\Sigma}_{11}^{-1} (x - \mu_1)} = \boxed{\int_{\Omega} G(\bar{x}(\omega)) P(d\omega)} \\ \bar{\Sigma}_{Y|\bar{x}} &= \bar{\Sigma}_{22} - \bar{\Sigma}_{21} \bar{\Sigma}_{11}^{-1} \bar{\Sigma}_{12} \geq 0. = \int_{\Omega} Y(\omega) P(d\omega) \end{aligned}$$

$$\mu_{Y|\bar{x}} = \mu_2 + \begin{pmatrix} \sigma_{21} \\ \sigma_{11} \end{pmatrix} (x - \mu_1).$$

$$\frac{\bar{x} - \mu}{\sigma_{11}} \sim N(0, 1).$$

$$E(\bar{x}) = E(Y) = 0 \quad \alpha \bar{x} \rightarrow Y. \quad \text{Linear Representation}$$

$$\min_{\alpha} E(Y - \alpha \bar{x})^2 \Rightarrow \alpha_{\text{opt}} = \frac{E(\bar{x} Y)}{E(\bar{x}^2)}. \quad \text{Projection!}$$

$$\underline{\Sigma_{Y|X}} = \sigma_{22} - \frac{\sigma_{12}\sigma_{21}}{\sigma_{11}} = \sigma_{22} - \frac{(\sigma_{12})^2}{\sigma_{11}} \geq 0. \quad (E(XY))^2 \leq E X^2 E Y^2$$

$$\sigma_{12}^2 \leq \sigma_{11} \cdot \sigma_{22} \quad (E((X-\mu_1)(Y-\mu_2)))^2 \leq E(X-\mu_1)^2 E(Y-\mu_2)^2$$

$$E(Y) = E(E(Y|X)) \quad \text{Var}(Y) = E(\text{Var}(Y|X)) + \text{Var}(E(Y|X))$$

$$\text{Var}(Y|X) = E((Y - E(Y|X))^2 | X)$$

$$\begin{aligned} \text{Var}(Y) &= E(Y - EY)^2 = E(Y - E(Y|X))^2 + E(E(Y|X) - EY)^2 \\ &= E(\underbrace{E(Y - E(Y|X))^2}_{\text{Var}(Y|X)} | X) + E(E(Y|X) - E(E(Y|X)))^2 \end{aligned}$$

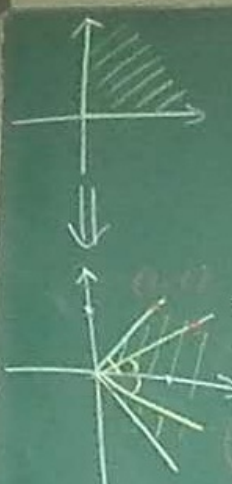
Nonlinear. Operation. Case by Case.

$$(X, Y) \sim N(0, 0, \sigma_1^2, \sigma_2^2, \underline{\rho}). \quad (X, Y) \rightarrow \boxed{G} \rightarrow (X', Y').$$

Hard Limiter. $G(x) = \text{sgn}(x) = \begin{cases} 1 & x \geq 0 \\ -1 & x < 0. \end{cases}$

$$X' \sim \begin{pmatrix} -1 & 1 \\ 1-p & p \end{pmatrix} \quad \mu_1 = 0 \Rightarrow p = \frac{1}{2} \quad Y' \sim \begin{pmatrix} -1 & 1 \\ 1-p & p \end{pmatrix} \quad \mu_1 = \mu_2 = 0$$

$$E(X') = E(Y') = 0 \quad E(X'Y') = 1 \cdot P(XY \geq 0) - 1 \cdot P(XY < 0)$$



$$x' = \frac{x}{\sigma_1} + \frac{y}{\sigma_2}, \quad y' = \frac{y}{\sigma_2} - \frac{x}{\sigma_1}$$

$$x'^2 + y'^2 - 2\rho x'y' = 2u^2 + 2v^2 - 2\rho(u^2 - v^2)$$

$$= 2((1-\rho)u^2 + (1+\rho)v^2)$$

$$= \frac{2}{\pi\sqrt{1-\rho^2}} \iint_{\Delta} \exp\left(-\frac{1}{2(1-\rho^2)}((1-\rho)u^2 + (1+\rho)v^2)\right) du dv$$

$$u' = \frac{u}{\sqrt{1+\rho}}, \quad v' = \frac{v}{\sqrt{1-\rho}}$$

$$= \frac{2}{\pi\sqrt{1-\rho^2}} \iint_{\Delta'} \exp(-(u'^2 + v'^2)) \sqrt{1-\rho^2} du' dv'$$

$\frac{dx dy}{du dv} = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} = -2$

$$E(X'Y') = P(X'Y' \geq 0) - P(X'Y' < 0) = 2P(X'Y' \geq 0) - 1$$

$$P(X'Y' \geq 0) = 2 \int_0^\infty \int_0^\infty f_{X,Y}(x,y) dx dy, \quad \mu_1 = \mu_2 = 0$$

$$= \frac{2}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \int_0^\infty \int_0^\infty \exp\left(-\frac{1}{2(1-\rho^2)}\left(\left(\frac{x-\mu_1}{\sigma_1}\right)^2 + \left(\frac{y-\mu_2}{\sigma_2}\right)^2 - 2\rho\left(\frac{x-\mu_1}{\sigma_1}\right)\left(\frac{y-\mu_2}{\sigma_2}\right)\right)\right) dx dy$$

$$x' = \frac{x-\mu_1}{\sigma_1}, \quad y' = \frac{y-\mu_2}{\sigma_2}$$

$$= \frac{2}{2\pi\sqrt{1-\rho^2}} \int_0^\infty \int_0^\infty \exp\left(-\frac{1}{2(1-\rho^2)}(x'^2 + y'^2 - 2\rho x'y')\right) dx' dy'$$

$$\frac{\pi}{\pi} \iint_{\Delta'} \exp(-(u'^2 + v'^2)) du' dv', \quad u' = r \cos \theta, \quad v' = r \sin \theta.$$

$$= \frac{2}{\pi} \iint \exp(-r^2) r dr d\theta, \quad \phi = \arctan \sqrt{\frac{1+p}{1-p}}$$

$$= \frac{2}{\pi} \left(\int_0^\infty \exp(-r^2) r dr \right) \cdot 2\phi = \frac{2}{\pi} \phi = \frac{2}{\pi} \arctan \sqrt{\frac{1+p}{1-p}}$$

$$\tan(\theta) = t$$

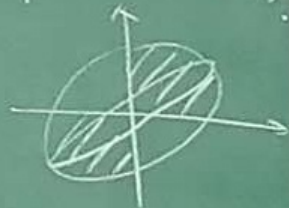
$$= \frac{1}{\pi} \arccos \left(\frac{1 - \frac{1+p}{1-p}}{1 + \frac{1+p}{1-p}} \right) = \frac{1}{\pi} \arccos(-p)$$

$$\Rightarrow \cos(2\theta) = \frac{1-t^2}{1+t^2}$$

$$\arcsin(-p) + \arccos(-p) = \frac{\pi}{2}$$

$$P(XY \geq 0)$$

$$= \frac{1}{\pi} \left(\frac{\pi}{2} - \arcsin(p) \right) = \frac{1}{\pi} \left(\frac{\pi}{2} + \arcsin(p) \right)$$



$$= \frac{1}{2} + \frac{1}{\pi} \arcsin(p). \quad (\text{Arcsin Law})$$

$$E(X'Y') = 2P(XY \geq 0) - 1 = \frac{2}{\pi} \arcsin(p)$$