

# Last lecture

Minimum mean square error estimation ([Ch 4.9](#))

- Recap
  - Constant estimators
  - Unconstrained estimators
  - Linear estimators
- Examples

# Agenda

## Joint Gaussian Distribution ([Ch 4.11](#))

- Motivation
- Facts
- Examples

# **Joint Gaussian distribution**

# Motivation

Describe the joint distribution of (Height, Weight)=( $X, Y$ ) of the class

- Metrics:  $\mu_X, \mu_Y, \sigma_X^2, \sigma_Y^2, \rho_{X,Y}$
- Is  $aX + bY$  Gaussian?

Def. 4.11.1 RV  $X$  and  $Y$  are said to be **jointly Gaussian** if every linear combination  $aX + bY$  is a Gaussian random variable

- $$f_{X,Y}(u, v) = \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho^2}} \exp\left(-\frac{\left(\frac{u-\mu_X}{\sigma_X}\right)^2 + \left(\frac{v-\mu_Y}{\sigma_Y}\right)^2 - 2\rho\left(\frac{u-\mu_X}{\sigma_X}\right)\left(\frac{v-\mu_Y}{\sigma_Y}\right)}{2\sqrt{1-\rho^2}}\right)$$

# Justifying a bivariate normal

A bivariate normal  $f_{X,Y}(u, v) = C \times \exp(-P(u, v))$

- $P(u, v) = au^2 + buv + cv^2 + du + ev + f$
- $P(u, v) \rightarrow \infty$  as  $|u| + |v| \rightarrow \infty$
- $a, c > 0, b^2 - 4ac < 0$

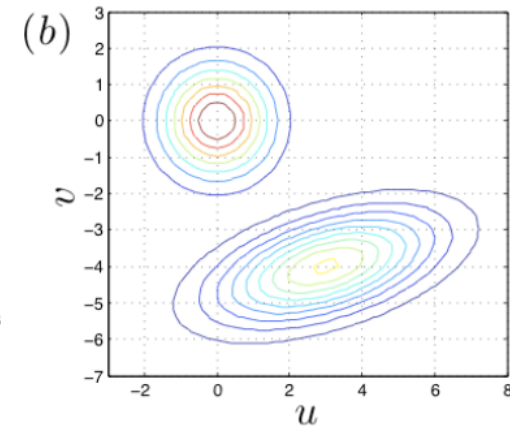
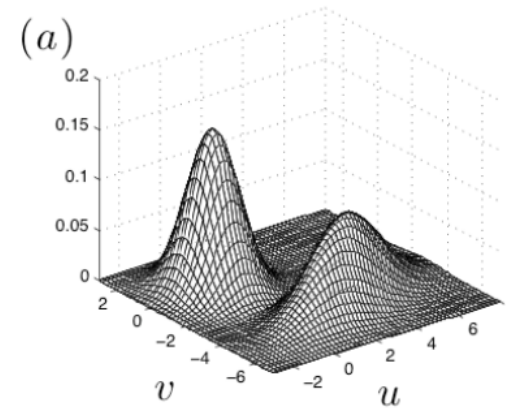
# Standard 2-d Normal to Bivariate Normal

Let  $W, Z$  be independent standard normal

- $$f_{W,Z}(\alpha, \beta) = \frac{e^{-\frac{\alpha^2}{2}}}{\sqrt{2\pi}} \frac{e^{-\frac{\beta^2}{2}}}{\sqrt{2\pi}} = \frac{e^{-\frac{\alpha^2 + \beta^2}{2}}}{2\pi}$$

- $$\begin{pmatrix} X \\ Y \end{pmatrix} = A \begin{pmatrix} W \\ Z \end{pmatrix} + \begin{pmatrix} \mu_X \\ \mu_Y \end{pmatrix}$$

- $$A = \begin{bmatrix} \sqrt{\frac{\sigma_X^2(1+\rho)}{2}} & -\sqrt{\frac{\sigma_X^2(1-\rho)}{2}} \\ \sqrt{\frac{\sigma_Y^2(1+\rho)}{2}} & \sqrt{\frac{\sigma_Y^2(1-\rho)}{2}} \end{bmatrix}$$



# Facts

If  $X$  and  $Y$  forms the bivariate normal with  $\mu_X, \mu_Y, \sigma_X^2, \sigma_Y^2, \rho$

- $X \sim N(\mu_X, \sigma_X^2)$  and  $Y \sim N(\mu_Y, \sigma_Y^2)$
- $aX + bY$  is a Gaussian for any  $a, b \in \mathbb{R}$
- $\rho = \rho_{X,Y}$ ,  $X$  and  $Y$  are independent iif  $\rho = 0$
- $E[Y|X] = \hat{E}[Y|X]$
- $(Y|X = u) \sim N(\hat{E}[Y|X = u], \sigma_e^2)$

# Example

Let  $X$  and  $Y$  be jointly Gaussian with mean  $(0,0)$ .  $\sigma_X^2 = 5$ ,  $\sigma_Y^2 = 2$  and  $Cov(X, Y) = -1$ . Find  $P\{X + 2Y \geq 1\}$  in terms of  $\Phi$  function

- $Z = X + 2Y$  is Gaussian
- $\mu_Z =$
- $Var(Z) =$



# Example

Let  $X$  and  $Y$  be jointly Gaussian RV with mean 0, variance 1, and  $\text{Cov}(X, Y) = \rho$ . Find  $E[Y^2|X]$

- $\hat{E}[Y|X] =$
- For any RV,  $\text{Var}(Z) = E[Z^2] - (E[Z])^2$

# Example

Let  $X$  and  $Y$  be jointly Gaussian RV with mean 0, variance 1, and  $\text{Cov}(X, Y) = 0.5$ . Find

- $\text{Var}(3X - 2Y)$
- $P\{(3X - 2Y)^2 \leq 28\}$  in terms of  $\Phi$
- $E[Y|X = 3]$