

(10)

## Continuous Random Variable III

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Student Notebooks

★ Conditioned PDFs, given another random variable

$$P_{X|Y}(x|y) = P(X=x | Y=y) = \frac{P_{XY}(x, y)}{P_Y(y)}$$

{ This is problematic in }  
 { continuous case }

~~$$P_{X|Y}(x|y) = \frac{P_{XY}(x, y)}{P_Y(y)} \quad \text{if } P_Y(y) > 0$$~~

⇒ Let us consider a condition of  $x < X < x + \delta$   
 given  $y < Y < y + \epsilon$

$$P(x < X < x + \delta | y < Y < y + \epsilon)$$

$$= \frac{f_{XY}(x, y) \delta \epsilon}{f_Y(y) \epsilon} = \frac{f_{XY}(x, y)}{f_Y(y)} \delta$$

⇒ Let us define  $f_{X|Y}(x|y) = \frac{f_{XY}(x, y)}{f_Y(y)}$

$$P(X \in A | Y = y) = \int_A f_{X|Y}(x|y) dx$$



### ★ Total probability and expectation theorem

$$f_X(x) = \int_{-\infty}^{\infty} f_Y(y) f_{X|Y}(x|y) dy$$

$$E[X|Y=y] = \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx$$

$$E[X] = \int_{-\infty}^{\infty} f_Y(y) E[X|Y=y] dy$$

### ★ Independence

$$f_{XY}(x, y) = f_X(x) f_Y(y) \quad \forall x \text{ and } y$$

### ★ Independent Standard normals

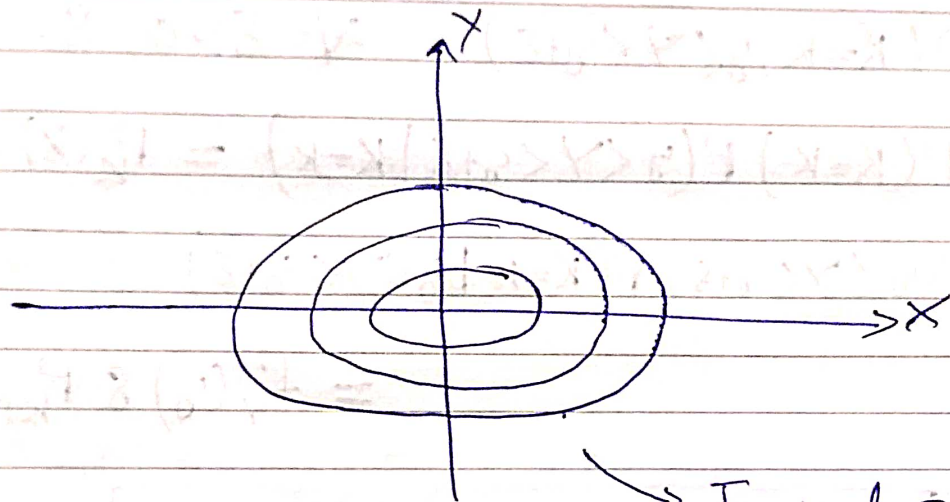
⇒ Let  $X$  and  $Y$  be two independent normal variable distributed normally.

$$f_X = \frac{1}{\sigma_X \sqrt{2\pi}} e^{\left(-\frac{1}{2} \frac{(x-\mu_X)^2}{\sigma_X^2}\right)}$$

$$f_Y = \frac{1}{\sigma_Y \sqrt{2\pi}} e^{\left(-\frac{1}{2} \frac{(y-\mu_Y)^2}{\sigma_Y^2}\right)}$$

$$\text{So, } f_{xy}(x, y) = f_x(x) f_y(y)$$

$$\Rightarrow \frac{1}{2\pi\sigma_x\sigma_y} e^{\left\{ -\frac{(x-\mu_x)^2}{2\sigma_x^2} - \frac{(y-\mu_y)^2}{2\sigma_y^2} \right\}}$$



Typical Contour plot of bivariate distribution.

Note

→ The contours are ellipse when major & minor axis are always aligned along either X or Y axis.

→ It can not make some arbitrary angle.

★ Bayes rule

$$f_{x|y}(x|y) = \frac{f_x(x) f_{y|x}(y|x)}{f_y(y)}$$

{Posterior}

{Prior}



★ The Bayes rule: One discrete and one continuous random variable

$K$ : discrete       $Y$ : continuous

$$\Rightarrow P(K=k, y \leq Y < y+\delta) \quad \forall \delta > 0, \delta \approx 0$$

$$\Rightarrow P(K=k) P(y \leq Y < y+\delta | K=k) = P_K(k) f_{Y|K}(y|k) \delta$$

$$\begin{aligned} \Rightarrow P(y \leq Y < y+\delta) P(K=k | y \leq Y < y+\delta) \\ = f_Y(y) \delta P_{K|Y}(k|y) \end{aligned}$$

$$\Rightarrow \boxed{P_{K|Y}(k|y) = \frac{P_K(k) f_{Y|K}(y|k)}{f_Y(y)}}$$

$$\Rightarrow \boxed{f_{Y|K}(y|k) = \frac{f_Y(y) P_{K|Y}(k|y)}{P_K(k)}}$$

