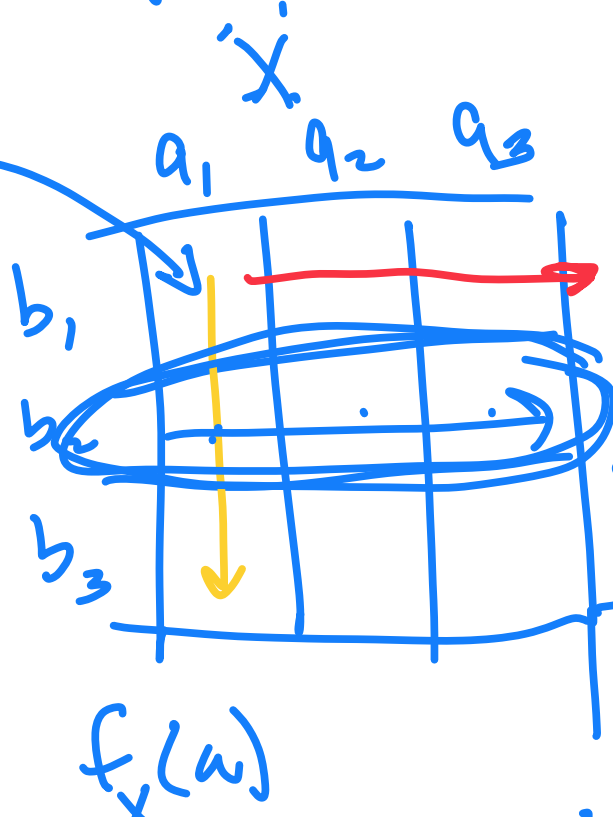


# Discrete Joint RV's

$$P(X=a_i, Y=b_j)$$

$$P(X=a_i | Y=b_j) = \frac{f_{xy}(a_i, b_j)}{f_y(b_j)}$$



$$f_{xy}(a, b)$$

$$f_y(b)$$

$$f_y(b_{2j})$$

$$\sum_i P(X=a_i | Y=b_{2j}) = 1$$

$$f_x(a)$$

$$P(X=a_i)$$

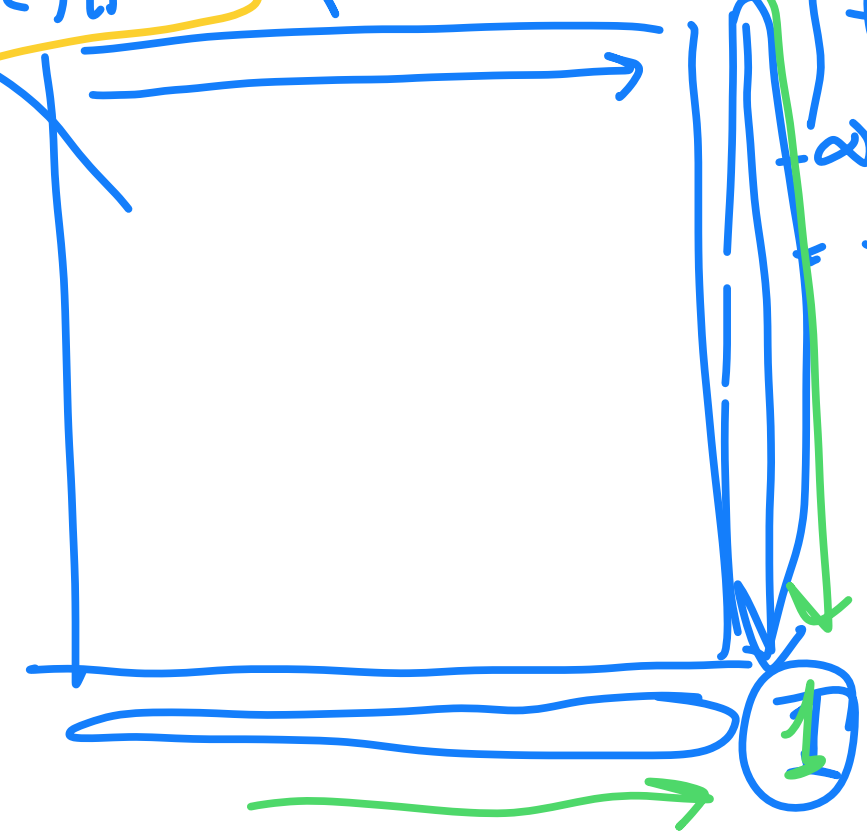
# Continuous Joint RV's

$$f_{xy}(x,y)$$

$$1 = \iint f_{xy}(x,y) dx dy$$

$$\int_{-\infty}^{\infty} f_{xy}(x,y) dx = f_y(y)$$

$P(X=x | Y=y)$   
 $f_{xy}(x,y)$   
 $f_y(y)$



$$f(x, y) = \begin{cases} x^2 + \frac{4}{3}xy + y^2 & 0 \leq x, y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$f_x(x) = \int_0^1 (x^2 + \frac{4}{3}xy + y^2) dy$$

$$f_y(y) = \int_0^1 (x^2 + \frac{4}{3}xy + y^2) dx$$



# Expectation

$$E(X) = \int_a^b \int_c^d x f_{xy}(x, y) dx dy$$

$$= \int_a^b x \left[ \int_c^d f_{xy}(x, y) dy \right] dx$$

$$= \int_a^b x f_x(x) dx$$

# Independence

$f_{XY}(x, y)$

$X, Y$  independent.

$$\frac{f_{XY}(x, y)}{f_Y(y)} = f_X(x) \Rightarrow \underbrace{f_{XY}(x, y)}_{\forall x, y} = f_X(x) f_Y(y)$$

discrete Expectations With Joint RVs.

$$E[g(X, Y)] = \sum_i \sum_j g(a_i, b_j) f_{XY}(a_i, b_j)$$

cont.

$$E[g(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{XY}(x, y) dx dy.$$

- Linearity of  $E$

$$E[rX + sY + t] = rE[X] + sE[Y] + t$$

$$E[X+Y] = E[X] + E[Y]$$

$$X^2 + 2XY + Y^2$$

$$\text{Var}[X+Y] = E[(X+Y)^2] - E[X+Y]^2$$

$$= E[X^2] + 2E[XY] + E[Y^2] - E[X]^2 - 2E[X]E[Y] - E[Y]^2$$

$$= \text{Var}[X] + \text{Var}[Y] + 2[\underbrace{E[XY]}_{\text{Cov}(X,Y)} - \underbrace{E[X]E[Y]}]$$

# Covariance

$$\text{Cov}(X, Y) = E[XY] - E[X]E[Y]$$
$$= E[(X - E[X])(Y - E[Y])]$$

$\sigma^2$  - variance

$\sigma_{XY}$  - cov

mittelw.  
m

abs.  
Kq.

Property:  $X, Y$  independent  $\Rightarrow$   ~~$\sigma_{XY} = 0$~~   $\Rightarrow \text{cov}[X, Y] = 0$



$\text{Var}[X+Y]$        $X, Y$  independent

$$= \text{Var}[X] + \text{Var}[Y]$$

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Correlation

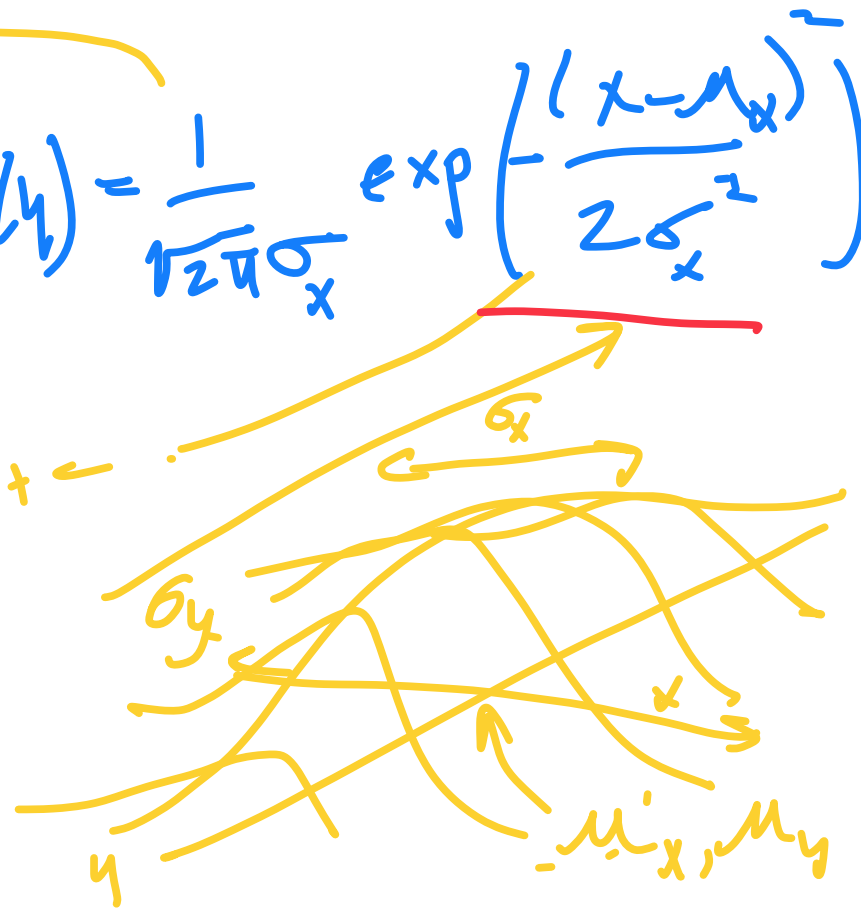
$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}[X] \text{Var}[Y]}} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}$$

# Bivariate Normal Distribution

Independence

$$f_{XY}(x, y) = f(x, y) = f_X(x) f_Y(y) = \frac{1}{\sqrt{2\pi}\sigma_x} \exp\left(-\frac{(x-\mu_x)^2}{2\sigma_x^2}\right)$$

$$\frac{1}{\sqrt{2\pi}\sigma_y} \exp\left(-\frac{(y-\mu_y)^2}{2\sigma_y^2}\right)$$



# Bivariate Normal

Not independent      correlation  $\rho$

$$f(x, y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left(-\frac{1}{2(1-\rho^2)} \times\right.$$

$$\left. \left[ \frac{(x-\mu_x)^2}{2\sigma_x^2} + \frac{(y-\mu_y)^2}{2\sigma_y^2} - \frac{2\rho(x-\mu_x)(y-\mu_y)}{\sigma_x\sigma_y} \right] \right)$$