

$$\frac{E_j}{1}$$

$$P = 0,5$$

$$N = 10$$

$$K = 3$$

$$\binom{10}{3} P^3 (1-P)^{10-3}$$

Binomial medio muestral: $MP = 4$

"

varianza

$$\begin{array}{c} \uparrow \quad \uparrow \\ 10 \quad 0,4 \end{array}$$

$$N P (1-P)$$

.

Ej - 2

$$f_{\underline{X}\underline{Y}} = f_{\underline{X}} f_{\underline{Y}}$$

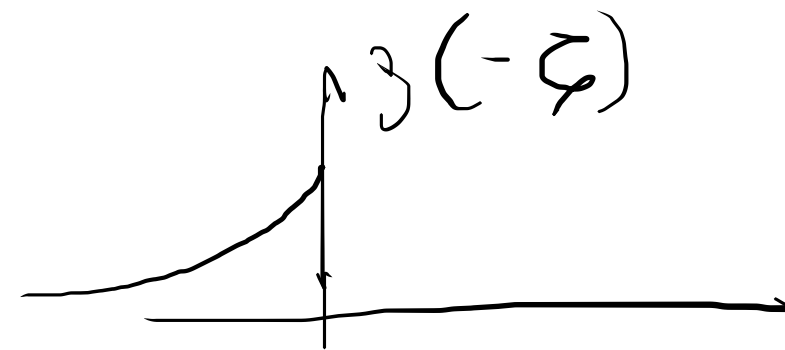
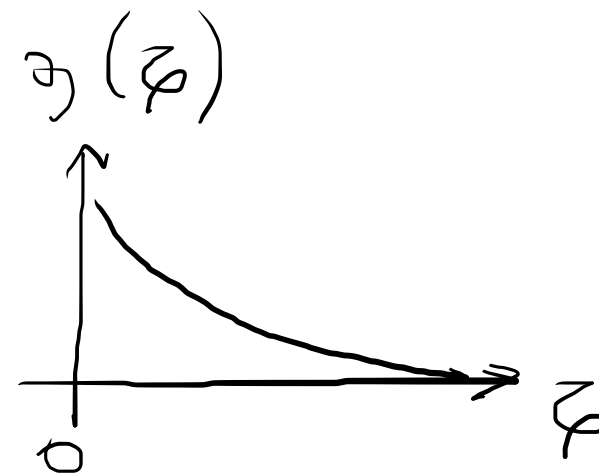
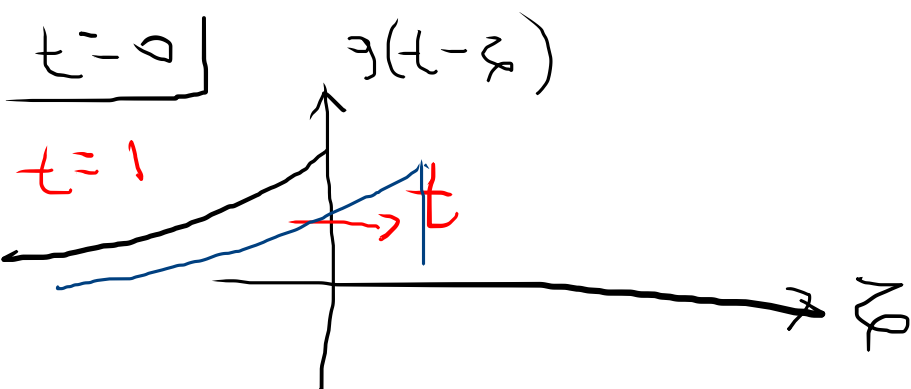
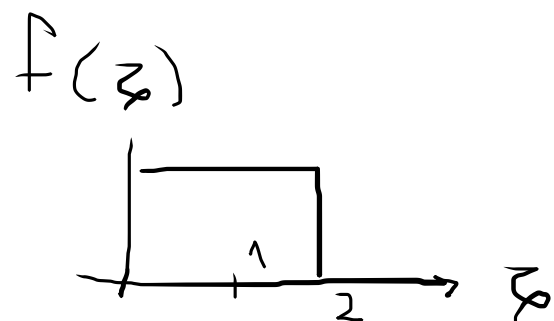
↑
indep.

Producto de Convolución

$f(t), g(t)$ reales

$$(f * g)(t) = \int_{-\infty}^{+\infty} f(\tau) g(t - \tau) d\tau$$

← desplazamiento



Linear $(\lambda f) * g = \lambda (f * g)$; $(f_1 + f_2) * g = f_1 * g + f_2 * g$

Associative $(f * g) * h = f * (g * h)$

Commutative $f * g = g * f$.

$$X = 1\ 2\ 3\ 4$$

$$\text{conv } X, X$$

$m=1$

1	2	3	4
1			

$m=2$

4	3	2	1
4	3	2	1

$m=4$

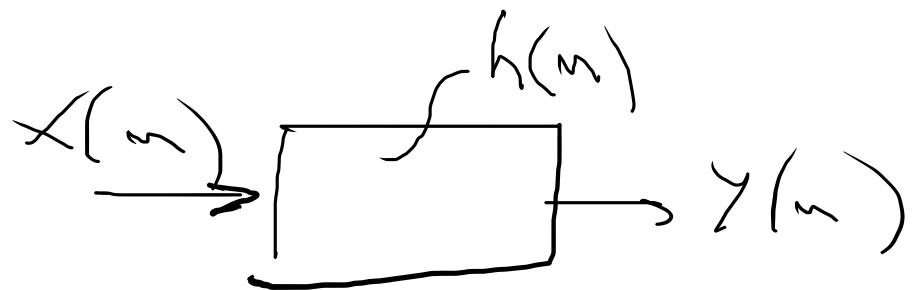
4	3	2	1
---	---	---	---

$$X^e = 4\ 3\ 2\ 1$$

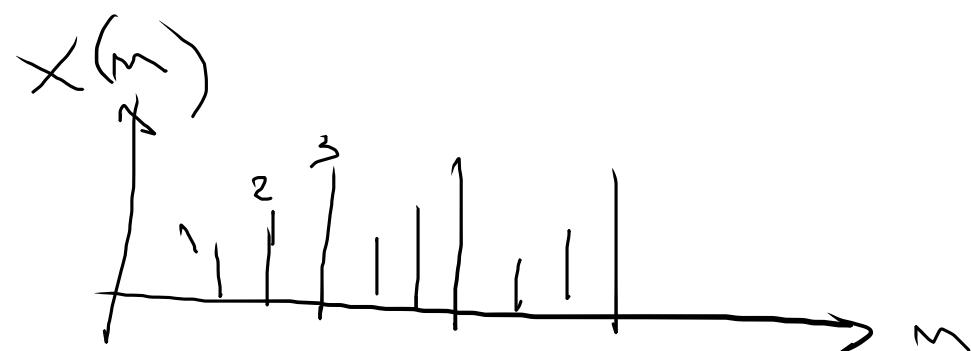
$$X \rightarrow m$$

$$Y \rightarrow m$$

$$X \star Y \rightarrow m + m - 1$$



$$y(n) = x(n] * h(n]$$



$$x(n] * h(n] = ?$$

Promedios móviles.

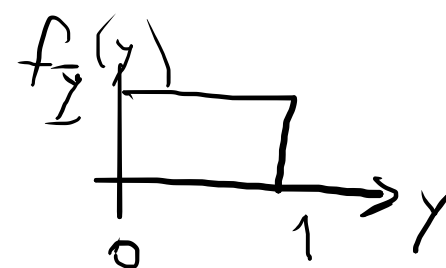
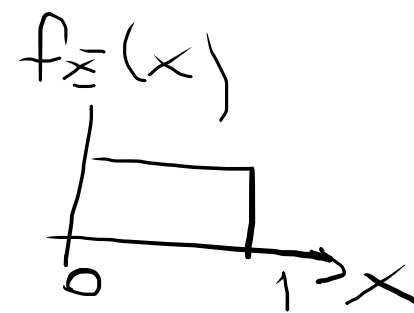
Bolsa.

$$\bar{Z} = \bar{X} + \bar{Y} \quad \bar{X}, \bar{Y} \sim U[0, 1]$$

$$P(\bar{Z} \leq z) = P(\bar{X} + \bar{Y} \leq z) = P(\bar{X} \leq z - \bar{Y}, \bar{Y} \leq \bar{Y})$$

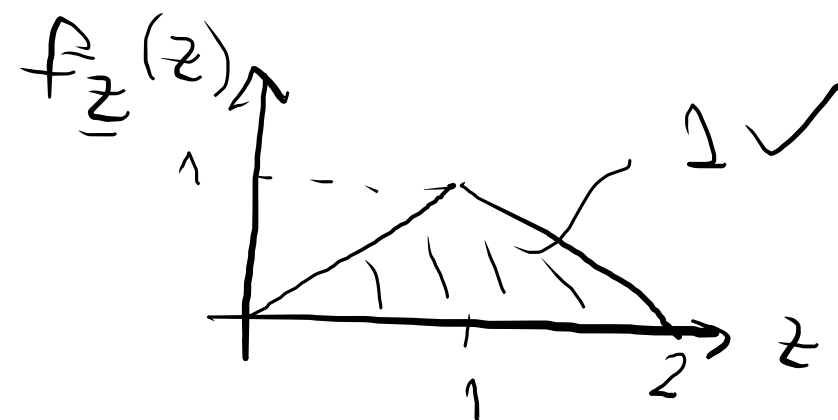
$$\text{Indep.} \rightarrow P(\bar{Z} \leq z) = P(\bar{X} \leq z - \bar{Y}) \cdot P(\bar{Y} \leq \bar{Y})$$

$$f_{\bar{Z}}(z) = \int_{-\infty}^{+\infty} \underbrace{f_{\bar{X}}(z - \gamma)}_{[0, 1] \neq 0 \rightarrow 1} f_{\bar{Y}}(\gamma) d\gamma$$



$[0, 1] \neq 0 \rightarrow 1$

$$\textcircled{1} \quad f_{\bar{Z}}(z) = \int_0^z 1 \cdot 1 d\gamma = z, \quad z \in [0, 1]$$

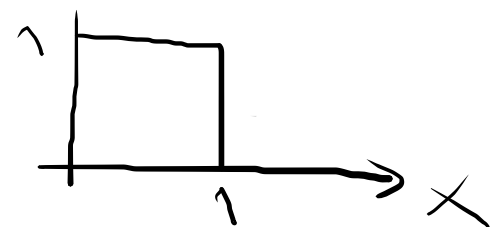


$$\frac{\text{Base} \times \text{altitude}}{2} = 1$$

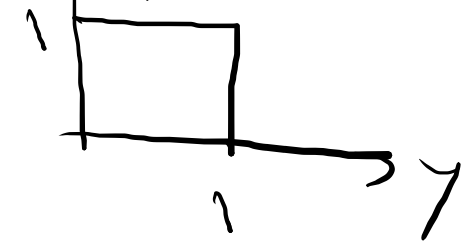
$$\textcircled{2} \quad f_{\bar{Z}}(z) = \int_{z-1}^1 1 \cdot 1 d\gamma = 1 - (z - 1) \quad z \in [1, 2]$$

$$= 2 - z$$

$$f_z(x)$$

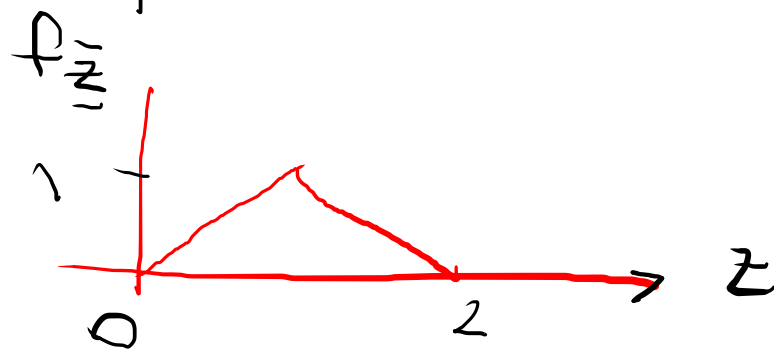
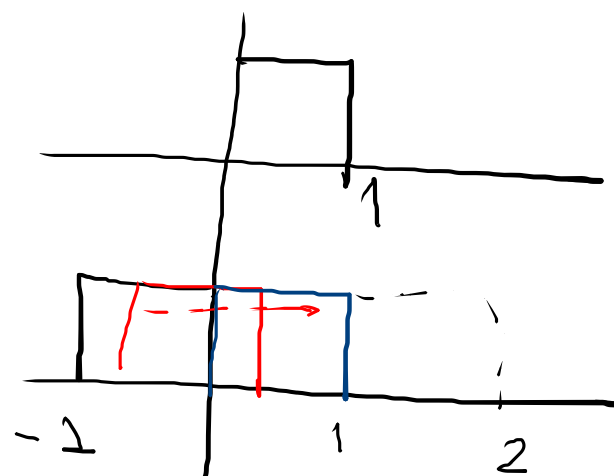


$$f_z(y)$$



$$f_z * f_z$$

$$t=0$$



$$x \sim N(\overset{\mu}{0}, \overset{\sigma^2}{1})$$

$$\bar{z} = \bar{x} + \bar{y}$$

$$y \sim N(0, 1)$$

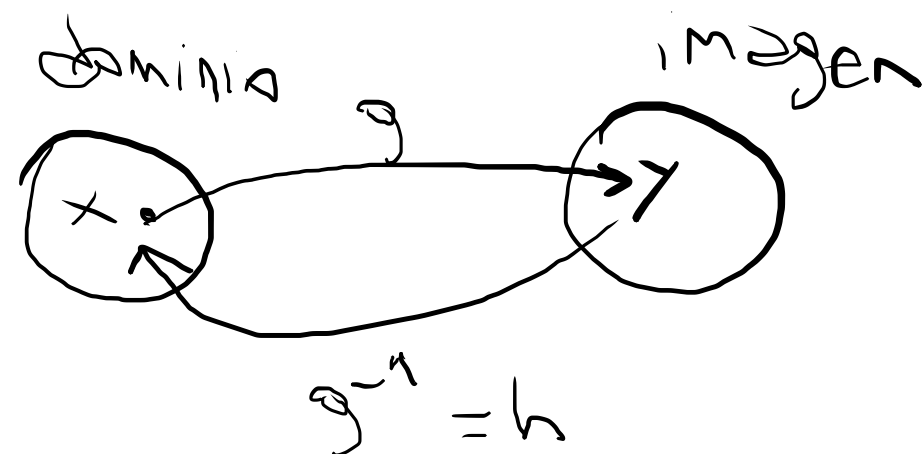
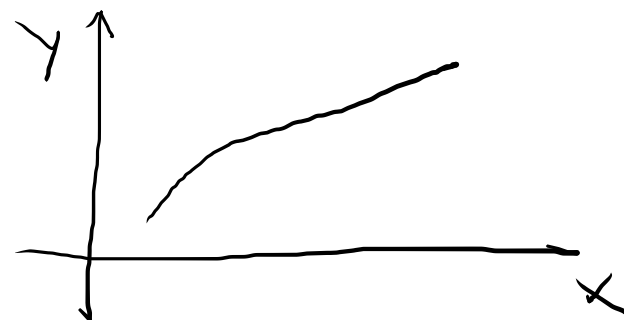
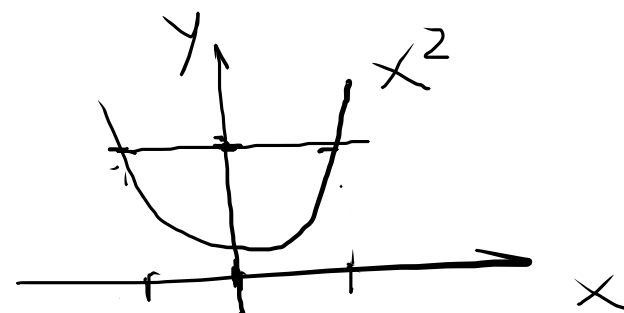
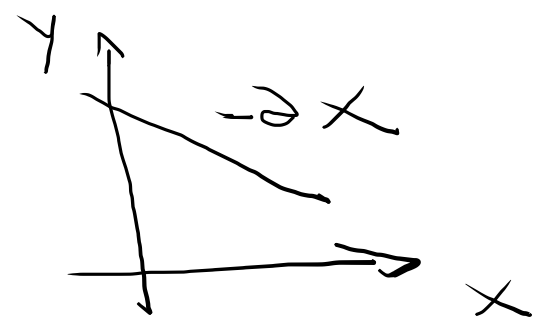
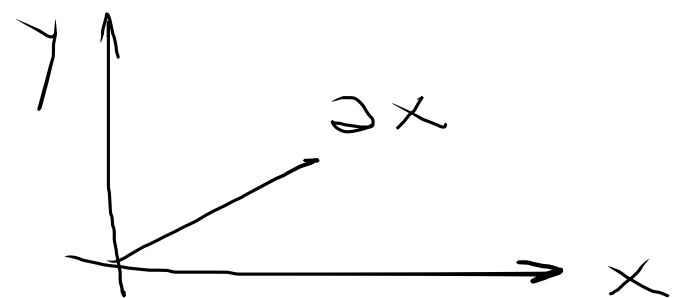
$$E[\bar{z}] = 0 = \mu$$

iid

$$\text{var}[\bar{z}] = \text{var}[\bar{x}] + \text{var}[\bar{y}] = 2 = \sigma^2$$

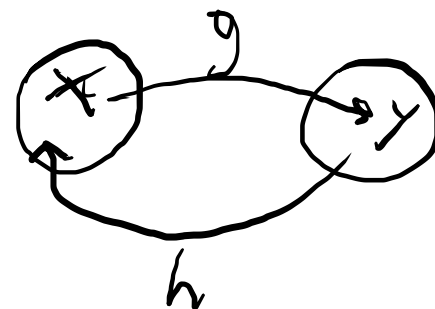
$$\sigma_z = \sqrt{2} = \sqrt{\text{var}[\bar{z}]}$$

$$f_{\bar{z}}(z) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(z-\mu)^2}{2\sigma^2}} = \frac{1}{\sqrt{4\pi}} e^{-\frac{z^2}{4}}$$



$$f_{\bar{y}}(y) = f_{\bar{x}}(h(y)) \left| \frac{dh}{dy}(y) \right|$$

$$\bar{y} = g(\bar{x})$$



$$\bar{y} = g(\bar{x}) \rightarrow \bar{x} = h(\bar{y}) \quad h(x) = g^{-1}(x)$$

$$P(\bar{x} \leq x) = \int_{-\infty}^x f_{\bar{x}}(x) dx$$

$$P(\bar{y} \leq y) = \int_{-\infty}^{h(y)} f_{\bar{x}}(x) dx \leftarrow P(\bar{x} \leq h(y))$$

$$h'(y) = \frac{dh(y)}{dy}$$

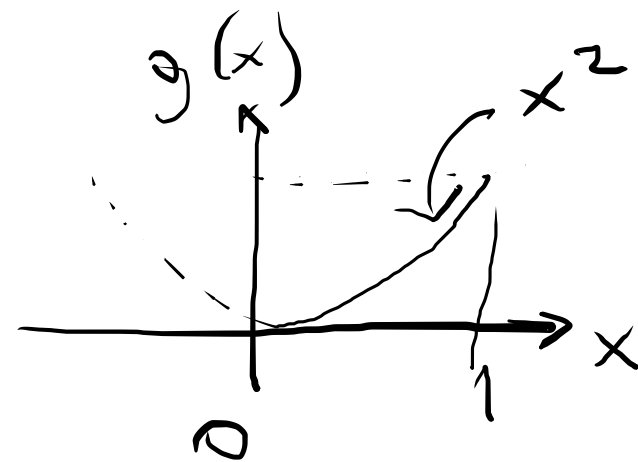
$$\frac{dP}{dy} = f_{\bar{y}}(y) = \frac{d}{dx} \left(P(\bar{x} \leq h(y)) \right) \xrightarrow{\geq 0} \xrightarrow{\geq 0}$$

Regla de la cadena: $f_{\bar{y}}(y) = f_{\bar{x}}(h(y)) |h'(y)|$

$$\bar{x} \sim U[0,1] \rightarrow f_{\bar{x}} = 1, x \in [0,1]$$

$$\bar{y} = \bar{x}^2 \quad g(x) = x^2$$

$$h(x) = \sqrt{x}$$



$$h(y) = \sqrt{y}$$

$$\frac{dh}{dy} = \frac{1}{2\sqrt{y}}, \quad f_{\bar{x}}(\sqrt{y}) = 1$$

\uparrow
[0,1]



$$f_{\bar{y}}(y) = \underbrace{f_{\bar{x}}(h(y))}_1 \cdot \left| \frac{dh}{dy}(y) \right| = \frac{1}{2\sqrt{y}} \quad \int_0^1 \frac{1}{2\sqrt{y}} dy = \sqrt{y} \Big|_0^1 = 1$$

Distribution ~ Gaussienne Bivariable

Matrice de Covariance :

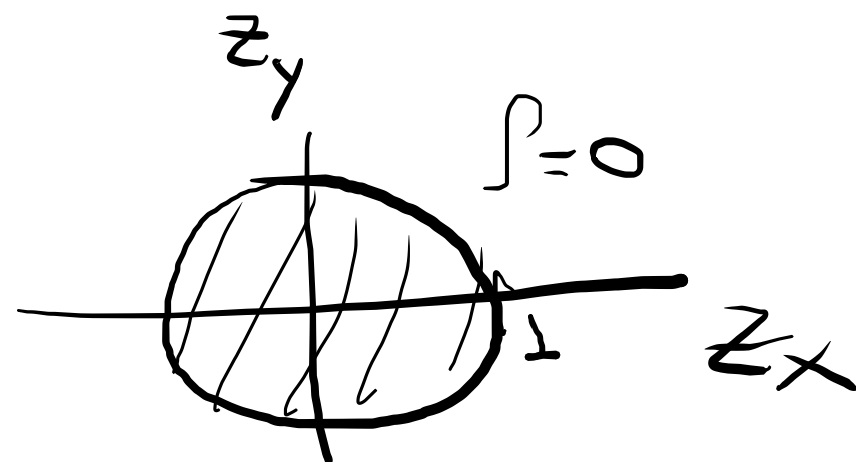
$$\Sigma = \begin{bmatrix} \sigma_x^2 & \rho_{xy} \sigma_x \sigma_y \\ \rho_{xy} \sigma_x \sigma_y & \sigma_y^2 \end{bmatrix}$$

$$|\Sigma| = \sigma_x^2 \sigma_y^2 - \rho_{xy}^2 \sigma_x^2 \sigma_y^2$$

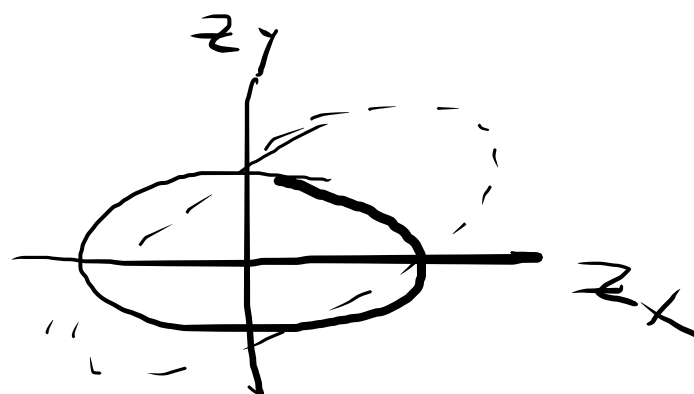
$$|\rho_{xy}| = 1 \Rightarrow \rho_{xy}^2 = 1 \Rightarrow |\Sigma| = 0 \Rightarrow \text{Singular}$$

$$\rho_{xy} = 0$$

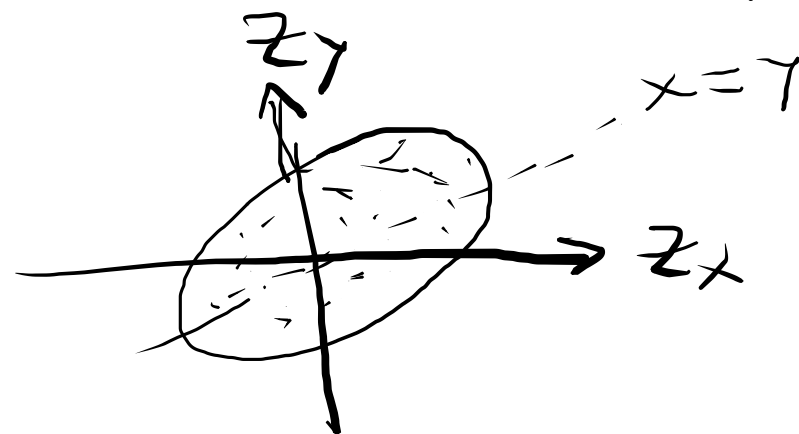
$$z_x^2 + z_y^2 \leq 1$$



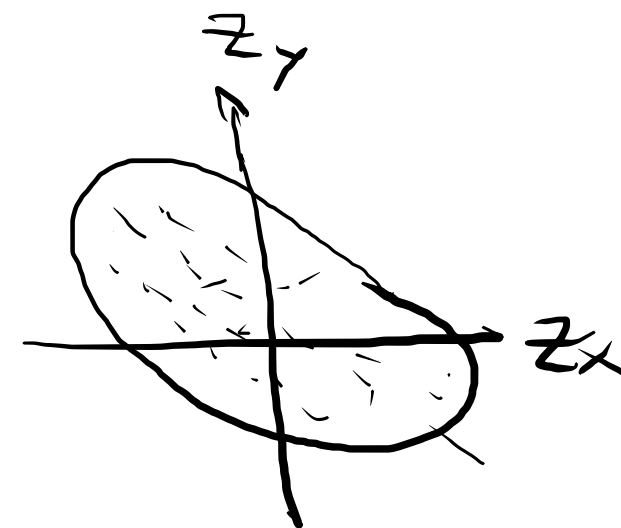
$$z_x^2 - 2\rho_{xy}z_xz_y + z_y^2 \leq r$$



$$\rho > 0$$

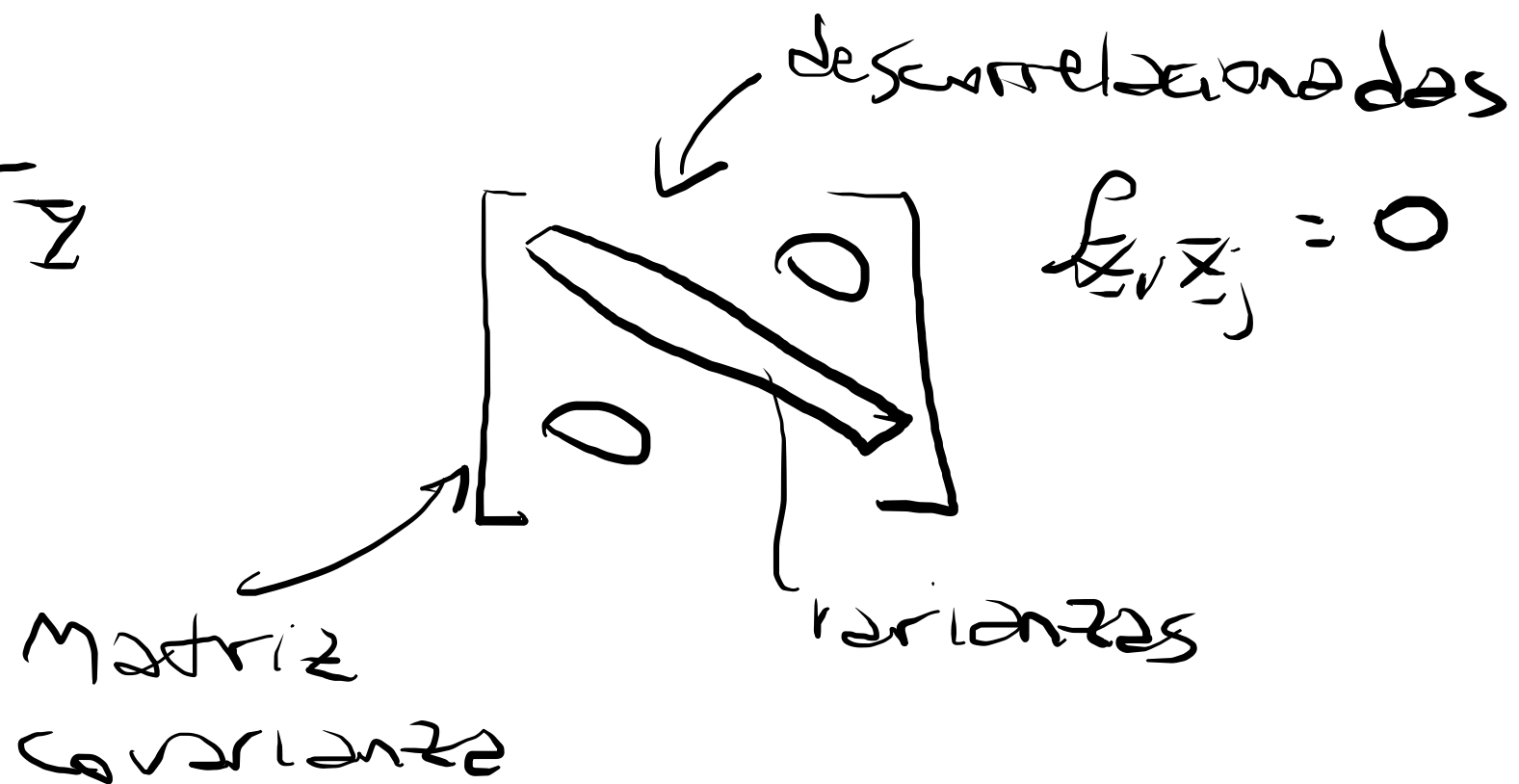


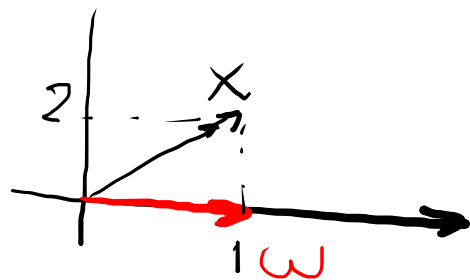
$$\rho < 0$$



$$\text{cov}[\bar{x}, \bar{x}] = \underbrace{\rho_{\bar{x}\bar{x}}}_1 \sigma_{\bar{x}} \sigma_{\bar{x}} = \sigma_{\bar{x}}^2$$

$$\text{cov}[\bar{x}, \bar{y}] = \rho_{\bar{x}\bar{y}} \sigma_{\bar{x}} \sigma_{\bar{y}}$$



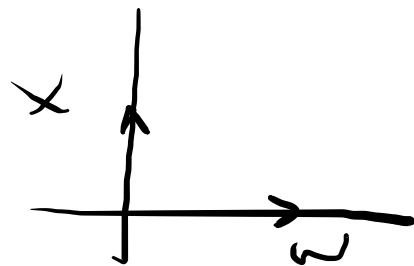


$$w = [1, 0]^T = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$x = [1, 2]^T = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$w^T \cdot x = [1 \ 0] \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 1$$

Normalización de vectores $\rightarrow \frac{w}{\|w\|_2}$



$$z^T x = 0$$

$$\underline{z} = \omega^T \underline{x}$$

$$\mu_z = \omega^T \mu_{\underline{x}}$$

(solo para escalares)

$$\sigma_{\underline{z}}^2 = \omega^2 \sigma_{\underline{x}}^2$$

$$\omega^T \omega = \|\omega\|^2$$

$$\omega^T \Sigma \omega$$