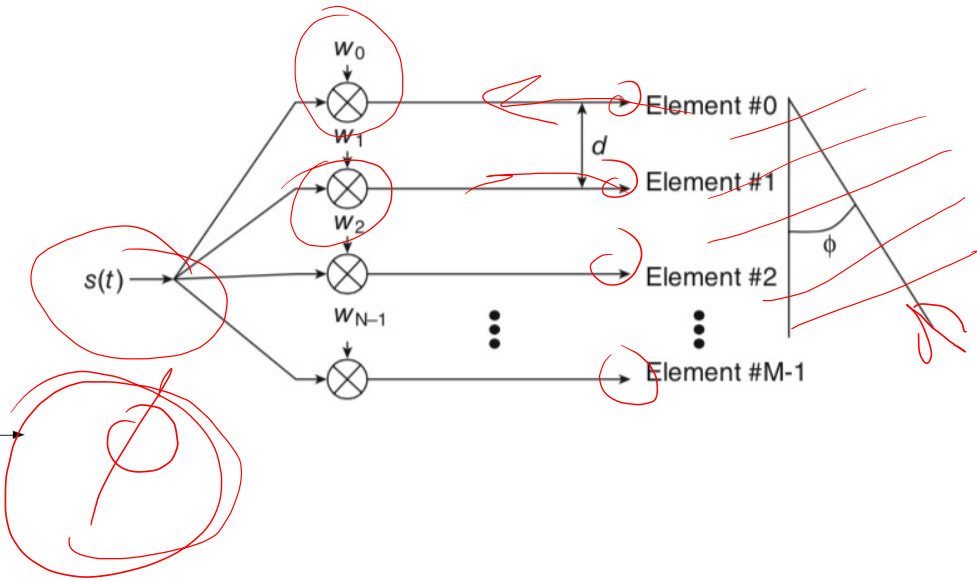
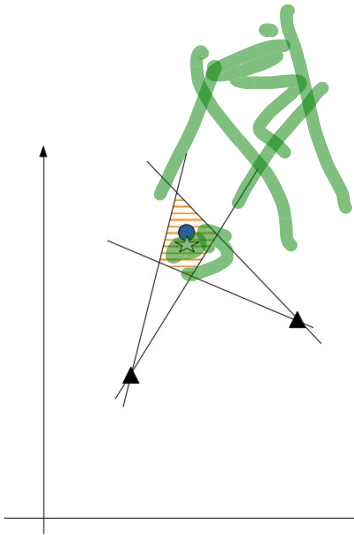
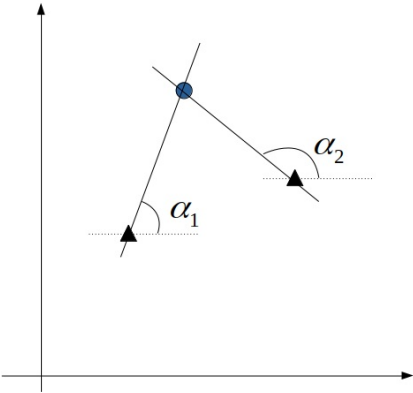
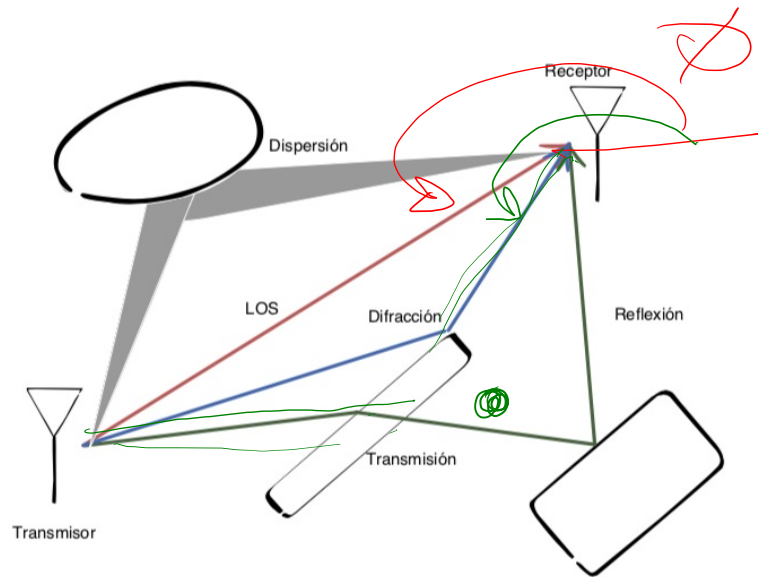


# Multiangulación: DOA/AOA

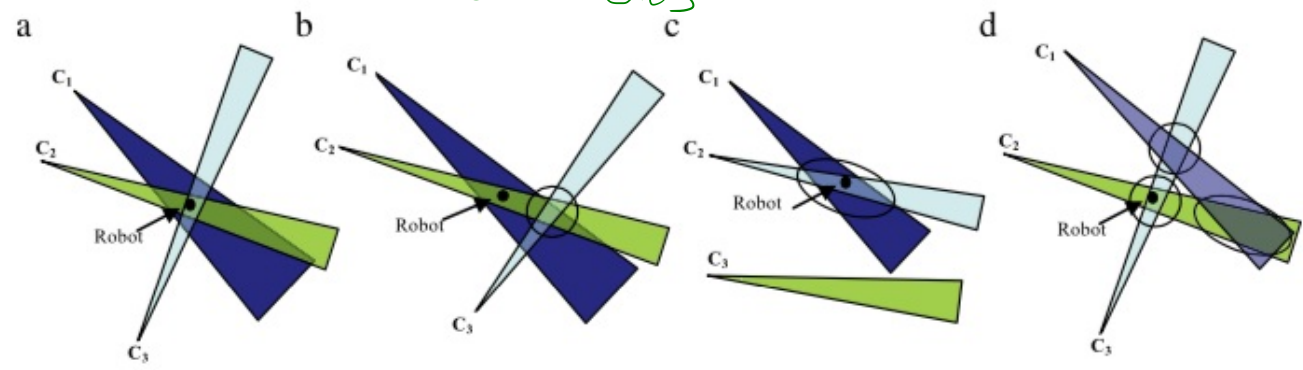


$$\hat{x} = \frac{[(x_2 \tan \theta_2 - x_1 \tan \theta_1) - (y_2 - y_1)]}{(\tan \theta_2 - \tan \theta_1)}$$
$$\hat{y} = \frac{[(x_2 - x_1) \tan \theta_1 \tan \theta_2 + (y_1 \tan \theta_2 - y_2 \tan \theta_1)]}{(\tan \theta_2 - \tan \theta_1)}$$

# Multiangulación: DOA/AOA



OUTLIERS



# Multiangulación: DOA/AOA

$$\mathbf{s}^0 = [x^0, y^0]^T \quad \text{MÓVIL / OBSERVADOR}$$

$$\mathbf{b}_i^0 = [x_i^0, y_i^0]^T \in \mathbb{R}^2 \quad (M)$$

$$\underline{\alpha} = [\alpha_1, \alpha_2, \dots, \alpha_M] = \underline{\alpha}^0 + \underline{n} \rightarrow \text{GAUSIANO}$$

MEAN CERO  
VARIANZA

$$C_{\alpha\alpha} = \begin{bmatrix} \sigma_{\alpha_1}^2 & 0 & 0 \\ 0 & \sigma_{\alpha_2}^2 & 0 \\ 0 & 0 & \sigma_{\alpha_M}^2 \end{bmatrix}$$

$$\underline{b} = [b_1, b_2, \dots, b_M] = \underline{b}^0 + \underline{m}$$

$$\underline{b}^0 = [b_1^0, b_2^0, \dots, b_M^0]$$

$$C_{\psi\psi} = \begin{bmatrix} \psi_{11} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \psi_{MM} \end{bmatrix}$$

## Multiangulación: algoritmo ML (Maximum Likelihood)

$$\Theta = [\Theta_1^T, \Theta_2^T]^T = [\underbrace{\mathbf{s}^{0T}}, \underbrace{\mathbf{b}_i^{0T}}]^T = [x^0, y^0, x_1^0, y_1^0, \dots, x_M^0, y_M^0]^T$$

$$\beta = [\underbrace{\alpha^T}, \underbrace{\mathbf{b}^T}]^T$$

$$\hat{\Theta}_{1,ML} = \underset{s_0}{\operatorname{argmin}} \underbrace{[\alpha - \underbrace{\mathcal{D}_1(\theta_1)}]}^T \underbrace{C_{\alpha\alpha}^{-1}} [\alpha - \mathcal{D}_1(\theta_1)]$$

$$\mathcal{D}_1(\theta_1) = [\mathcal{D}_{1,1}(\theta_1), \dots, \mathcal{D}_{1,n}(\theta_1)]$$

$$\mathcal{D}_{1,i}(\theta) = f_{\theta}^{-1} \cdot \frac{y_0 - y_i}{x_0 - x_i}$$

$$\hat{\theta}_{1,ML} = \underset{\theta_1}{\operatorname{argmin}} [\underbrace{x - g_1(\theta_1)}_{\text{error}}]^T \underbrace{C_{xx}^{-1}}_{\text{covariance}} [\underbrace{x - g_1(\theta_1)}_{\text{error}}] \quad \downarrow$$

$$\hat{\theta}_{1,k+1} = \hat{\theta}_{1,k} + \underbrace{\left( H_{1k}^T C_{xx} H_{1k} \right)^{-1}}_{\text{covariance}} \underbrace{H_{1k}^T C_{xx} (x - g_1(\theta_{1,k}))}_{\text{error}}$$

$$g_1(\theta_{1,k}) = \begin{bmatrix} 1 & \frac{y_k - y_1}{x_k - x_1} \\ g & \frac{y_k - y_m}{x_k - x_m} \end{bmatrix}$$

error

$$H_{1k} = \frac{\partial g_1 / \partial \theta_1}{\partial g_1 / \partial \theta_{1,k}}$$

$$\begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Multiangulación: algoritmo LS (Least Square)

$$\alpha_i^0 = \tan^{-1} \frac{y^0 - y_i^0}{x^0 - x_i^0}$$

$$\tan \alpha_i^0 = \frac{y^0 - y_i^0}{x^0 - x_i^0} = \frac{\sin \alpha_i^0}{\cos \alpha_i^0}$$

$$(y^0 - y_i^0) \cos \alpha_i^0 = (x^0 - x_i^0) \sin \alpha_i^0$$

$$x_i^0 \sin \alpha_i^0 - y_i^0 \cos \alpha_i^0 = x^0 \sin \alpha_i^0 - y^0 \cos \alpha_i^0$$

$$H = G \cdot S$$

$$\begin{bmatrix} x_1^0 \sin \alpha_1^0 - y_1^0 \cos \alpha_1^0 \\ \vdots \\ x_m^0 \sin \alpha_m^0 - y_m^0 \cos \alpha_m^0 \end{bmatrix} = \begin{bmatrix} \sin \alpha_1^0 & -\cos \alpha_1^0 \\ \vdots & \vdots \\ \sin \alpha_m^0 & -\cos \alpha_m^0 \end{bmatrix} \begin{bmatrix} x^0 \\ y^0 \end{bmatrix}$$

$$H = G \cdot S$$

$$\begin{bmatrix} x_1^0 \sin \alpha_1^0 - y_1^0 \cos \alpha_1^0 \\ \vdots \\ x_m^0 \sin \alpha_m^0 - y_m^0 \cos \alpha_m^0 \end{bmatrix} = \begin{bmatrix} \sin \alpha_1^0 & -\cos \alpha_1^0 \\ \vdots & \vdots \\ \sin \alpha_m^0 & -\cos \alpha_m^0 \end{bmatrix} \begin{bmatrix} x^0 \\ y^0 \end{bmatrix}$$

$$\hat{S}_L = (G^T G)^{-1} G^T H$$

$$\begin{matrix} 6 & 4 \\ M \times 2 & M \times 1 \end{matrix}$$

$$2 \times M \times M \times 1$$

$$(2 \times 1)$$

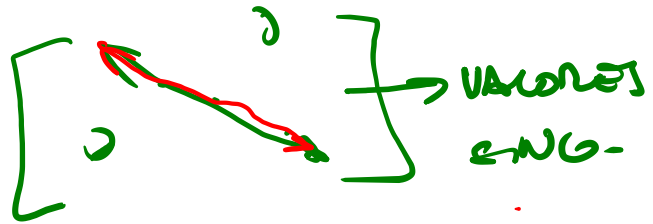
Multiangulación: algoritmo TLS (Total Least Square)

$$\hat{\Theta}_{1,TLS} = (\mathbf{G}^T \mathbf{G} - \sigma_s^2 \mathbf{I})^{-1} \mathbf{G}^T \mathbf{h}$$

①  $\sigma_s \rightarrow$  MENOR VALOR SINGULAR DE  $\begin{bmatrix} \mathbf{G}^T & \mathbf{h} \end{bmatrix}$

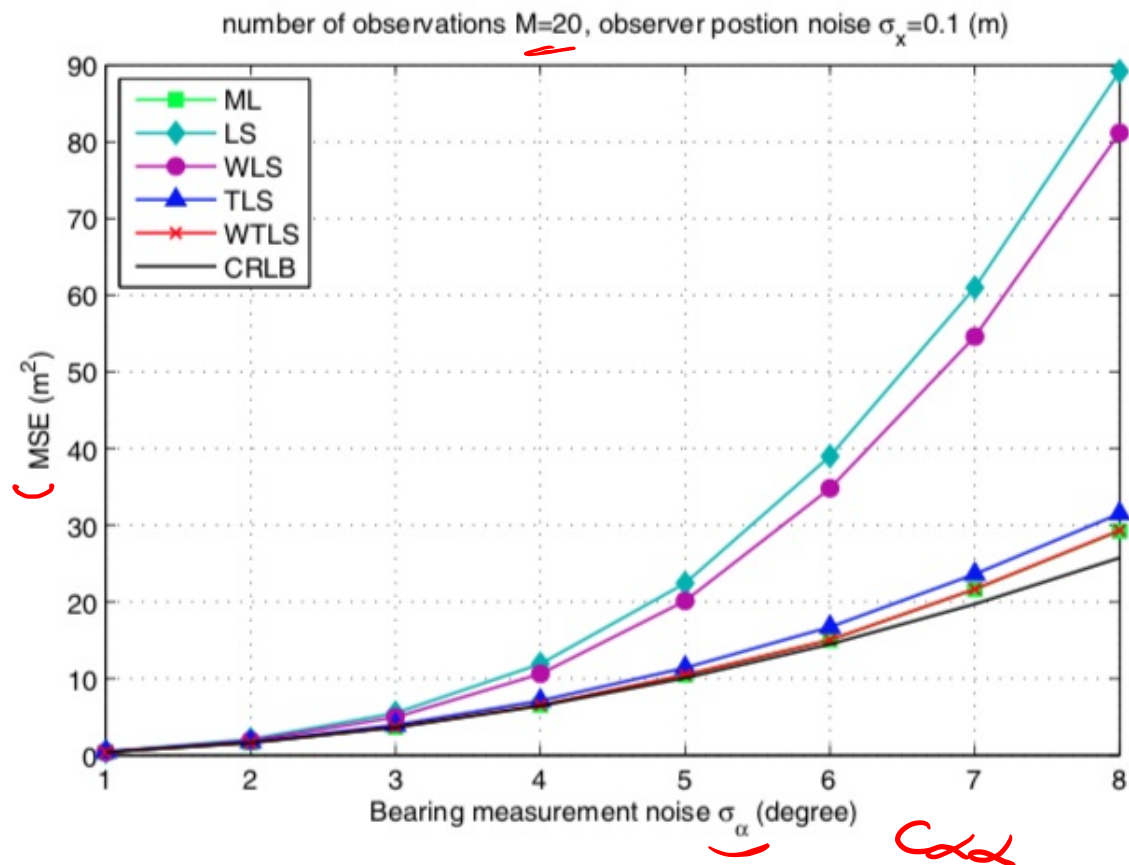
$\mathbf{A}_{MN} = \mathbf{U} \underbrace{\left( \sum \right)}_{\text{VALORES SING.}} \mathbf{V}^T$   $\xrightarrow{NN}$

$\mathbf{A}_{NN} = \begin{matrix} \nearrow \text{AUTOV.} \\ \searrow \text{AUTOV.} \end{matrix}$





# Multiangulación: comparación



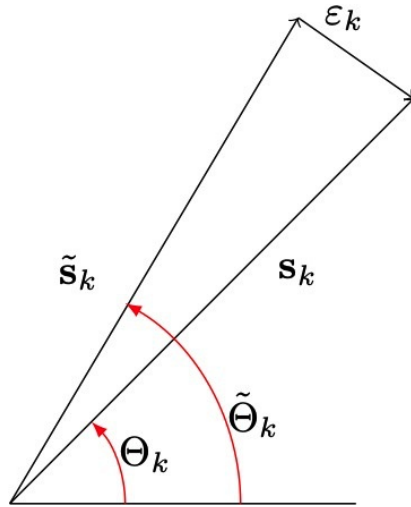
Multiangulación: Estimador Bayesiano pseudo-lineal

$$p(\mathbf{s}|\boldsymbol{\alpha}) = \frac{p(\boldsymbol{\alpha}|\mathbf{s}) \cdot p(\mathbf{s}_0)}{p(\boldsymbol{\alpha})}$$

$$p(\mathbf{s}|\boldsymbol{\alpha}) = \frac{p(\boldsymbol{\alpha}|\mathbf{s}) \cdot p(\mathbf{s}_0)}{p(\boldsymbol{\alpha})}$$

⌋

# Multiangulación: Estimador Bayesiano pseudo-lineal



Multiangulación: Estimador Bayesiano pseudo-lineal

$$\hat{\mathbf{s}}_{BP\text{LE}} = \mathbf{s}_{mean} + (\mathbf{C}_{ss}^{-1} + \mathbf{G}^T \mathbf{C}_{\eta\eta}^{-1} \mathbf{G})^{-1} \mathbf{G}^T \mathbf{C}_{\eta\eta}^{-1} (\mathbf{h} - \mathbf{G} \cdot \mathbf{s}_{mean})$$