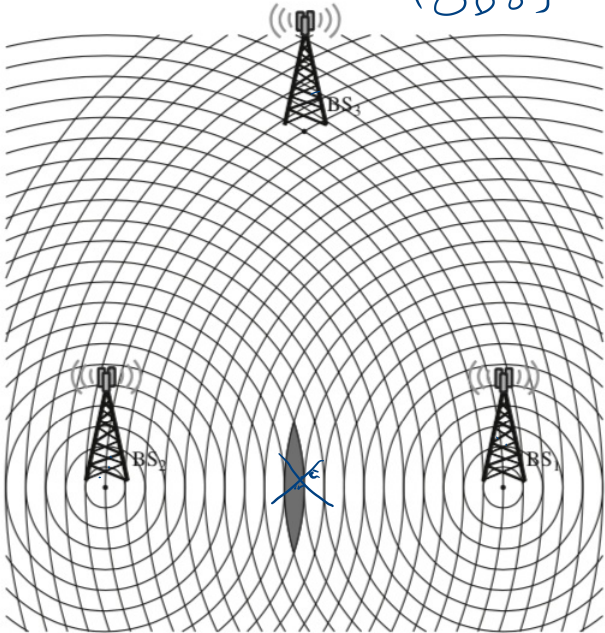
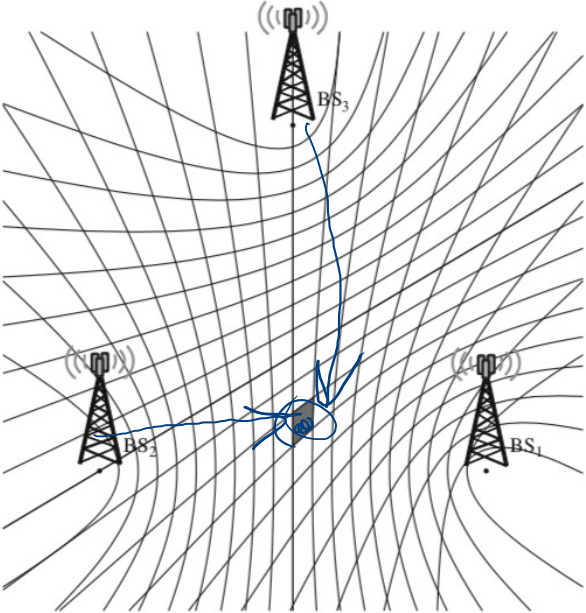


Multilateración Hiperbólica

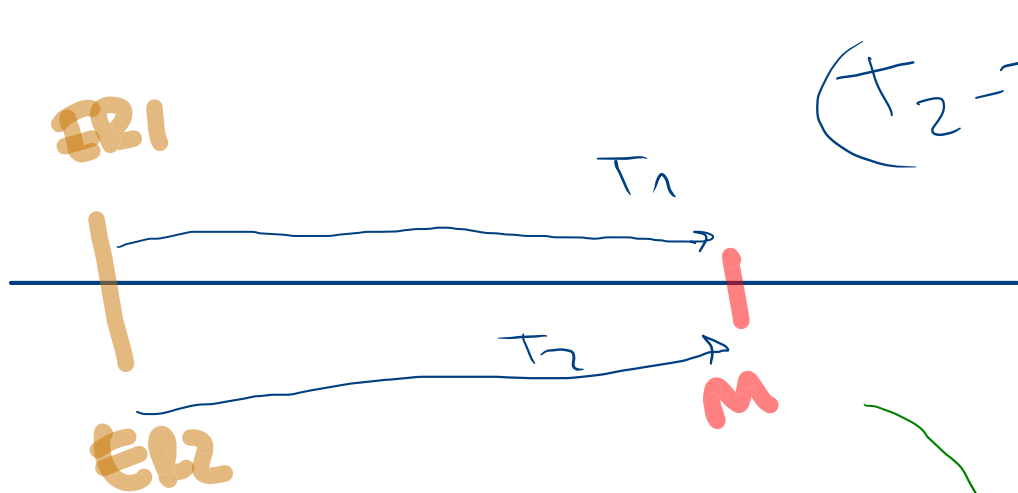
Sincronización  
+000



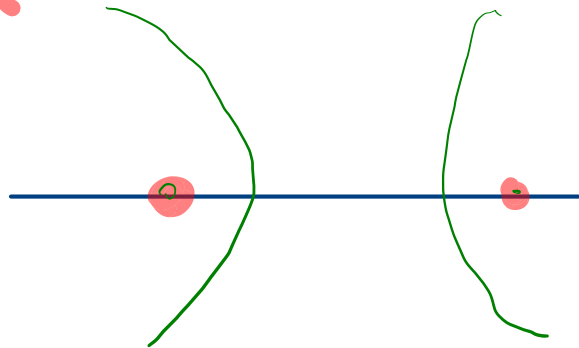
Sincronizar Low  
CAS ER.



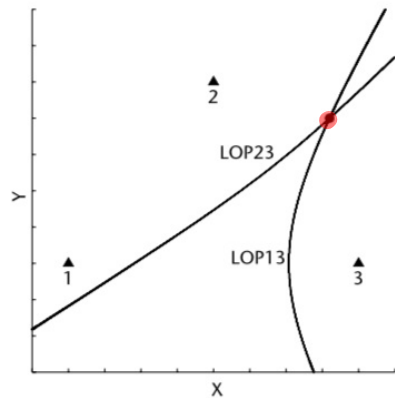
$\Delta_{12}$



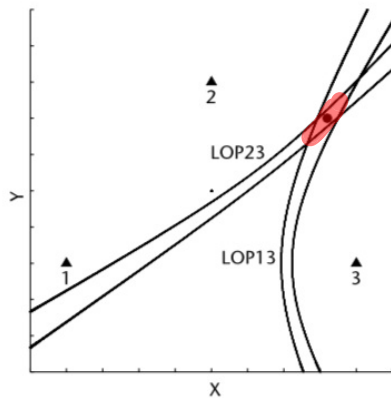
$$(T_2 - T_1) \left( \frac{\Delta T \times c}{\Delta d} \right)$$



# Multilateración Hiperbólica



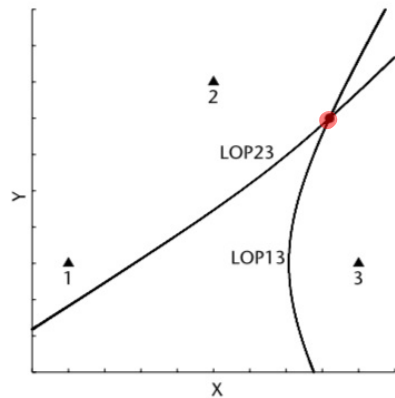
$$\Delta t = c \Delta t$$



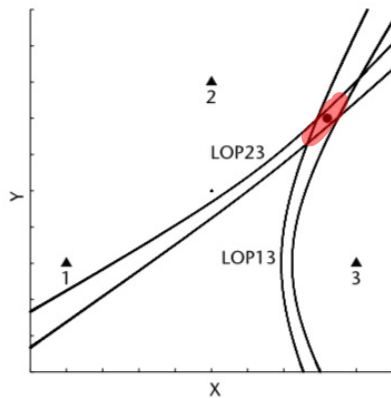
$$\Delta d = c(\Delta t + n)$$

$\Delta t$

# Multilateración Hiperbólica



$$\Delta d = c \Delta t$$



$$\Delta d = c(\Delta t + n)$$

$\Delta t$

**No requiere sincronización de reloj entre el móvil y las ER**

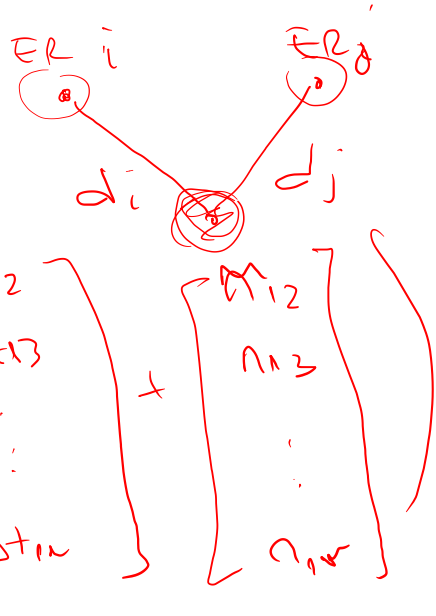
# TDOA

**Requiere sincronización de reloj entre las ER, puede  
requerir marcas de tiempo, requiere mayor ancho de banda**

# Multilateración Hiperbólica

$$\Delta \hat{r}_k = (d_i - d_j) = c(\Delta t_{ij} + n_{ij})$$

$$\begin{bmatrix} d_1 - d_2 \\ d_1 - d_3 \\ \vdots \\ d_1 - d_n \end{bmatrix} = \hat{\Delta R} = \begin{bmatrix} c(\Delta t_{12} + n_{12}) \\ c(\Delta t_{13} + n_{13}) \\ \vdots \\ c(\Delta t_{1n} + n_{1n}) \end{bmatrix} = c \left( \begin{bmatrix} \Delta t_{12} \\ \Delta t_{13} \\ \vdots \\ \Delta t_{1n} \end{bmatrix} + \begin{bmatrix} n_{12} \\ n_{13} \\ \vdots \\ n_{1n} \end{bmatrix} \right)$$



$$f(s) = \begin{bmatrix} d_1 - d_2 \\ d_1 - d_3 \\ \vdots \\ d_1 - d_n \end{bmatrix} = \begin{bmatrix} \sqrt{(x-x_1)^2 + (y-y_1)^2} - \sqrt{(x-x_2)^2 + (y-y_2)^2} \\ \sqrt{(x-x_1)^2 + (y-y_1)^2} - \sqrt{(x-x_3)^2 + (y-y_3)^2} \\ \vdots \\ \sqrt{(x-x_1)^2 + (y-y_1)^2} - \sqrt{(x-x_n)^2 + (y-y_n)^2} \end{bmatrix}$$

$$\hat{\Delta}_2 = f(s) + N$$

$$s = \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{Eq}_i = \begin{bmatrix} x_i \\ y_i \end{bmatrix}$$

$$\hat{\Delta R} - F(s) = N$$

$$S = \begin{pmatrix} x \\ 1 \end{pmatrix}?$$

$$\argmin_s \left( \hat{\Delta R} - F(s) \right)^T (\Delta R - F(s))$$

$$\hat{S} = \argmin_s (N^T N)$$

$$\hat{S} = \argmin_s \left[ \left( \hat{\Delta R} - F(s) \right)^T \left( \hat{\Delta R} - F(s) \right) \right]$$

$\downarrow$   
 $\begin{bmatrix} c(\Delta T + n) \\ \vdots \end{bmatrix}$

$\begin{bmatrix} \sqrt{x_1 + x \dots} & \sqrt{x_2 + x \dots} \\ x_1 + x & \dots & x_N + x \dots \end{bmatrix}$



$$\hat{S} = \underset{S}{\text{argmin}} \left( (\hat{\Delta} - \underbrace{F(s)})^T (\Delta - \overline{F(s)}) \right)$$

expansion en  $S_0$  (cond. linéaire)

$$\hat{F}(s) = F(s_0) + \frac{\partial F}{\partial x} (x - x_0) + \frac{\partial F}{\partial y} (y - y_0) + \dots$$

$$\hat{F}(s) = F(s_0) + \begin{bmatrix} \frac{\partial F}{\partial x} & \frac{\partial F}{\partial y} \end{bmatrix} \begin{bmatrix} x - x_0 \\ y - y_0 \end{bmatrix}$$

$$\hat{F}(s) = \begin{bmatrix} f_1(s_0) \\ f_2(s_0) \\ \vdots \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \\ \vdots & \vdots \end{bmatrix}_{s_0} \begin{bmatrix} x - x_0 \\ y - y_0 \end{bmatrix}$$

$$\hat{F}(s) = \begin{bmatrix} f_1(s) \\ f_2(s) \\ \vdots \end{bmatrix} \approx \begin{bmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \\ \vdots & \vdots \end{bmatrix} \begin{bmatrix} x_1 - x_0 & y_2 - y_0 \\ x_1 - x_0 & y_2 - y_0 \end{bmatrix}$$

$$\hat{F}(s) = F(s_0) + H \bigg|_{s=s_0} (s - s_0) = F(s_0) + H \begin{bmatrix} x - x_0 \\ y - y_0 \end{bmatrix}$$

$$\hat{s} = \underset{s}{\operatorname{argmin}} \left\{ (\hat{\Delta}_L - F(s_0) - H(s - s_0))^T (\hat{\Delta}_R - F(s_0) - H(s - s_0)) \right\}$$

$$\hat{S} = \underset{S}{\operatorname{argmin}} \left\{ (\hat{\Delta R} - \hat{F}(s_0) - \underbrace{H(s-s_0)})^T (\hat{\Delta R} - \hat{F}(s_0) - \underbrace{H(s-s_0)}) \right\}$$

$$\hat{S} = \underset{S}{\operatorname{argmin}} \left\{ \underbrace{\left[ (\hat{\Delta R} - \hat{F}(s_0) - HS) - HS \right]^T}_{\Delta R} [\dots] \right\}$$

$$\hat{S} = \underset{S}{\operatorname{argmin}} \left\{ \left[ \hat{\Delta R} - HS \right]^T \left[ \hat{\Delta R} - HS \right] \right\} \text{ is it?}$$

$$\boxed{\hat{S} = (H^T H)^{-1} H^T \hat{\Delta R}}$$

$$\hat{S} = (H^T H)^{-1} H^T \Delta R$$

$$\hat{S} = (H^T H)^{-1} H^T (\hat{\Delta R} - F(s_0) + H s_0)$$

$$\hat{S} = (H^T H)^{-1} H^T (\hat{\Delta R} - F(s_0)) + \underline{S_0}$$

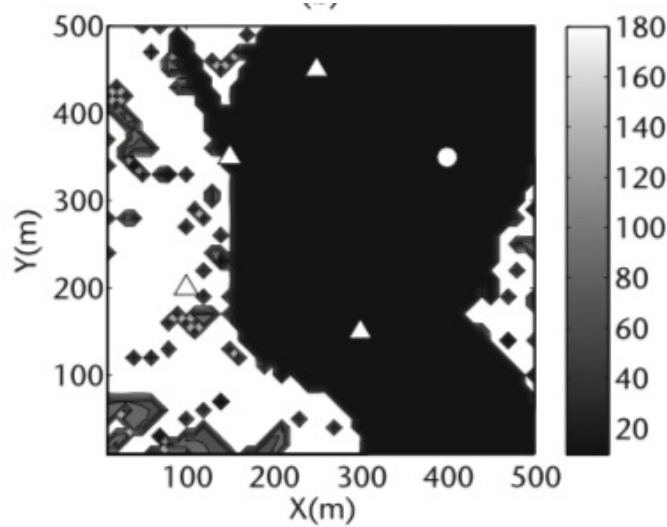
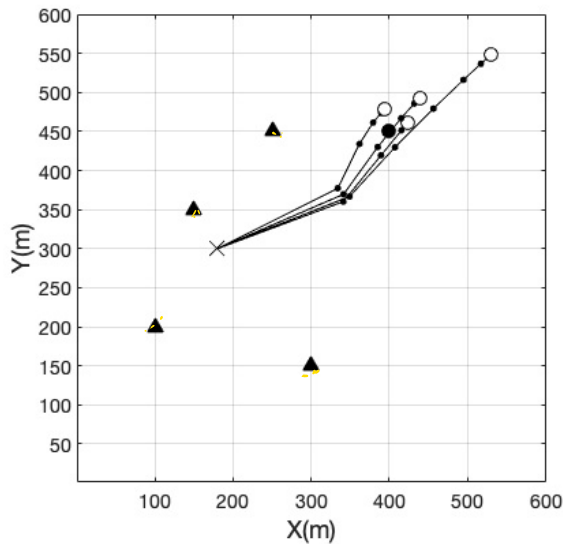
$$\hat{S}_{k+1} = \hat{S}_k + (H^T H)^{-1}_{s_k} H^T (\hat{\Delta R} - F(s_k))$$

## Multilateración Hiperbólica

$$\hat{S} = \arg \min_s \{(\Delta \hat{R} - F(s))^T (\Delta \hat{R} - F(s))\}$$

# Multilateración Hiperbólica: Ejercicio

Correr el programa ch02fig11.m para la misma CI y luego para distintas condiciones iniciales fuera del polígono que forman las ER y el móvil dentro y fuera también.



# Multilateración Hiperbólica: Filtro de Kalman

$$\nabla h \left[ \frac{\partial f_1}{\partial x} \quad \frac{\partial f_1}{\partial y} \quad 0 \quad 0 \right]$$

$$h = \left[ \sqrt{(x-x_1)^2 + (y-y_1)^2} - \sqrt{(x-x_2)^2 + (y-y_2)^2} \right]$$

~~$$\hat{\mathbf{x}}_{k|k-1} = \mathbf{F}_k \hat{\mathbf{x}}_{k-1|k-1}$$~~

$$\mathbf{P}_{k|k-1} = \mathbf{F}_k \mathbf{P}_{k-1|k-1} \mathbf{F}_k^T + \mathbf{G}_k \mathbf{Q}_k \mathbf{G}_k^T$$

$$\hat{\mathbf{x}}_{k|k} = \hat{\mathbf{x}}_{k|k-1} + \mathbf{W}_k [\hat{\mathbf{z}}_k - \hat{\mathbf{h}}(\hat{\mathbf{x}}_{k|k-1})]$$

$$\mathbf{P}_{k|k} = \mathbf{P}_{k|k-1} - \mathbf{W}_k \mathbf{S}_k \mathbf{W}_k^T$$

$$\mathbf{W}_k = \mathbf{P}_{k|k-1} (\nabla^T \mathbf{h}_k^x) \mathbf{S}_k^{-1}$$

$$\mathbf{S}_k = (\nabla \mathbf{h}_k^x) \mathbf{P}_{k|k-1} (\nabla^T \mathbf{h}_k^x) + \mathbf{R}_k$$

# Multilateración Hiperbólica: Método de Lavenberg-Marquardt

$$\mathbf{x}_{k+1} = \mathbf{x}_k + (\mathbf{A}^k + \lambda^k \mathbf{I})^{-1} \cdot \mathbf{g}_k$$

$$\mathbf{s}_k = \mathbf{s}_k + (\mathbf{W}^T \mathbf{H})^{-1} \mathbf{H}^T (\hat{\Delta \mathbf{L}} - \mathbf{F}(\mathbf{s}_k))$$

$$\mathbf{A}^k = [\Phi_k^T \Sigma^{-1} \Phi_k]$$

$$\Phi_k = \begin{bmatrix} \frac{x_k - x_1}{\sqrt{(x_k - x_1)^2 + (y_k - y_1)^2}} \\ \vdots \\ \frac{x_k - x_n}{\sqrt{(x_k - x_1)^2 + (y_k - y_1)^2}} \end{bmatrix}$$

$$\frac{y_k - y_1}{\sqrt{(x_k - x_1)^2 + (y_k - y_1)^2}}$$



$$\mathbf{x}_{k+1} = \mathbf{x}_k + (\mathbf{A}^k + \lambda^k \mathbf{I})^{-1} \cdot \mathbf{g}_k$$

$$S_k = S_k + (W^T H)^{-1} H^T (\hat{\Delta}_k - F(s_k))$$

$$\Lambda_k = [\Phi^T \Sigma^{-1} \Phi]$$

$$g_k = \Phi^T \Sigma^{-1} (\hat{\Delta}_k - \hat{d}_k)$$

dist. med.

$$\frac{(\hat{d}_1 - \hat{d}_2)}{\sqrt{(x_1 - x_k)^2 + (y_1 - y_k)^2}}$$

$$\lambda^k = \sum_{i=1}^{k_{\max}} \max \{ A_k / i, i \}$$

$k_{\max} \sim 5$

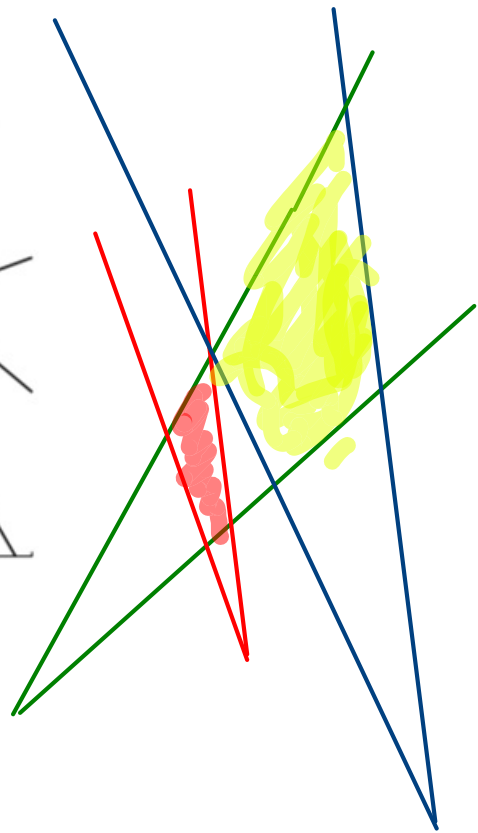
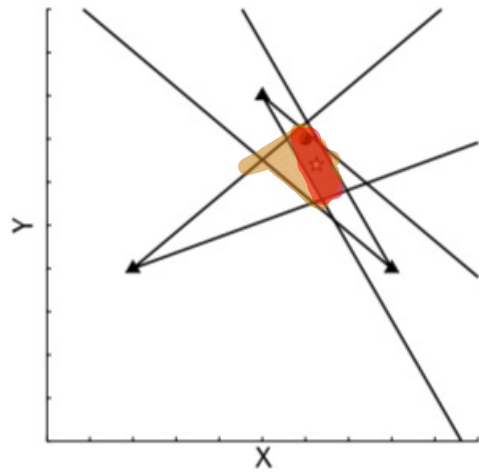
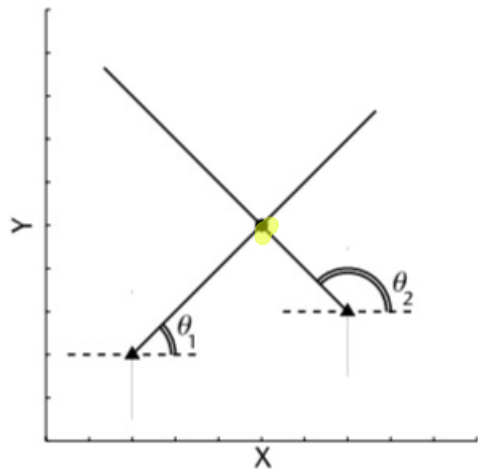
$$\begin{bmatrix} 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$J = \frac{E(x_k) - E(x_{k+1})}{H_k^T (\Delta^k H_k + g_k)} \rightarrow \underline{\text{escalon}}$$

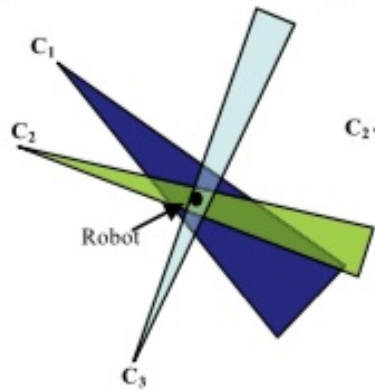
$$E(x_k) = \underbrace{[d - \hat{d}_k]}^T \underbrace{\bar{Z}^{-1}} \underbrace{[d - \hat{d}_k]}$$

si  $\rho > 0 \rightarrow$  usar el alg. normal  
 $\rho \leq 0 \quad \Delta_{k+1} = \Delta_k \quad g_{k+1} = g_k$

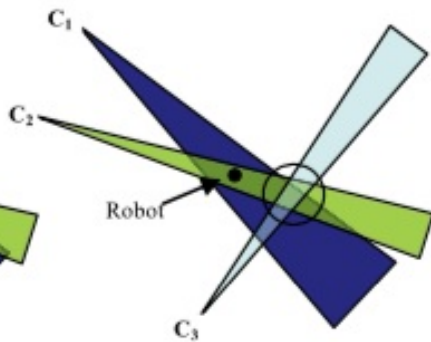
Multiangulación: DOA/AOA



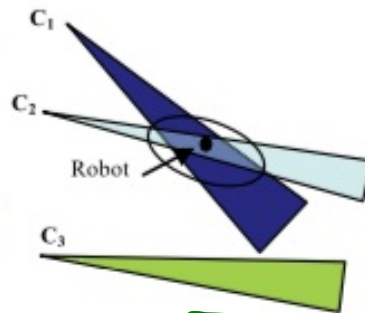
a



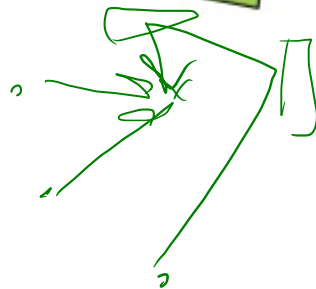
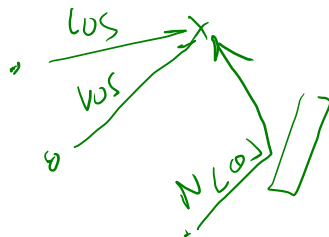
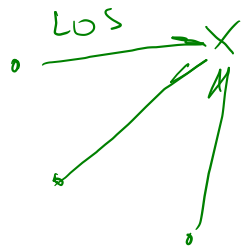
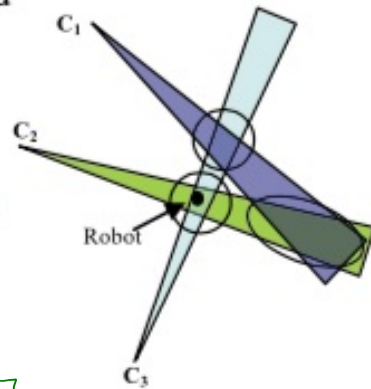
b



c



d



OUTLIER