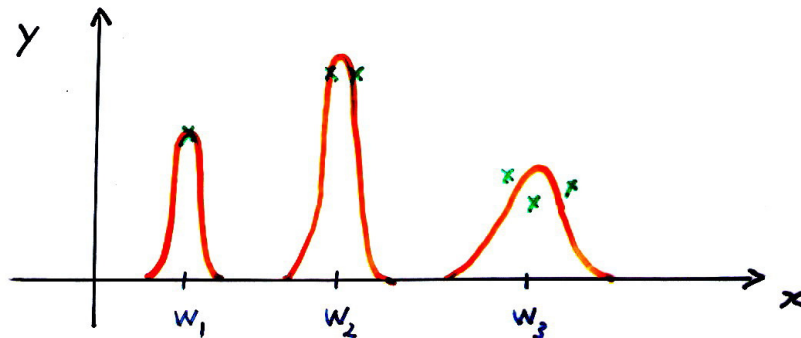


RBF - Radial Basis Function

$$y = f(x) \quad (x_i, y_i)$$



$$f_i(x) = f_i(u_i) \quad u_i = |x - w_i|^2$$

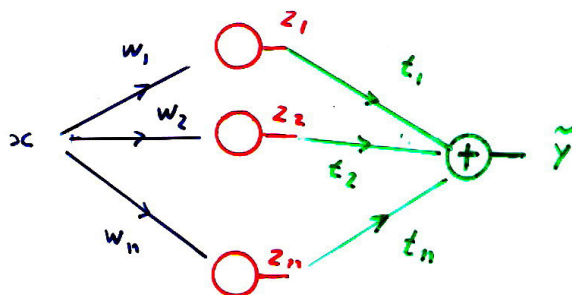
$$f_i \quad e^{-u_i / 2\sigma_i^2}$$

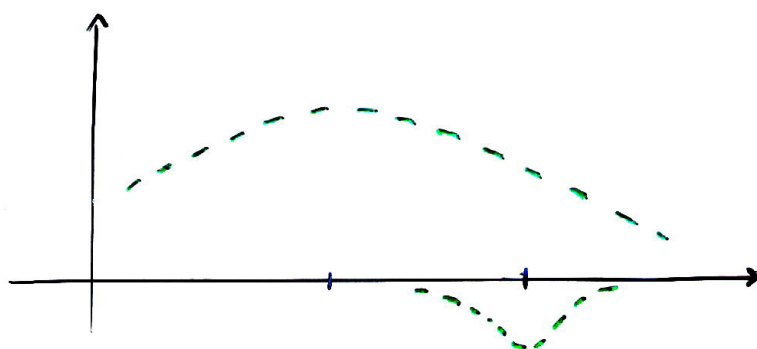
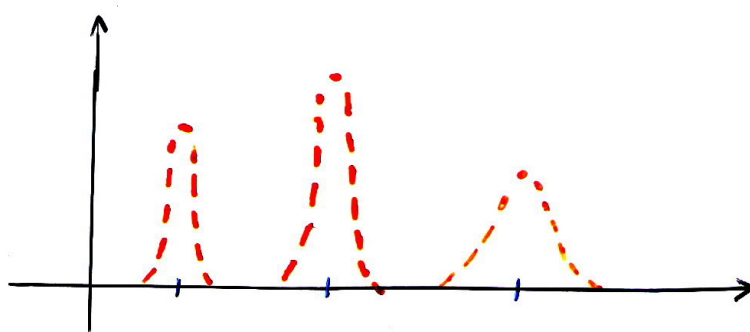
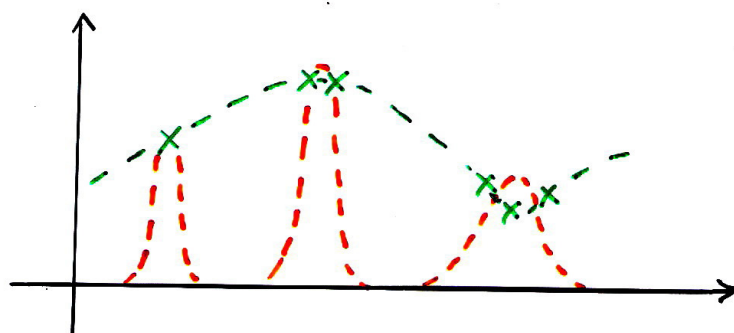
$$\frac{1}{1 + u_i / \sigma_i^2}$$

$$f_i \in [0, 1] \quad \begin{cases} x \approx w_i & f_i \approx 1 \\ x \neq w_i & f_i \approx 0 \end{cases}$$

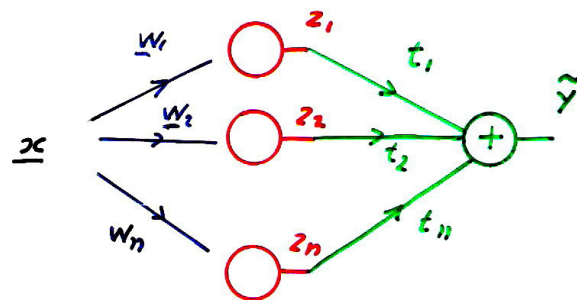
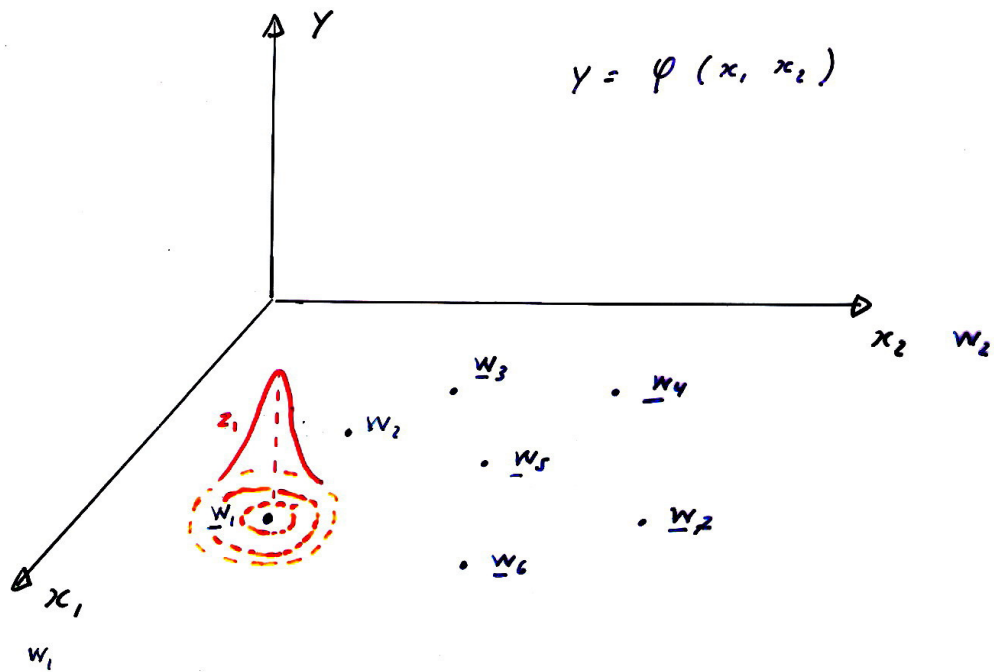
$$z_i = f_i(x) \quad \begin{cases} \approx 1 & x \approx w_i \\ \approx 0 & x \neq w_i \end{cases}$$

$$y_i = \sum_i t_i z_i \quad \begin{cases} \approx t_i & x \approx w_i \\ \approx 0 & x \neq w_i \quad \forall i \end{cases}$$





RBF - Radial Basis Function



$$u_i = d_i^2 \\ = \|x - w_i\|^2$$

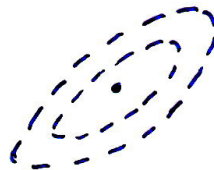
$$z_i = e^{-\frac{u_i}{2\sigma_i^2}}$$

$$\tilde{y} = \sum t_i z_i$$

Generalizar ξ_0

$$u_i = \|\underline{x} - \underline{w}_i\|^2$$

$$u_i = (\underline{x} - \underline{w}_i)^t \underline{K}_i (\underline{x} - \underline{w}_i) = \underline{d}_i^t \underline{K}_i \underline{d}_i$$



$$\underline{d}_i = \underline{x} - \underline{w}_i = \underline{d}_l^i]$$

$$\underline{K}_i = \underline{I} \quad u_i = \underline{d}_i^t \underline{I} \underline{d}_i = \underline{d}_i^t \underline{d}_i = \|\underline{d}_i\|^2$$

i.e. $u_i = \text{cte} \Rightarrow \text{esfera}$

$$\underline{K}_i \text{ diagonal} + \quad u_i = \sum_l k_{ll} d_l^2$$

elipsoide com eixos alinhados

$\underline{K}_i$ triangular superior
+

$$u_i = \sum_l \sum_{m \geq l} k_{lm} d_l d_m$$

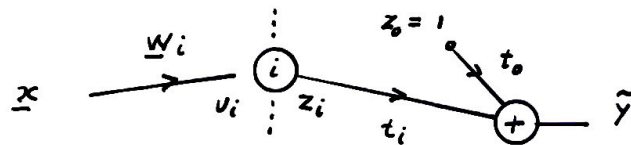
$$k_{lm} = 0 \Leftarrow m < l$$

elipsoide

Parâmetros

$$t_i \quad k_{lm}^i \quad w_l^i$$

Treinar por BP !



$$F = E(\varepsilon^2) \quad \varepsilon^2 = (y - \tilde{y})^2$$

$$\Delta p = -\alpha \frac{d\varepsilon^2}{dp} = 2\alpha \varepsilon \frac{d\tilde{y}}{dp}$$

$$\tilde{y} = \sum_i t_i z_i$$

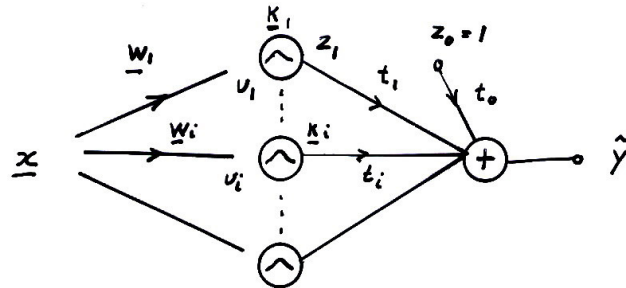
$$z_i = e^{-u_i}$$

$$u_i = \underline{d}_i^t \underline{K}_i \underline{d}_i = \sum_l \sum_{m \geq l} k_{lm}^i d_l^i d_m^i$$

$$d_l^i = x_l - w_l^i$$

Signal Feedforward 8

Derivadas Parciais



$$\underline{d}_i = \underline{x} - \underline{w}_i$$

$$d_l^i = x_l - w_l^i$$

$$\frac{d d_l^i}{d w_l^i} = -1$$

$$v_i = \underline{d}_i^t \underline{k}_i \underline{d}_i = \sum_l \sum_{m \geq l} k_{lm}^i d_l^i d_m^i$$

$$s_l^i = \sum_{m \geq l} k_{lm}^i d_m^i$$

$$v_i = \sum_l d_l^i s_l^i$$

$$\frac{d v_i}{d k_{lm}^i} = d_l^i d_m^i$$

$$\frac{d v_i}{d d_l^i} = k_{ll}^i d_l^i + s_l^i$$

$$z_i = e^{-v_i}$$

$$\frac{dz_i}{dv_i} = -z_i$$

$$\tilde{y} = \sum_{i=0} t_i z_i$$

$$\frac{d\tilde{y}}{dt_i} = z_i$$

$$z_0 = 1$$

$$\frac{d\tilde{y}}{dz_i} = t_i$$

Error feedback

$$\Delta t_i = 2\alpha \varepsilon \frac{d\tilde{y}}{dt_i}$$

$$= 2\alpha \varepsilon z_i$$

$$\Delta k_{lm}^i = 2\alpha \varepsilon \frac{d\tilde{y}}{dk_{lm}^i} = 2\alpha \varepsilon \frac{d\tilde{y}}{dz_i} \frac{dz_i}{dv_i} \frac{dv_i}{dk_{lm}^i}$$

$$= 2\alpha \varepsilon (t_i)(-z_i)(d_l^i d_m^i)$$

$$= -2\alpha \varepsilon t_i z_i d_l^i d_m^i$$

$$\Delta w_l^i = 2\alpha \varepsilon \frac{d\tilde{y}}{dw_l^i} = 2\alpha \varepsilon \frac{d\tilde{y}}{dz_i} \frac{dz_i}{dv_i} \frac{dv_i}{dw_l^i} \frac{dw_l^i}{dd_l^i} \frac{dd_l^i}{dw_l^i}$$

$$= 2\alpha \varepsilon (t_i)(-z_i) (d_l^i k_{ll}^i + s_l^i) (-1)$$

$$= 2\alpha \varepsilon t_i z_i (d_l^i k_{ll}^i + s_l^i)$$

Caso particular

$$\underline{K}_i = k_i \underline{I}$$

$$d_{ij} = x_j - w_{ij}$$

$$\frac{d}{dw_{ij}} d_{ij} = -1$$

$$D_i^2 = \sum_j d_{ij}^2$$

$$\frac{d}{d_{ij}} D_i^2 = 2 d_{ij}$$

$$v_i = k_i D_i^2$$

$$\frac{dv_i}{dk_i} = D_i^2$$

$$\frac{dv_i}{dD_i^2} = k_i$$

$$z_i = e^{-v_i}$$

$$\frac{dz_i}{dv_i} = -z_i$$

$$\tilde{y} = \sum_{i=0} t_i z_i$$

$$\frac{d\tilde{y}}{dt_i} = z_i$$

$$\frac{d\tilde{y}}{dz_i} = t_i$$

$$\Delta p = -2 \alpha \varepsilon \frac{d\tilde{y}}{dp}$$

$$\frac{d\tilde{y}}{dt_i} = z_i$$

$$\frac{d\tilde{y}}{dk_i} = \frac{d\tilde{y}}{dz_i} \frac{dz_i}{dv_i} \frac{dv_i}{dk_i} = t_i (-z_i) D_i^2$$

$$\frac{d\tilde{y}}{dw_{ij}} = \frac{d\tilde{y}}{dz_i} \frac{dz_i}{dv_i} \frac{dv_i}{dD_i^2} \frac{dD_i^2}{dd_{ij}} \frac{dd_{ij}}{dw_{ij}} = t_i (-z_i) k_i 2 d_{ij} (-1)$$

Complexidade - # parâmetros $\left\{ \begin{array}{l} N \text{ entradas} \\ M \text{ neurônios} \end{array} \right. \begin{array}{l} 10 \\ 10 \end{array}$

K triangular superior

$$M \left[N + \frac{N(N+1)}{2} + 1 \right] + 1 \quad 661$$

K diagonal

$$M \left[N + N + 1 \right] + 1 \quad 211$$

K identidade

$$M \left[N + 1 + 1 \right] + 1 \quad 121$$

Comparando com feed forward clássico (BP)

$$M \left[N + 1 + 1 \right] + 1 \quad 121$$

Ativação dos neurônios $\left\{ \begin{array}{l} \text{RBF} - \text{local} \\ \text{FF} - \text{global} \end{array} \right.$

Simplificações

(ou valores iniciais)

W

aleatórios

entradas

clusterização (ART)

K

elipse (diagonal)

circular (identidade)

$$\sigma = \frac{d_{\max}}{\sqrt{2N}}$$

← entre centros W

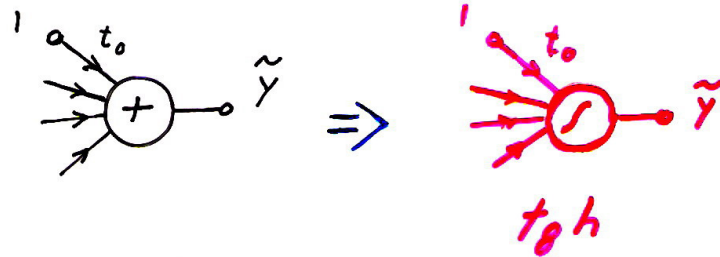
← nº clusters

local - global

⇒ Treinar !

RBF como classificador

Neurônio de saída

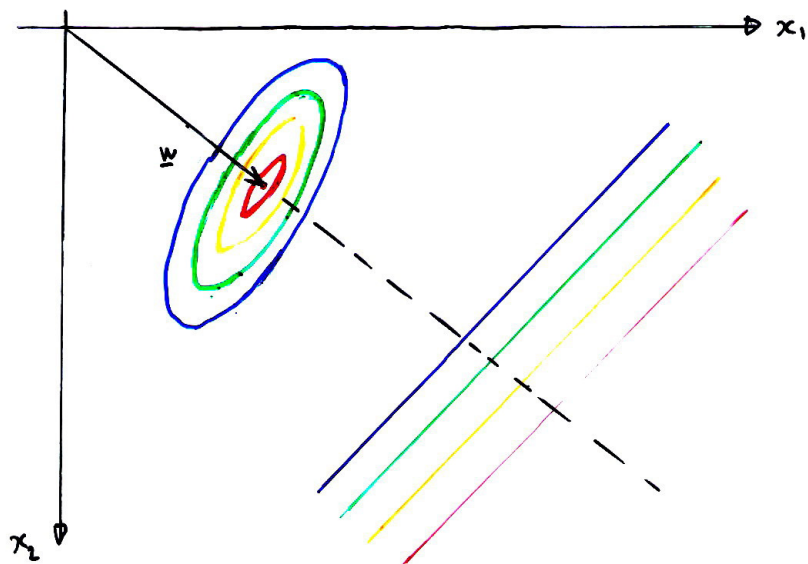
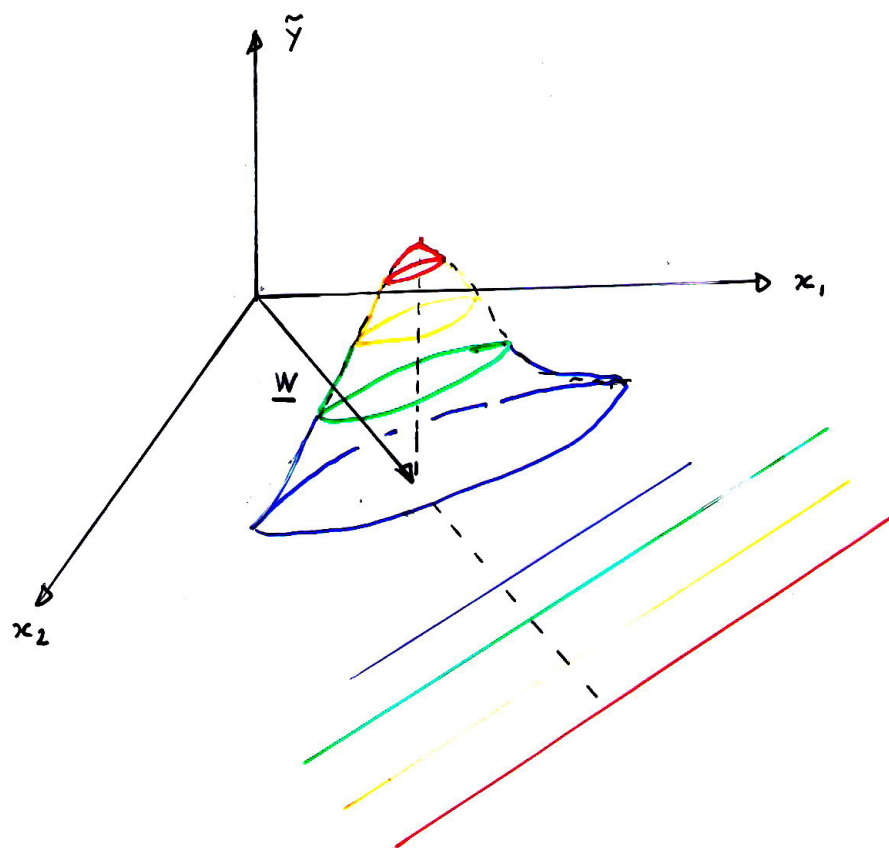


Separadores elementares

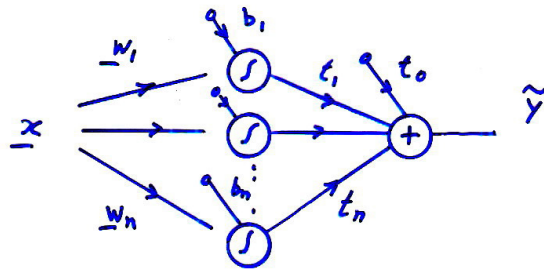
$\underline{K} = I \rightarrow$ esferas

$\underline{K} = \text{diag} \rightarrow$ elipsoides eixos alinhados

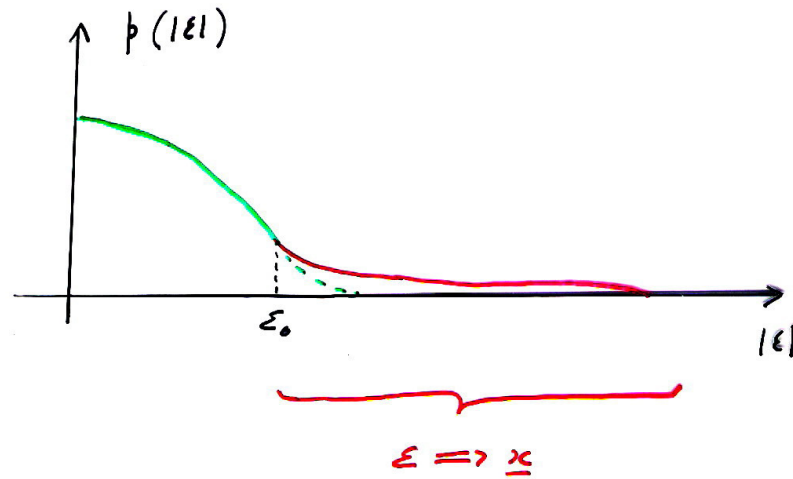
$\underline{K} = \text{triang sup} \rightarrow$ elipsoides



Auxílio à BP



$$(\underline{x}, y) \Rightarrow (\underline{x}, y, \tilde{y}, \epsilon) \Rightarrow (\underline{x}, \epsilon)$$



$$|\epsilon| > \epsilon_0 \quad \underline{x} \in X_0$$

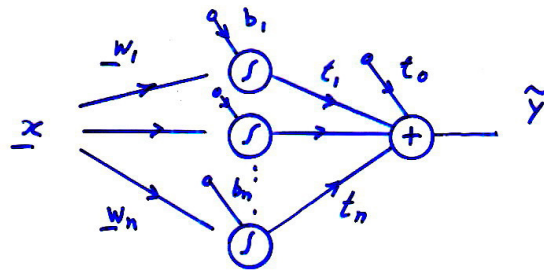
Clusterizer $\underline{x} \in X_0$

ART

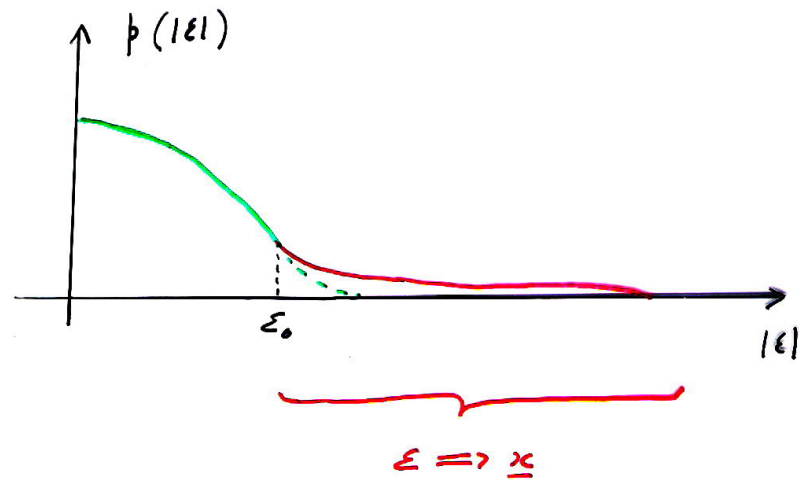
histograma $\Rightarrow r_0$

$\Rightarrow$ centros $\underline{m}_i$

Auxílio à BP



$$(\underline{x}, y) \Rightarrow (\underline{x}, y, \tilde{y}, \varepsilon) \Rightarrow (\underline{x}, \varepsilon)$$



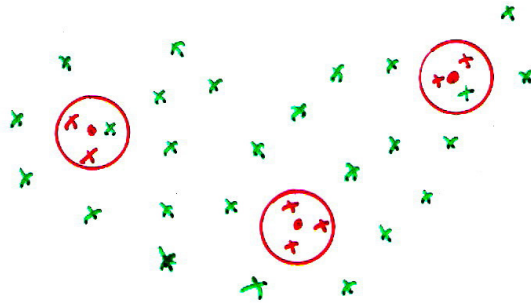
$$|\varepsilon| > \varepsilon_0 \quad \underline{x} \in X_0$$

Clusterizer $\underline{x} \in X_0$

ART

histograma $\Rightarrow r_0$

$\Rightarrow$ centros $\underline{m}_i$



Classificador: LB dos grandes erros

Classe dos grandes erros, X_1

Verificar população - identificar outliers



Classificar todos os x por região

$$\underline{x} \in X_2$$

mesmos centros

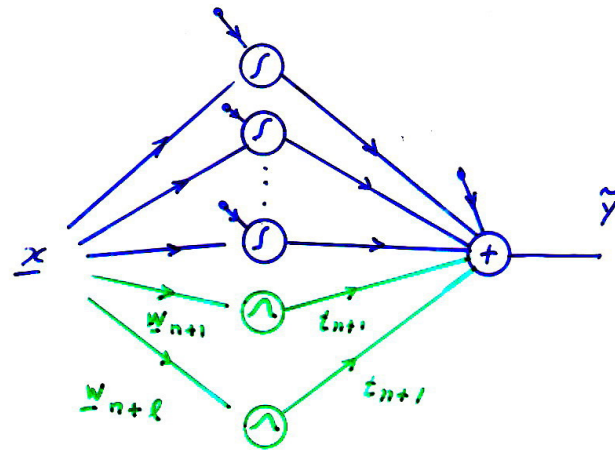
$$r_2 > r_0$$

Montar novo conjunto de treinamento

(apenas a região de grandes erros)
aumentada

$$(\underline{x}, y) \quad \forall \underline{x} \in X_2$$

Rede aumentada



Neurônios de base radial

início

$$w_{n+i} = m_i$$

um ou + por cento

esférico, $\sigma \approx r_0$

Treinar por BP

apenas os novos neurônios,
esféricos

apenas com (x, y) /

$$x \in X_2$$