Radial Basis Function Networks - Learning Algorithm for variable centres

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1 Notation

Weights (coefficients)	w	Centre	R
Target Output	t	Hidden Node/Centre	h
		Iterator	
Actual Output	у	Number of Centres	m
Input Pattern	x	Output Iterator	0
Input Dimension	i	Number of Outputs	k
Iterator		Trainser of Gaspass	11
Input Dimension	d	Standard Deviation	σ
Number of Inputs	n	Learning Rates	$\eta_R, \eta_\sigma,$
			η_w
Input Pattern	n	Bias unit	b
Iterator	p	Dias unit	u

2 Algorithm Pseudocode

- 1. Choose the number (m) and initial coordinates of the centres (R) of the RBF functions.
- 2. Choose the initial value of the spread parameter (σ) for each centre (R).
- 3. Initialise the weights/coefficients (w) to small random values [-1,1].
- 4. For each epoch (e)
- 5. For each input vector/pattern $(\mathbf{x}^{(p)})$
- 6. Calculate the output (y) of each output node (o) using eq. 1.
- 7. Update the network parameters (w, R, σ) using eqs. 6, 7, 8, 9.
- 8. end for (p = n)
- 9. end for (e = total epochs)

Note: Steps 1 and 2 can be performed using a clustering algorithm such as Kohonen SOMs.

3 Equations

1. The output of node o for pattern p is:

$$y_o^{(p)} = w_{bo} + \sum_{h=1}^m w_{ho} \phi(x^{(p)}, R_h, \sigma_h)$$

2. Gaussian function:

$$\phi(x^{(p)}, R_h, \sigma_h) = exp\left(-\frac{||x^{(p)} - R_h||^2}{2\sigma_h^2}\right)$$

3. Euclidean distance:

$$||x^{(p)} - R_h||^2 = \sum_{i=1}^d (x_i^{(p)} - R_{hi})^2$$

4. Total Error:

$$E = \frac{1}{2} \sum_{p=1}^{n} \sum_{o=1}^{k} \left[\varepsilon_o^{(p)} \right]^2$$

5. Error of output neuron o for pattern p:

$$\varepsilon_o^{(p)} = t_o^{(p)} - y_o^{(p)}$$

6. Weight/coefficient update equation (from hidden unit/centre h to output node o):

$$w'_{ho} = w_{ho} + \eta_w \varepsilon_o^{(p)} \phi(x^{(p)}, R_h, \sigma_h)$$

7. Weight/coefficient update equation (from bias unit b to output node o):

$$w_{bo}' = w_{bo} + \eta_w \varepsilon_o^{(p)}$$

8. Centre update equation (ithcoordinate of centre h):

$$R'_{hi} = R_{hi} + \eta_R \sum_{o=1}^{k} \varepsilon_o^{(p)} w_{ho} \phi(x^{(p)}, R_h, \sigma_h) \frac{x_i^{(p)} - R_{hi}}{\sigma_h^2}$$

9. Standard Deviation/width σ of Gaussian of centre h:

$$\sigma'_{h} = \sigma_{h} + \eta_{\sigma} \sum_{o=1}^{k} \varepsilon_{o}^{(p)} w_{ho} \phi(x^{(p)}, R_{h}, \sigma_{h}) \frac{||x^{(p)} - R_{h}||^{2}}{\sigma_{h}^{3}}$$