Analysis of Radial Basis Function Neural Networks (RBFNN)

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Abstract. In the following document, we have derived and analyzed variants of Radial Basis Function Neural Networks (RBFNN). There are four models we have worked on. The first part of the document consists of network diagrams, and a table of key equations and derivations of the back-propagation equations for all variant models.

The second part of the document contains test results of the rbfnn and its variants for X-OR classification problem.

1 Radial Basis Function Neural Networks

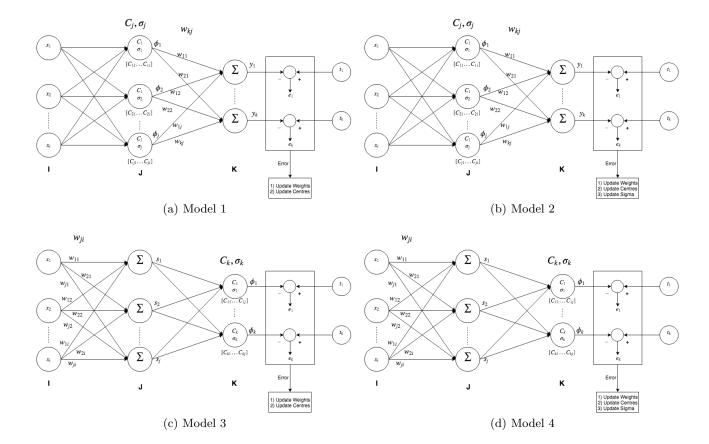


Table 1: Key equation, and back-propagation derivation results of all RBFNN models

	Key Equations	Gradient Descent Learning	Computation of $\frac{\partial E}{\partial w}$, $\frac{\partial E}{\partial c}$ & $\frac{\partial E}{\partial \sigma}$	
Model-1	$y_k = \sum\limits_j w_{kj} \phi_j$	Learning of weights:	$\frac{\partial E}{\partial m_{ki}} = \frac{\partial E}{\partial m_k} \cdot \frac{\partial y_k}{\partial m_{ki}} = -(t_k - y_k)\phi_j$	
	$\phi_j = e^{\frac{-z_j^2}{2\sigma_j^2}}$	$w_{kj}(t+1) = w_{kj}(t) - \eta_w \frac{\partial E}{\partial w_{kj}}$	$\partial w_{kj} = \partial y_k \partial w_{kj} = (\kappa g_k) \psi_j$	
	$z_j = X - C_j = \sqrt{\sum_i (x_i - c_{ji})^2}$	Learning of centers:	$\frac{\partial E}{\partial c_{ji}} = \left[\sum_{k} \frac{\partial E}{\partial y_{k}} \cdot \frac{\partial y_{k}}{\partial \phi_{j}} \right] \cdot \frac{\partial \phi_{j}}{\partial z_{j}} \cdot \frac{\partial z_{j}}{\partial c_{ji}}$	
	Cost function: $E = \frac{1}{2} \sum_{k} (t_k - y_k)^2$	$c_{ji}(t+1) = c_{ji}(t) - \eta_c \frac{\partial E}{\partial c_{ji}}$	$= \left[\sum_{k} -(t_k - y_k) w_{kj} \right] \times \frac{\phi_j}{\sigma_j^2} \times (x_i - c_{ji})$	
		Learning of weights:	$\frac{\partial E}{\partial w_{ki}} = \frac{\partial E}{\partial u_k} \cdot \frac{\partial y_k}{\partial w_{ki}} = -(t_k - y_k)\phi_j$	
Model-2	$y_k = \sum\limits_j w_{kj} \phi_j$	$w_{kj}(t+1) = w_{kj}(t) - \eta_w \frac{\partial E}{\partial w_{kj}}$	$\partial w_{kj} \partial y_k \partial w_{kj} (\overset{\bullet}{}_{} \overset{\bullet}{}_{} \overset{\bullet}{}_{})^{\gamma} f$	
	$\phi_j = e^{\frac{-z_j^2}{2\sigma_j^2}}$	Learning of centers:	$\frac{\partial E}{\partial c_{ji}} = \left[\sum_{k} \frac{\partial E}{\partial y_{k}} \cdot \frac{\partial y_{k}}{\partial \phi_{j}} \right] \cdot \frac{\partial \phi_{j}}{\partial z_{j}} \cdot \frac{\partial z_{j}}{\partial c_{ji}}$	
	$z_j = X - C_j = \sqrt{\sum_i (x_i - c_{ji})^2}$	$c_{ji}(t+1) = c_{ji}(t) - \eta_c \frac{\partial E}{\partial c_{ji}}$	$= \left[\sum\limits_{k} -(t_k - y_k)w_{kj}\right] \times \frac{\phi_j}{\sigma_j^2} \times (x_i - c_{ji})$	
	Cost function: $E = \frac{1}{2} \sum_{k} (t_k - y_k)^2$	Learning of sigma:	$\frac{\partial E}{\partial \sigma_j} = \left[\sum_{k} \frac{\partial E}{\partial y_k} \cdot \frac{\partial y_k}{\partial \phi_j} \right] \cdot \frac{\partial \phi_j}{\partial \sigma_j}$	
		$\sigma_j(t+1) = \sigma_j(t) - \eta_\sigma \frac{\partial E}{\partial \sigma_j}$	$=\left[\sum\limits_{k}-(t_{k}-y_{k})w_{kj} ight] imesrac{\phi_{j} imes z_{j}^{2}}{\sigma_{j}^{3}}$	
Model-3	$s_j = \sum\limits_i w_{ji} x_i$	Learning of weights:	$\frac{\partial E}{\partial w_{ji}} = \left[\sum_{k} \frac{\partial E}{\partial \phi_{k}} \cdot \frac{\partial \phi_{k}}{\partial z_{k}} \cdot \frac{\partial z_{k}}{\partial s_{j}}\right] \cdot \frac{\partial s_{j}}{\partial w_{ji}}$	
	$\phi_k = e^{rac{-z_k^2}{2\sigma_k^2}}$	$w_{ji}(t+1) = w_{ji}(t) - \eta_w \frac{\partial E}{\partial w_{ji}}$	$=\left[\sum\limits_{k}(t_{k}-\phi_{k}) imesrac{\phi_{k}}{\sigma_{k}^{2}} imes\left(s_{j}-c_{kj} ight) ight] imes x_{k}$	
	$z_k = S - C_k = \sqrt{\sum_i (s_j - c_{kj})^2}$	Learning of centers:	$\frac{\partial E}{\partial c_{kj}} = \frac{\partial E}{\partial \phi_k} \cdot \frac{\partial \phi_k}{\partial z_k} \cdot \frac{\partial z_k}{\partial c_{kj}}$	
	Cost function: $E = \frac{1}{2} \sum_{k} (t_k - \phi_k)^2$	$c_{kj}(t+1) = c_{kj}(t) - \eta_c \frac{\partial E}{\partial c_{kj}}$	$= -(t_k - \phi_k) \times \frac{\phi_k}{\sigma_k^2} \times (s_j - c_{kj})$	
		Learning of weights:	$\frac{\partial E}{\partial w_{ji}} = \left[\sum_{k} \frac{\partial E}{\partial \phi_{k}} \cdot \frac{\partial \phi_{k}}{\partial z_{k}} \cdot \frac{\partial z_{k}}{\partial s_{j}}\right] \cdot \frac{\partial s_{j}}{\partial w_{ji}}$	
Model-4	$s_j = \sum\limits_i w_{ji} x_i$	$w_{ji}(t+1) = w_{ji}(t) - \eta_w \frac{\partial E}{\partial w_{ji}}$	$= \left[\sum_{k} (t_k - \phi_k) \times \frac{\phi_k}{\sigma_k^2} \times (s_j - c_{kj})\right] \times x_i$	
	$\phi_k = e^{\frac{-z_k^-}{2\sigma_k^2}}$	Learning of centers:	$\frac{\partial E}{\partial c_{kj}} = \frac{\partial E}{\partial \phi_k} \cdot \frac{\partial \phi_k}{\partial z_k} \cdot \frac{\partial z_k}{\partial c_{kj}}$	
	$z_k = S - C_k = \sqrt{\sum_i (s_j - c_{kj})^2}$	$c_{kj}(t+1) = c_{kj}(t) - \eta_c \frac{\partial E}{\partial c_{kj}}$	$= -(t_k - \phi_k) \times \frac{\phi_k}{\sigma_k^2} \times (s_j - c_{kj})$	
	Cost function: $E = \frac{1}{2} \sum_{k} (t_k - \phi_k)^2$	Learning of sigma:	$\frac{\partial E}{\partial \sigma_k} = \frac{\partial E}{\partial \phi_k} \cdot \frac{\partial \phi_k}{\partial \sigma_k}$	
		$\sigma_k(t+1) = \sigma_k(t) - \eta_\sigma \frac{\partial E}{\partial \sigma_k}$	$= -(t_k - \phi_k) \times \frac{\phi_k \times z_k^2}{\sigma_k^3}$	

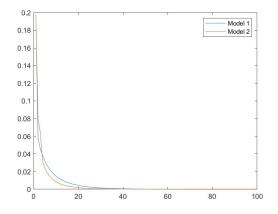
2 Training Results for XOR Problem

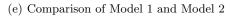
Table 2: RBF Network used and their respective learning parameters

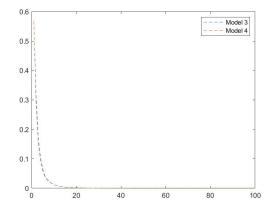
	Network configuration	Learning Parameters			
	Network configuration	Eta-Center	Eta-Weight	Eta-Sigma	
		η_c	η_w	η_σ	
Model-1	2-4-1	0.9099	0.4788	-	
Model-2	2-4-1	0.9099	0.4788	0.5	
Model-3	2-4-1	0.6	0.99	-	
Model-4	2-4-1	0.4	0.99	10^{-3}	

Table 3: Convergence Characteristics

	Number of iterations for convergence	MSE at convergence
Model-1	34	9.9083×10^{-4}
Model-2	25	9.1492×10^{-4}
Model-3	24	8.676×10^{-4}
Model-4	24	8.654×10^{-4}



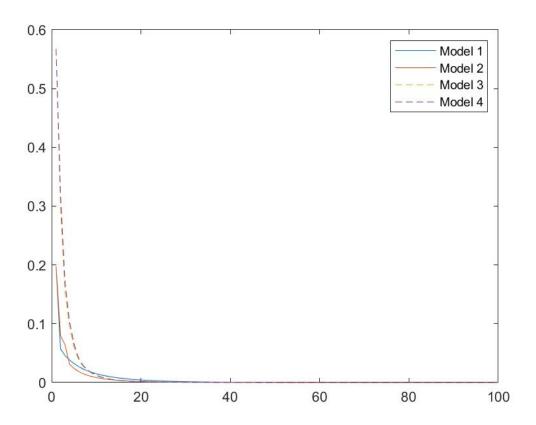




(f) Comparison of Model 3 and Model 4 $\,$

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Fig. 1: Convergence characteristics comparison of all Models



3 XOR classification results

Table 4: Training results for RBFNN $\,$

	Testing Data		Actual Output (AO)	Model Output (MO)	Deviation	Mean deviation
	0.1	0.1	0.1	0.1034	0.0034	
Model-1	0.1	0.9	0.9	0.8976	0.0024	0.0019
Wodel-1	0.9	0.1	0.9	0.8986	0.0014	
	0.9	0.9	0.1	0.1006	0.0006	
	0.1	0.1	0.1	0.1041	0.0041	0.0021
Model-2	0.1	0.9	0.9	0.8990	0.0010	
Wiodei-2	0.9	0.1	0.9	0.8989	0.0011	
	0.9	0.9	0.1	0.0979	0.0021	
	0.1	0.1	0.1	0.1026	0.0026	0.0011
Model-3	0.1	0.9	0.9	0.9000	0.0000	
Wodel-3	0.9	0.1	0.9	0.9000	0.0000	0.0011
	0.9	0.9	0.1	0.1017	0.0017	
	0.1	0.1	0.1	0.1029	0.0029	
Model-4	0.1	0.9	0.9	0.9000	0.0000	0.0012
1410061-4	0.9	0.1	0.9	0.9000	0.0000	
	0.9	0.9	0.1	0.1018	0.0018	