

Neural Network Controller

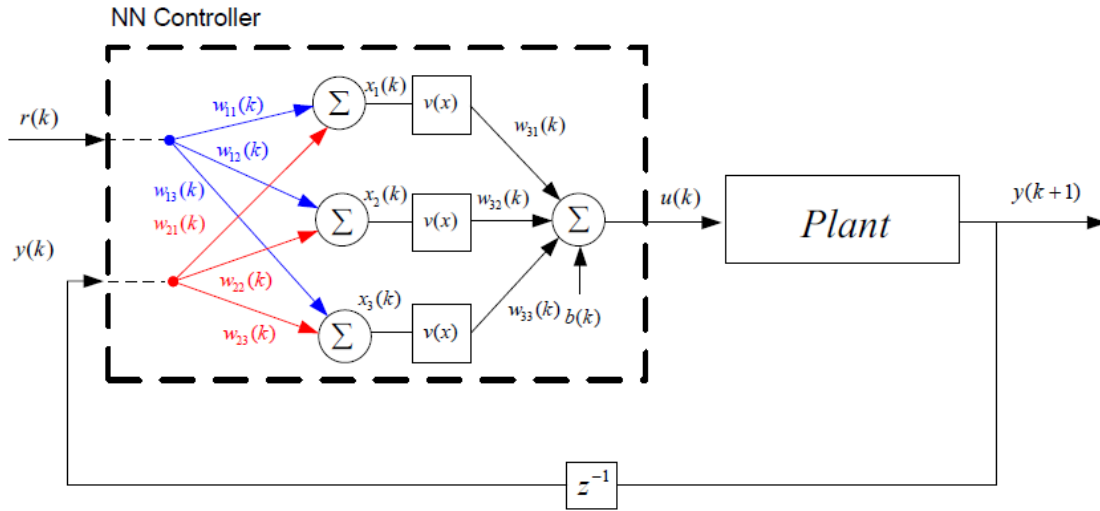


Fig.1 Neural Network Structure

Neural Network structure can be used as a controller directly as in fig.1. The performance of the controller is going to be evaluated on the following nonlinear time-varying plant :

$$y(k+1) = \frac{1.2(1-0.8e^{-0.1k})}{1+y^2(k)} y(k) + u(k) \quad (1)$$

The initial values of weights in NN controller are initialized as in Table 1.

Table 1.Symbols and Initial values

| Symbol | Value | Description |
|--|--|-----------------------------------|
| $W(0) = \begin{bmatrix} w_{11}(0) & w_{12}(0) & w_{13}(0) \\ w_{21}(0) & w_{22}(0) & w_{23}(0) \\ w_{31}(0) & w_{32}(0) & w_{33}(0) \end{bmatrix}$ $b(0)$ | $W(0) = \begin{bmatrix} -1.7502 & -0.8314 & -1.1564 \\ -0.2857 & -0.9792 & -0.5336 \\ -2.0026 & 0.9642 & 0.5201 \end{bmatrix}$ $b(0) = 10^{-5}$ | Weights in input and hidden layer |
| $u(0)$ | $u(0) = 0$ | Control Signal |
| $\eta(k)$ | $\eta(k) = 0.01$ | Learning Rate |
| t_s | $t_s = 0.01$ | Sampling Time |

The system jacobian information is approximated using the following equation:

$$\frac{\partial y(k+1)}{\partial u(k)} = \begin{cases} 1 & , \text{ if } u(k) - u(k-1) \cong 0 \\ \frac{y(k+1) - y(k)}{u(k) - u(k-1)} & , \text{ otherwise} \end{cases} \quad (2)$$

Utilize tan-sigmoid function ($v(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$) as activation functions in hidden layer.

minimize the following objective function:

$$J(e_r(k)) = \frac{1}{2} e_r^2(k) \quad (3) \quad \text{where} \quad e_r(k) = r(k) - y(k)$$

Reference Signal:

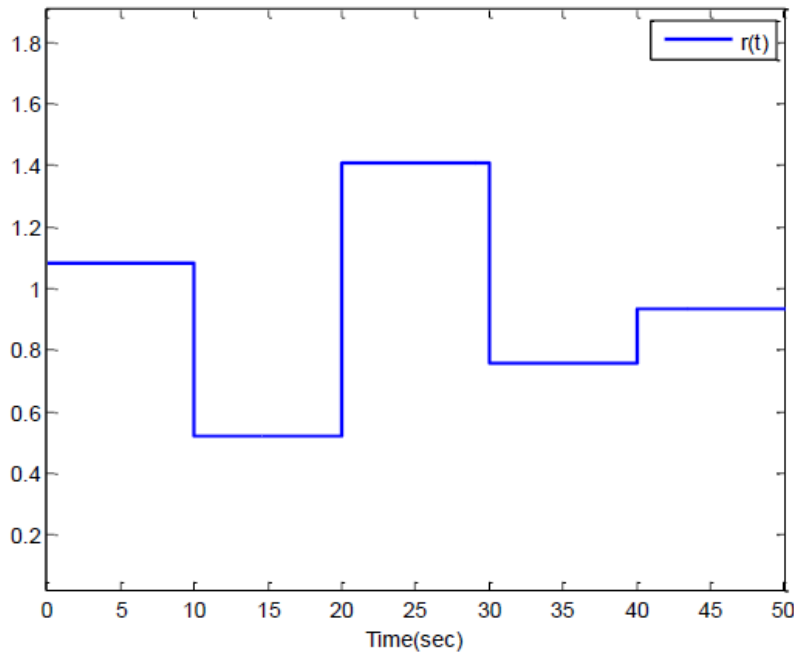


Fig.2 Reference Signal

Explanation of the code:

$etr(k) = r(k) - y(k)$. The back propagation algorithm is given below. It is used to update weights in order to find best PID-NN coefficients:

$$\frac{\partial J}{\partial w_{ij}} = \frac{\partial J}{\partial e} \frac{\partial e}{\partial y} \frac{\partial y}{\partial v_j} \frac{\partial v_j}{\partial w_{ij}}$$

$$\frac{\partial y}{\partial v_j} = \varphi_j'(v_j)$$

$v(x)$ is the activation func which is $\frac{e^x - e^{-x}}{e^x + e^{-x}}$

$$\frac{\partial J}{\partial e} = e$$

$$\frac{\partial e}{\partial y} = -1$$

$$\frac{\partial v_j}{\partial w_{ij}} = y_{Hj}$$

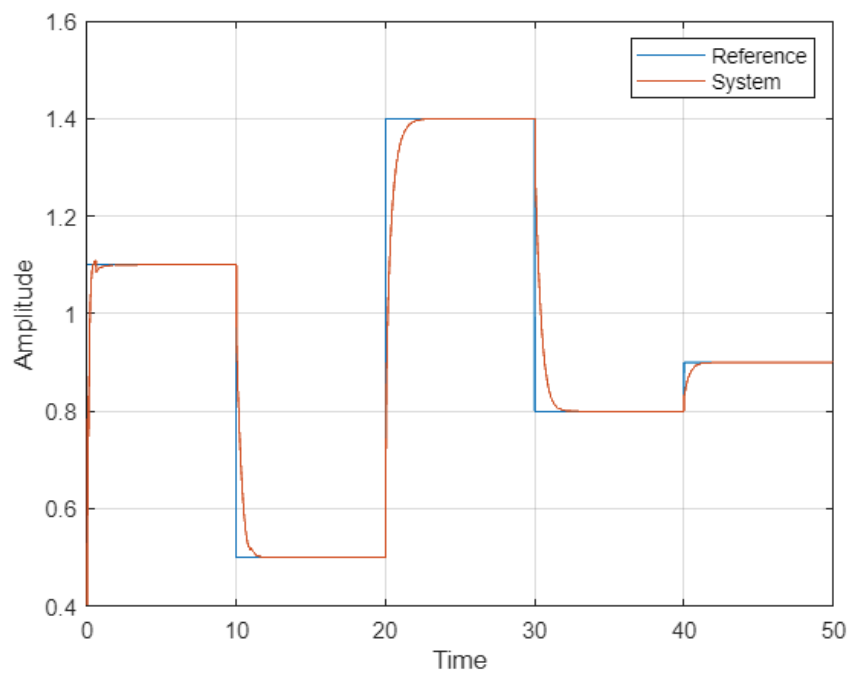
after the values are substituted we can obtain new w by this eq.

$$w_{ij} = (k + 1) = w_{ij}(k) + \eta \frac{\partial J}{\partial w_{ij}}$$

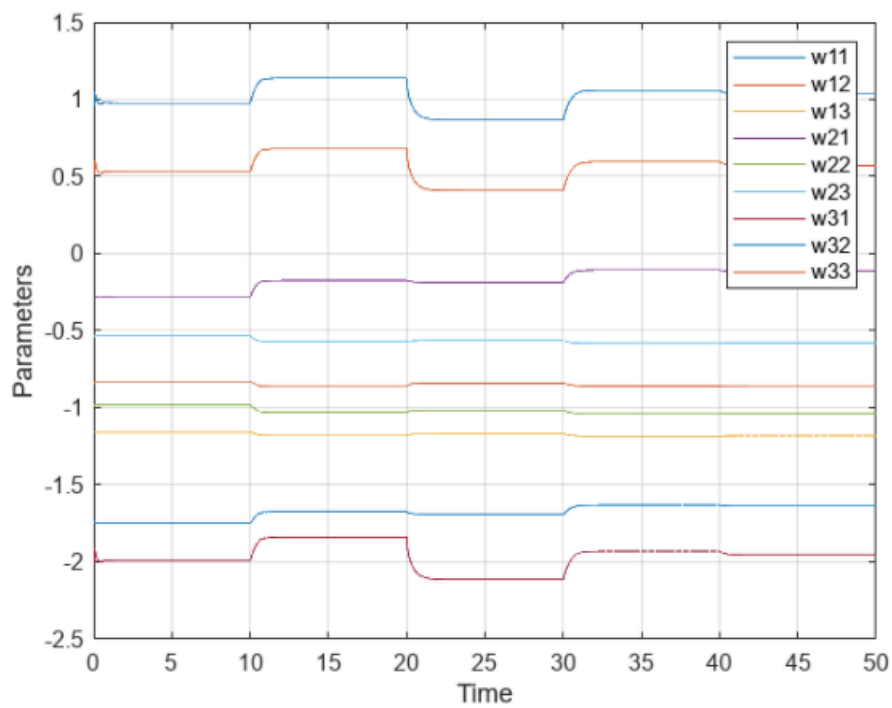
$$y(k + 1) = \frac{1.2(1 - 0.8e^{-0.1k})y(k)}{1 + y^2(k)} + u(k)$$

Gradient descent is used for cost function to be below the threshold.

Reference and system output signals:



Weight signals:



Controller signal:

