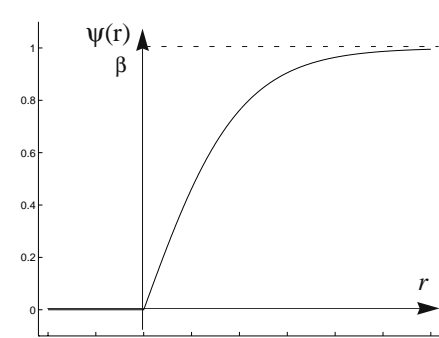
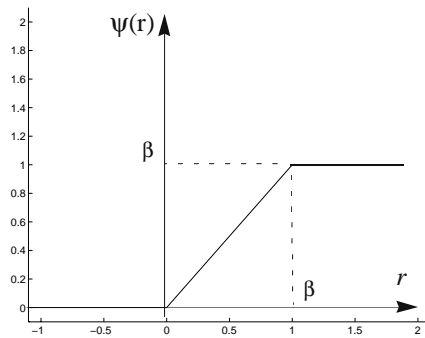
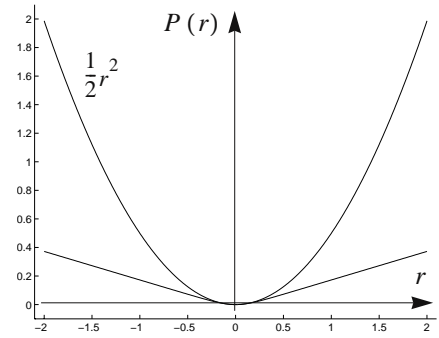
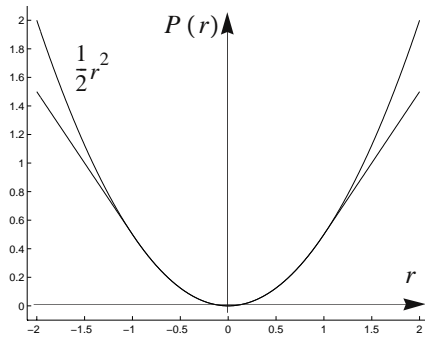


(a)

(b)

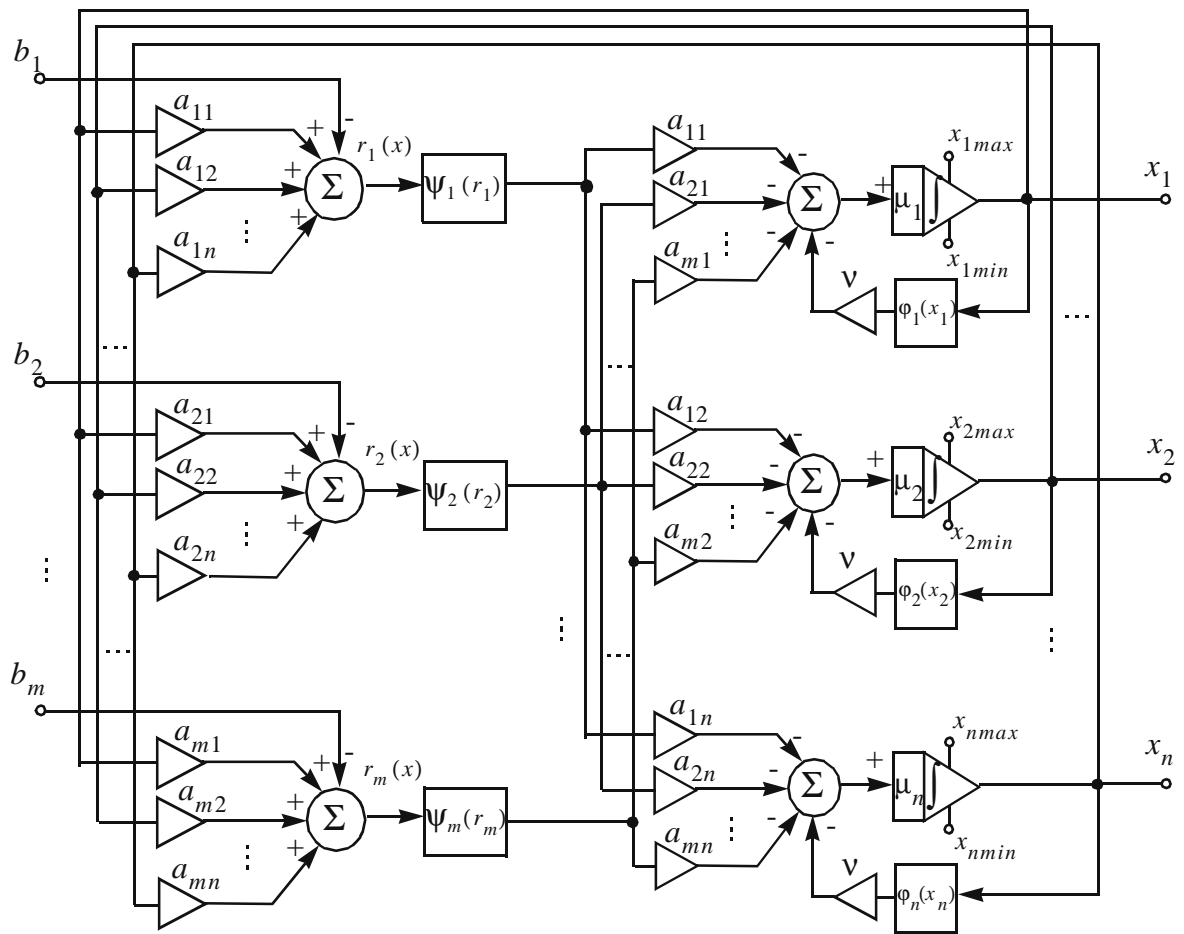


(c)

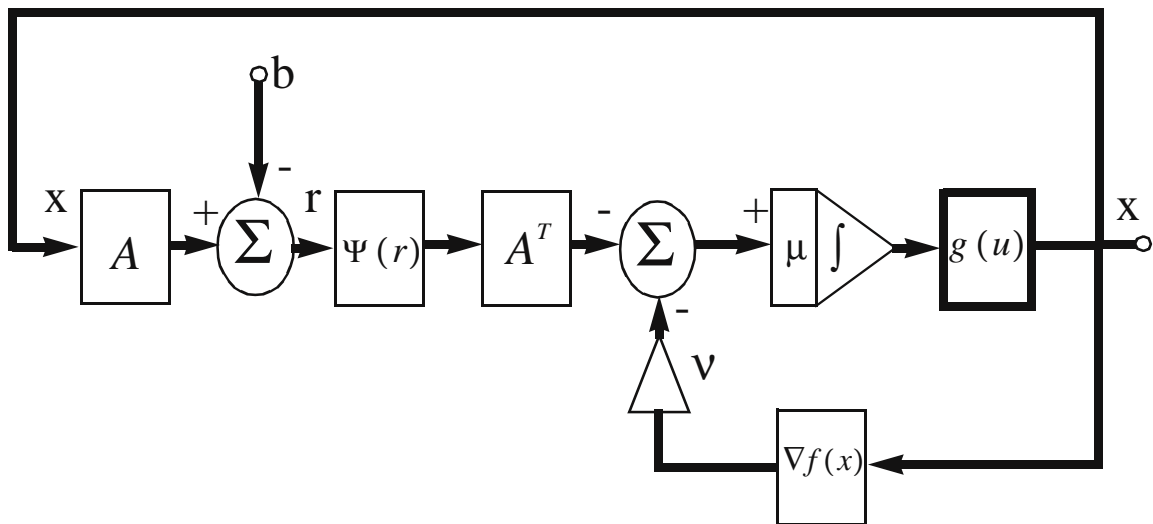
(d)

- (a) Quadratic function
- (b) Absolute values function
- (c) Huber's function
- (d) Logistic function

Figure 1: Penalty functions



(a)



(b)

(a) Detailed neural network  
 (b) Aggregated neural network

Figure 2: Neural network using penalty function approach

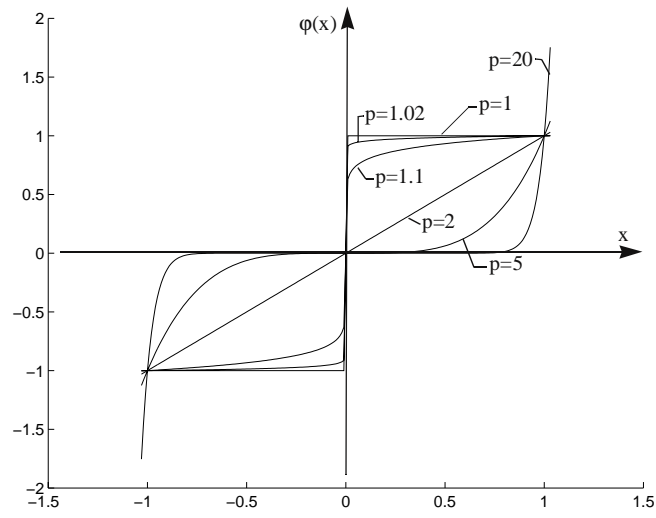


Figure 3: Exemplary plots of the activation function

$$\varphi(x_j) = \frac{\partial}{\partial x_j} f(x) \quad \text{with} \quad f(x) = \frac{1}{p} \|x\|_p^p$$

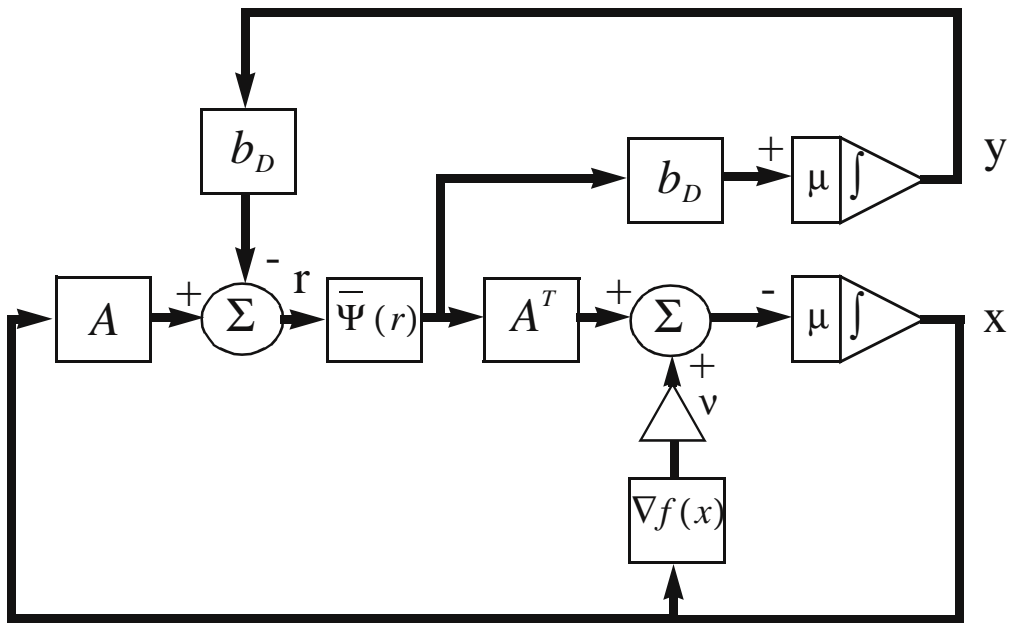
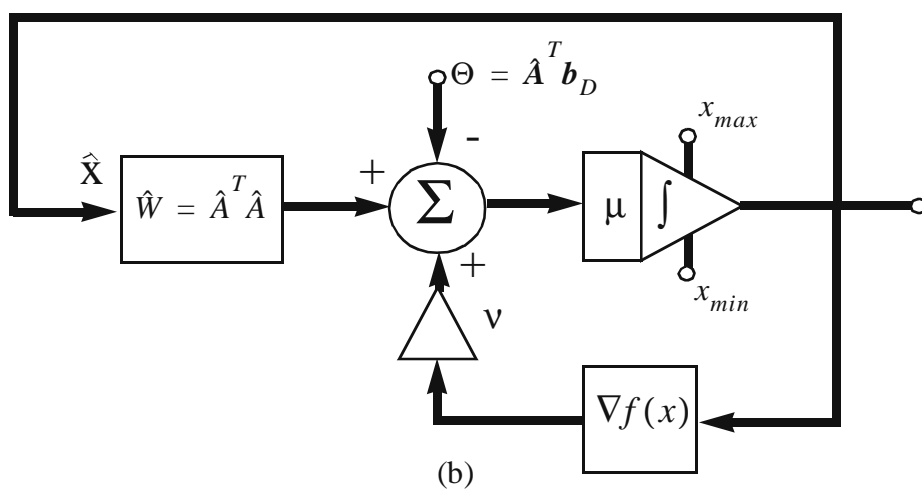
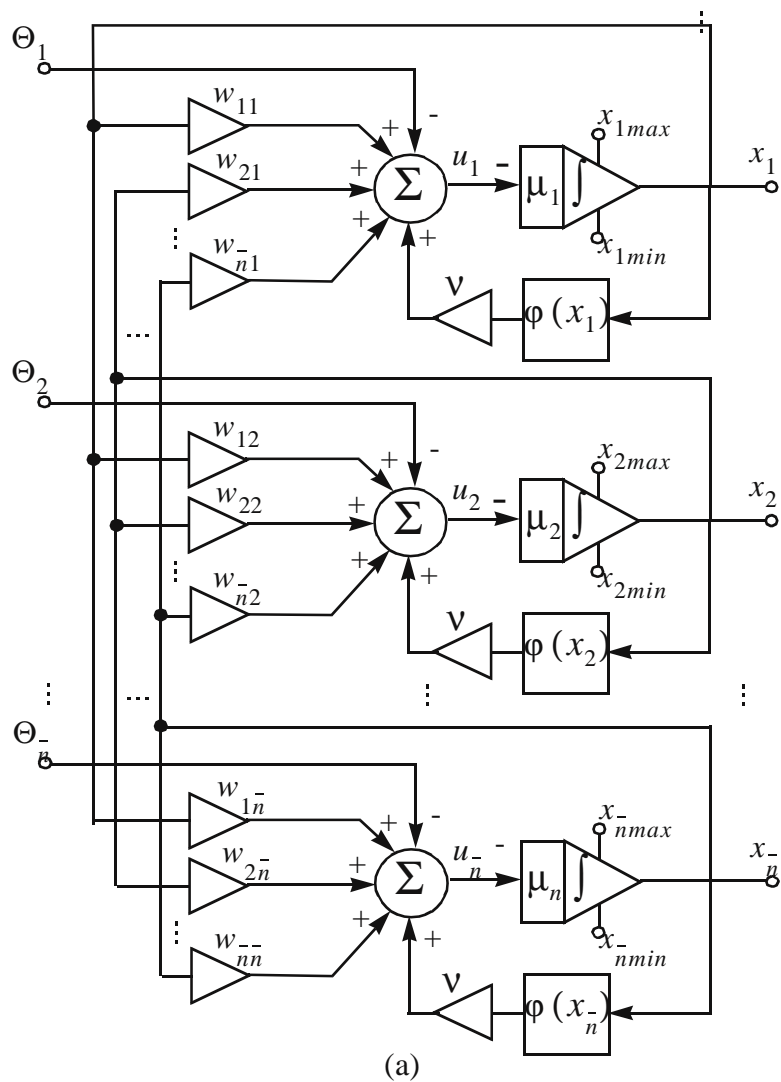


Figure 4: Neural network implementing transformation of inequality constraints into equality constraints



(a) Detailed neural network  
 (b) Aggregated neural network

Figure 5: Simplified neural network implementing transformation of inequality constraints into equality constraints

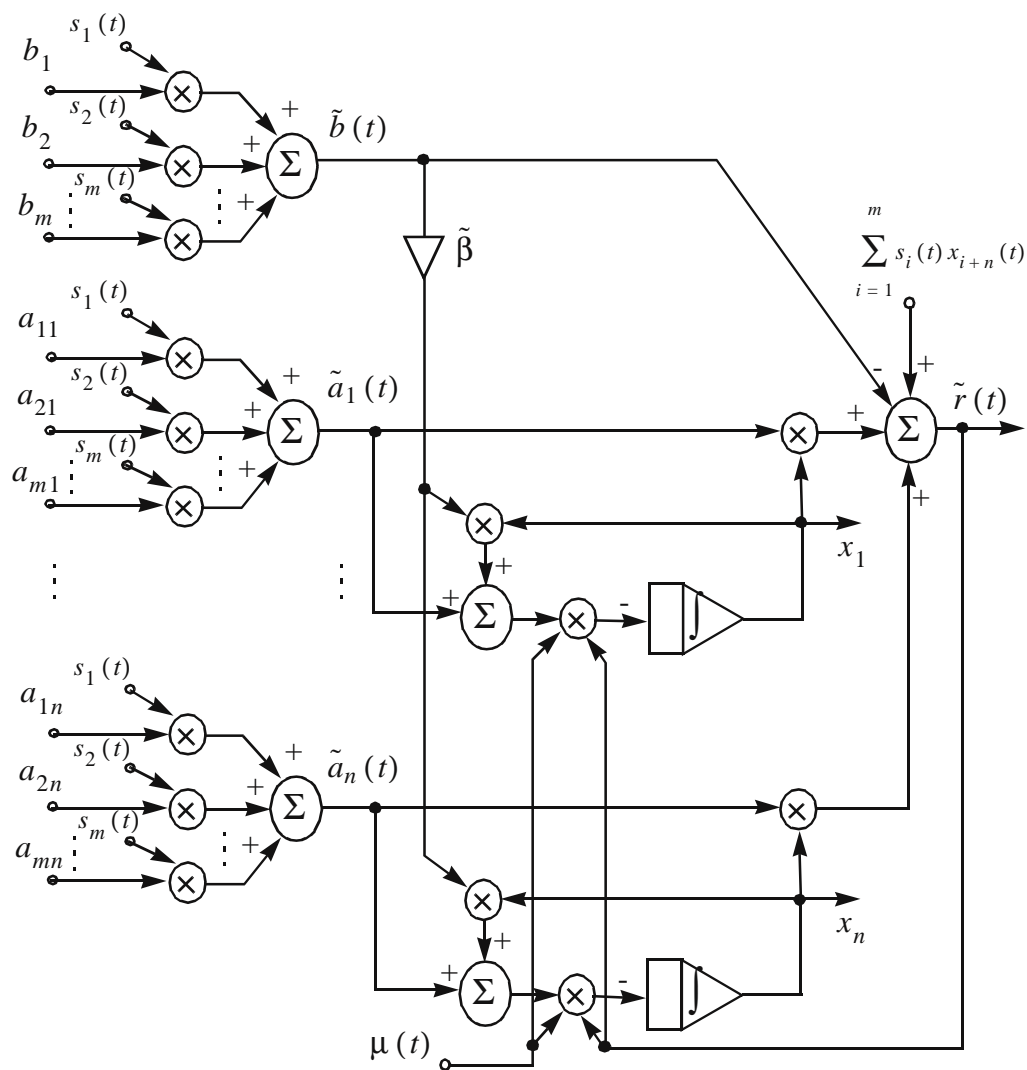
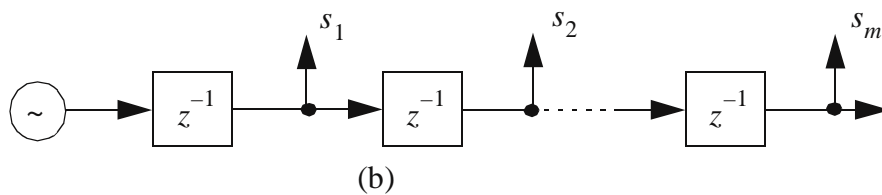
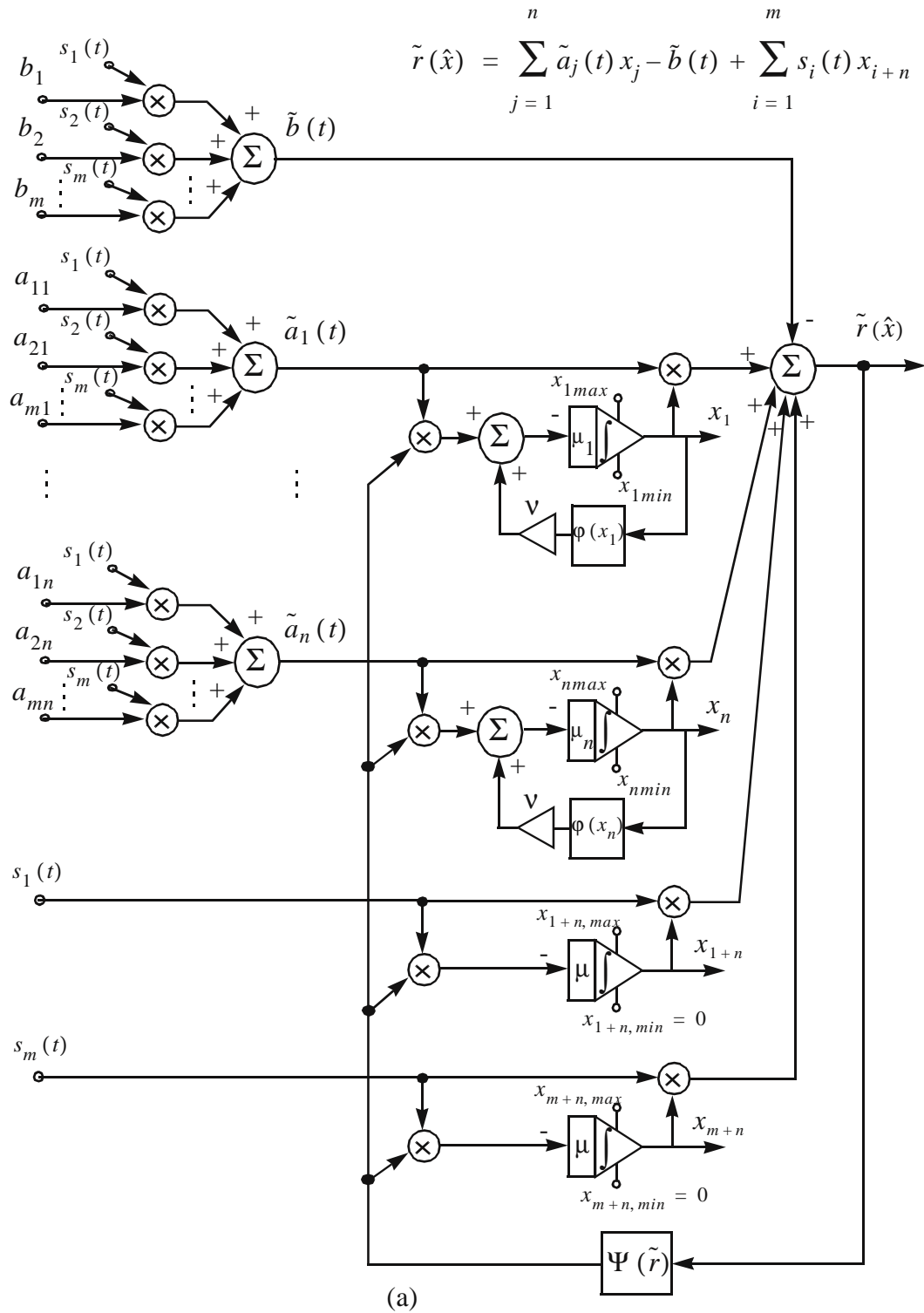


Figure 7: Generalised neural network implementing the regularised total least squares (RTLs) (for  $v = 0$ )



(a) Neuron with adaptive synaptic weights  
 (b) Unit delays

Figure 6: Neural network using random perturbation signal

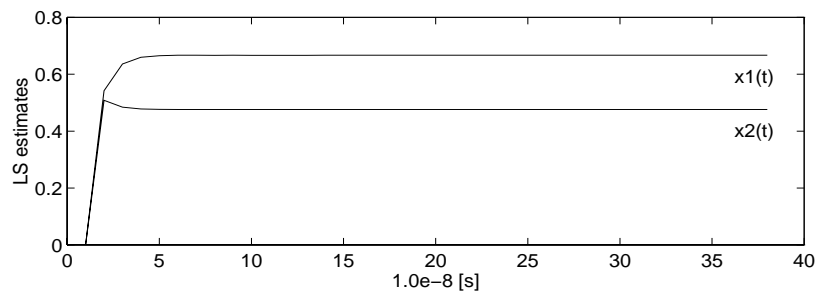
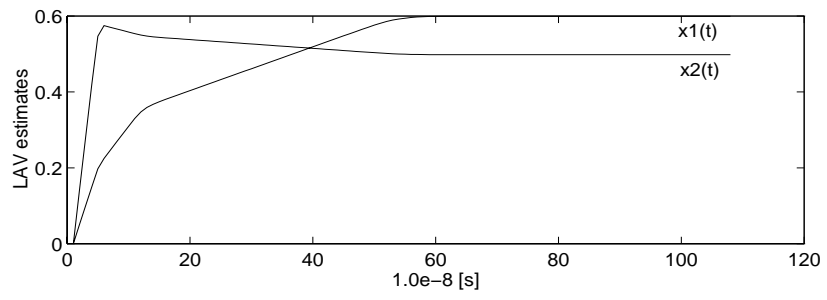


Figure 8: LAV and LS estimates obtained using the neural network of Figure 2

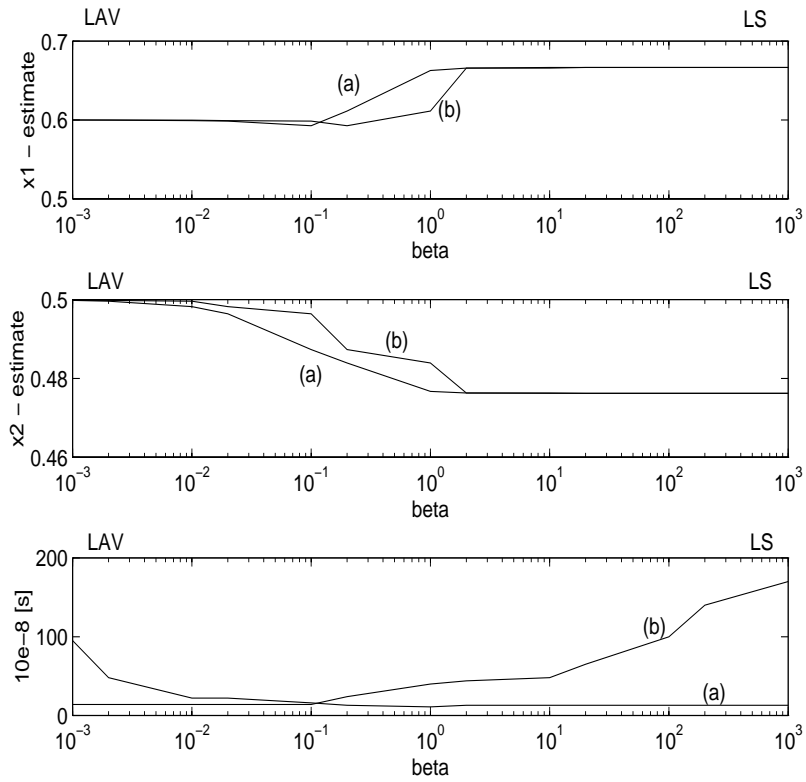
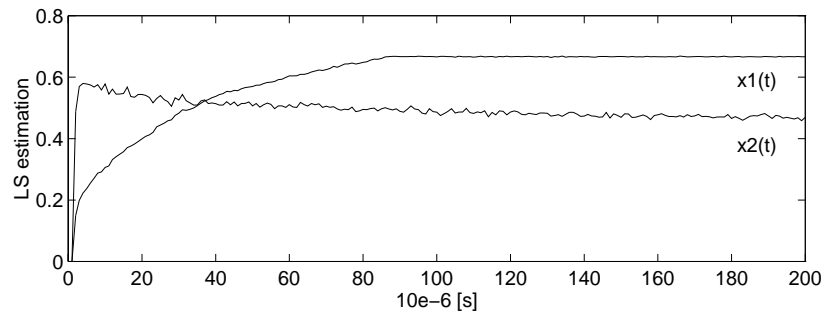
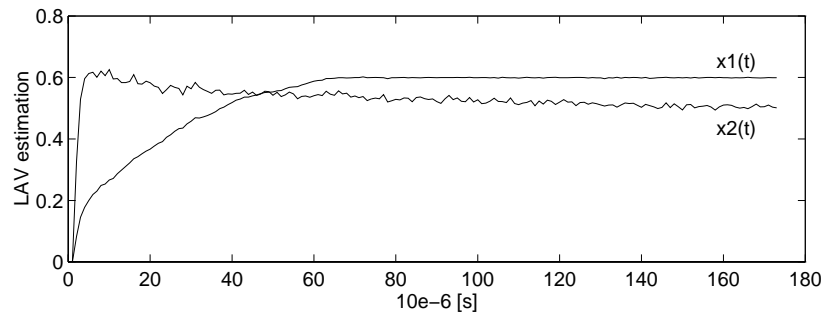


Figure 9: Dependence of the neural network solution on the value of the parameter  $\beta$  in the penalty function and the effect of using different penalty functions (a) - Eq.(8d) and (b) - Eq. (48).



*Figure 10: LAV and LS estimates obtained using the neural network of Figure 6.*

## LIST OF FIGURES

- (a) Quadratic function
- (b) Absolute values function
- (c) Huber's function
- (d) Logistic function

*Figure 1: Penalty functions*

- (a) Detailed neural network
- (b) Aggregated neural network

*Figure 2: Neural network using penalty function approach*

*Figure 3: Exemplary plots of the activation function*

$$\varphi(x_j) = \frac{\partial}{\partial x_j} f(x) \text{ with } f(x) = \frac{1}{p} \|x\|_p^p$$

*Figure 4: Neural network implementing transformation of inequality constraints into equality constraints*

- (a) Detailed neural network
- (b) Aggregated neural network

*Figure 5: Simplified neural network implementing transformation of inequality constraints into equality constraints*

- (a) Neuron with adaptive synaptic weights
- (b) Unit delays

*Figure 6: Neural network using random perturbation signal*

*Figure 7: Generalised network implementing the regularised total least squares (RTLTS)*

*Figure 8: LAV and LS estimates obtained using the neural network of Figure 2*

*Figure 9: Dependence of the neural network solution on the value of the parameter  $\beta$  in the penalty function and the effect of using different penalty functions (a) - Eq. (8d) and (b) - Eq. (48)*

*Figure 10: LAV and LS estimates obtained using the neural network of Figure 6.*