

# Takagi-Sugeno Model Identification Toolbox

March 10, 2021

Automatic static LiP model for an academic example

V1.0

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\$Id: Static\_Acad\_auto.m | Fri Feb 26 16:25:05 2021 +0100 | Axel Dürrbaum \$

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Example of automatic identification of a static MISO LiP TS model for given multiple inputs  $u$  and single output  $y$  with minimal requirements.

Determine the MISO LiP TS model

$$y(u) = \sum_{i=1}^{n_v} \phi_i(z) \cdot \left( \sum_{j=1}^{n_u} B_{i,j} \cdot u_j + c_i \right)$$

- for given vectors  $u_j, j = 1, \dots, n_u$  of  $n_u$  inputs and
- vector  $y$  of single output,
- with FCM membership function

$$\mu_i(x) = \left( \sum_{j=1}^{n_v} \left( \frac{\|z - v_i\|}{\|z - v_j\|} \right)^{\frac{2}{\nu - 1}} \right)^{-1}$$

- or Gaussian membership function

$$\mu_i(z) = e^{-\frac{\|z - v_i\|^2}{2 \cdot \sigma_i^2}}$$

- norm  $\|z - v_j\| = (z - v_j)^T \cdot A_j \cdot (z - v_j)$
- and fuzzy basis functions

$$\phi_i(z) = \frac{\mu_i(z)}{\sum_{j=1}^{n_v} \mu_j(z)}$$

- with the scheduling variable  $z = u$  (for input space clustering) or  $z = [u, y]$  (for product space clustering), and
- cluster centers  $v_i, i = 1, \dots, n_v$ .

## 1 Algorithm

1. Search the best TS model with the minimal MSE for  $n_v = \{2, 3, 4\}$  and  $\nu = \{1.05, 1.1, 1.2, 1.5, 2\}$ .
2. Select the TS model with minimal MSE of  $s$  multi-start tries for clustering and Least Squares estimation.
3. Optimize the TS model parameters  $(v_i, B_i, c_i)$  for each try.

## 2 Minimal required data

Inputs  $u \in \mathbb{R}^{N \times n_u}$  and output  $y \in \mathbb{R}^N$ , each with  $N$  data points

## 3 Identification data

Given is an academic example as a TS model with

- inputs  $u_1, u_2 \in [0, 2]$
- local model matrices

$$B = \begin{pmatrix} -4 & 4 \\ 4 & -2 \\ 2 & 1 \end{pmatrix}, \quad c = \begin{pmatrix} -2 \\ -4 \\ 1 \end{pmatrix}$$

- FCM membership functions ( $\nu = 1.2$ ) with Euclidian norm
- cluster centers

$$v = \begin{pmatrix} 0.5 & 0.5 \\ 0.5 & 1.5 \\ 1.5 & 1 \end{pmatrix}$$

Load data  $u, y$  with  $N = 50$  data-points without noise, generated from this model:

```
load( 'Data/AcadEx.mat' )
```

## 4 Structural parameters

Number of inputs  $n_u$  = number of columns in  $u$

```
Par.nu = size( u, 2 );
```

Number of clusters  $n_v$  = number of local models ( $n_v > 1$ )

```
Par.nv = [ 2, 3, 4 ];
```

Fuzziness parameter (FCM:  $\nu = \{1.05, \dots, 2\}$ , Gauss:  $\sigma_i^2$ )

```
Par.fuzzy = [ 1.05, 1.2, 2.0 ];
```

## 5 Optional settings

For more control over the approximation process.

Multi-Start: number of tries  $s$  (clustering & LS), default = 10

`Par.Tries = 10;`

Clustering: Fuzzy C-Means (FCM) / Gustafson-Kessel (GK) / KMeans (KMeans), default = 'FCM'

`Par.Clustering = 'FCM';`

Clustering in product space:  $u$  and  $y$  (true) or only input space  $u$  (false)

`Par.ProductSpace = true;`

Norm for clustering: 'Euclidian' or 'Mahalanobis', default = 'Euclidian'

`Par.Norm = 'Euclidian';`

Membership functions: 'FCM' or 'Gauss' type clustering

`Par.MSF = 'FCM';`

Least Squares estimation of local models: 'local' or 'global', default = 'global'

`Par.LS = 'global';`

Optimize TS model parameters: default='both'

- no optimization: 'none',
- only  $v$ : 'cluster',
- only local models  $(B_i, c_i)$ : 'model', or
- both  $v$  and  $B_i, c_i$ : 'both'

`Par.ParOpt = 'both';`

Optimize each try or only best try: default='each'

- each try: 'each',
- best try: 'best' (less computation time)

`Par.IterOpt = 'each';`

Plot clusters and residuals: 'none'/'iter'/'final', default='final'

`Par.Plots = 'final';`

Debug infos of algorithm progress: (0=none, 1=info, 2=detailed)

`Par.Debug = 2;`

## 6 Estimation of Static TS model parameters

Estimate the TS model with plot of clustering and correlation:

```
model = TSM_Static_auto( u, y, Par );
```

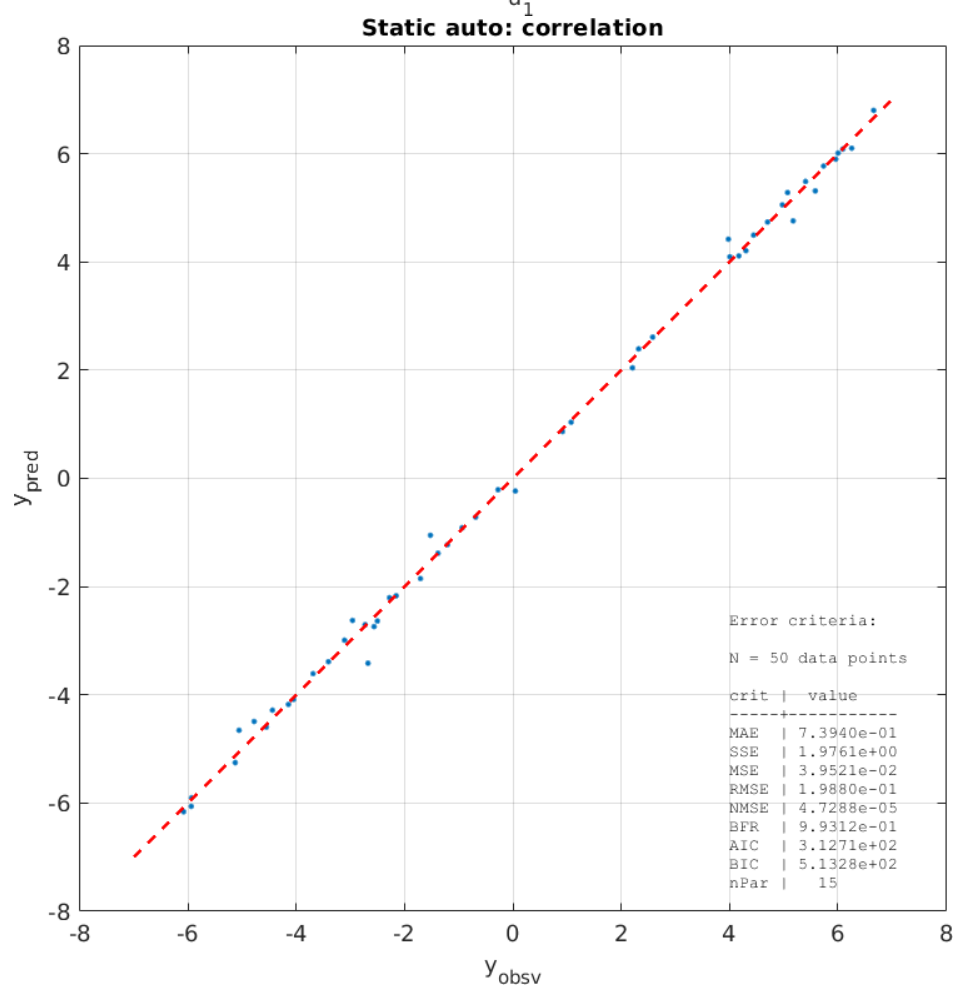
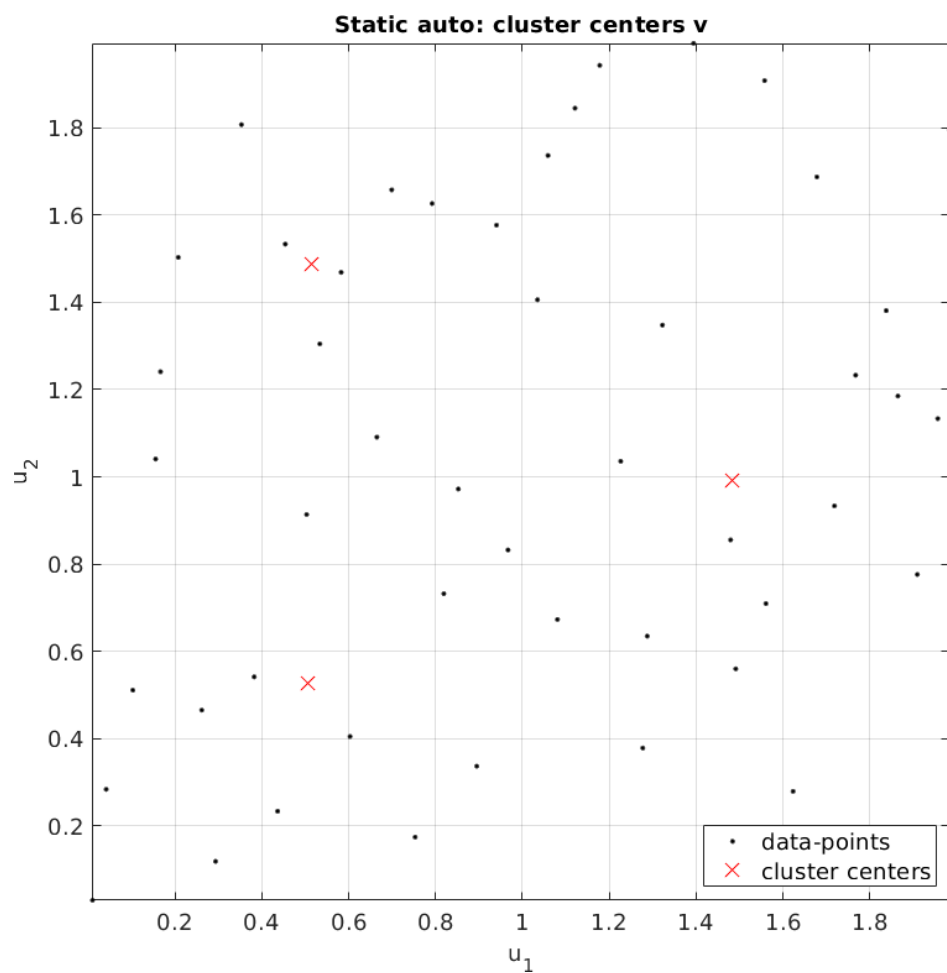
```
Iteration: nv= 2 / fuzzy=1.05
try 1: mse = 1.4333e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
try 2: mse = 1.4334e+00 / delta = +3.7371e-05 (nv= 2/fuzzy=1.05)
try 3: mse = 1.4378e+00 / delta = +4.5137e-03 (nv= 2/fuzzy=1.05)
try 4: mse = 1.4345e+00 / delta = +1.1573e-03 (nv= 2/fuzzy=1.05)
try 5: mse = 1.4349e+00 / delta = +1.5496e-03 (nv= 2/fuzzy=1.05)
try 6: mse = 1.4371e+00 / delta = +3.7533e-03 (nv= 2/fuzzy=1.05)
try 7: mse = 1.4312e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
try 8: mse = 1.4361e+00 / delta = +4.9482e-03 (nv= 2/fuzzy=1.05)
try 9: mse = 1.4376e+00 / delta = +6.4606e-03 (nv= 2/fuzzy=1.05)
try 10: mse = 1.4374e+00 / delta = +6.2157e-03 (nv= 2/fuzzy=1.05)
time = 0.248035 s
Iteration: nv= 2 / fuzzy=1.20
try 1: mse = 1.4835e+00 / delta = +5.2292e-02 (nv= 2/fuzzy=1.05)
try 2: mse = 1.4835e+00 / delta = +5.2292e-02 (nv= 2/fuzzy=1.05)
try 3: mse = 1.4839e+00 / delta = +5.2668e-02 (nv= 2/fuzzy=1.05)
try 4: mse = 1.4839e+00 / delta = +5.2668e-02 (nv= 2/fuzzy=1.05)
try 5: mse = 1.4839e+00 / delta = +5.2668e-02 (nv= 2/fuzzy=1.05)
try 6: mse = 1.4839e+00 / delta = +5.2668e-02 (nv= 2/fuzzy=1.05)
try 7: mse = 1.4839e+00 / delta = +5.2668e-02 (nv= 2/fuzzy=1.05)
try 8: mse = 1.4835e+00 / delta = +5.2292e-02 (nv= 2/fuzzy=1.05)
try 9: mse = 1.4839e+00 / delta = +5.2668e-02 (nv= 2/fuzzy=1.05)
try 10: mse = 1.4835e+00 / delta = +5.2292e-02 (nv= 2/fuzzy=1.05)
time = 0.059478 s
Iteration: nv= 2 / fuzzy=2.00
try 1: mse = 9.9954e-01 / delta = +0.0000e+00 (nv= 2/fuzzy= 2)
try 2: mse = 9.9954e-01 / delta = +2.2275e-11 (nv= 2/fuzzy= 2)
try 3: mse = 9.9954e-01 / delta = +1.6343e-10 (nv= 2/fuzzy= 2)
try 4: mse = 9.9954e-01 / delta = +1.6790e-11 (nv= 2/fuzzy= 2)
try 5: mse = 9.9954e-01 / delta = +1.3873e-10 (nv= 2/fuzzy= 2)
try 6: mse = 9.9954e-01 / delta = +1.7027e-10 (nv= 2/fuzzy= 2)
try 7: mse = 9.9954e-01 / delta = +1.9275e-11 (nv= 2/fuzzy= 2)
try 8: mse = 9.9954e-01 / delta = +2.1154e-11 (nv= 2/fuzzy= 2)
try 9: mse = 9.9954e-01 / delta = +1.5607e-10 (nv= 2/fuzzy= 2)
try 10: mse = 9.9954e-01 / delta = +1.4735e-11 (nv= 2/fuzzy= 2)
time = 0.048684 s
Iteration: nv= 3 / fuzzy=1.05
try 1: mse = 7.5872e-02 / delta = +0.0000e+00 (nv= 3/fuzzy=1.05)
try 2: mse = 7.5872e-02 / delta = +0.0000e+00 (nv= 3/fuzzy=1.05)
try 3: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 4: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 5: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 6: mse = 7.5872e-02 / delta = +4.3715e-15 (nv= 3/fuzzy=1.05)
try 7: mse = 7.5872e-02 / delta = +3.0623e-11 (nv= 3/fuzzy=1.05)
try 8: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 9: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 10: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
time = 0.112398 s
Iteration: nv= 3 / fuzzy=1.20
try 1: mse = 3.9521e-02 / delta = +0.0000e+00 (nv= 3/fuzzy= 1.2)
try 2: mse = 3.9521e-02 / delta = +6.1687e-15 (nv= 3/fuzzy= 1.2)
try 3: mse = 3.9521e-02 / delta = +0.0000e+00 (nv= 3/fuzzy= 1.2)
try 4: mse = 3.9521e-02 / delta = +0.0000e+00 (nv= 3/fuzzy= 1.2)
try 5: mse = 3.9521e-02 / delta = +8.5924e-13 (nv= 3/fuzzy= 1.2)
```

```

try 6: mse = 3.9521e-02 / delta = +8.2142e-13 (nv= 3/fuzzy= 1.2)
try 7: mse = 3.9521e-02 / delta = +0.0000e+00 (nv= 3/fuzzy= 1.2)
try 8: mse = 3.9521e-02 / delta = +0.0000e+00 (nv= 3/fuzzy= 1.2)
try 9: mse = 3.9521e-02 / delta = +0.0000e+00 (nv= 3/fuzzy= 1.2)
try 10: mse = 3.9521e-02 / delta = +1.7788e-13 (nv= 3/fuzzy= 1.2)
time = 0.038863 s
Iteration: nv= 3 / fuzzy=2.00
try 1: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 2: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 3: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 4: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 5: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 6: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 7: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 8: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 9: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
try 10: mse = 5.3089e-01 / delta = +4.9137e-01 (nv= 3/fuzzy= 1.2)
time = 0.094955 s
Iteration: nv= 4 / fuzzy=1.05
try 1: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 2: mse = 1.5795e-01 / delta = +1.1843e-01 (nv= 3/fuzzy= 1.2)
try 3: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 4: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 5: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 6: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 7: mse = 1.5795e-01 / delta = +1.1843e-01 (nv= 3/fuzzy= 1.2)
try 8: mse = 8.0882e-02 / delta = +4.1361e-02 (nv= 3/fuzzy= 1.2)
try 9: mse = 1.5795e-01 / delta = +1.1843e-01 (nv= 3/fuzzy= 1.2)
try 10: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
time = 0.234904 s
Iteration: nv= 4 / fuzzy=1.20
try 1: mse = 3.9728e-02 / delta = +2.0712e-04 (nv= 3/fuzzy= 1.2)
try 2: mse = 3.9729e-02 / delta = +2.0716e-04 (nv= 3/fuzzy= 1.2)
try 3: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 4: mse = 3.9729e-02 / delta = +2.0716e-04 (nv= 3/fuzzy= 1.2)
try 5: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 6: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 7: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 8: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 9: mse = 3.9729e-02 / delta = +2.0715e-04 (nv= 3/fuzzy= 1.2)
try 10: mse = 3.9728e-02 / delta = +2.0712e-04 (nv= 3/fuzzy= 1.2)
time = 0.157643 s
Iteration: nv= 4 / fuzzy=2.00
try 1: mse = 4.8537e-01 / delta = +4.4585e-01 (nv= 3/fuzzy= 1.2)
try 2: mse = 4.8545e-01 / delta = +4.4593e-01 (nv= 3/fuzzy= 1.2)
try 3: mse = 4.8537e-01 / delta = +4.4585e-01 (nv= 3/fuzzy= 1.2)
try 4: mse = 4.8538e-01 / delta = +4.4586e-01 (nv= 3/fuzzy= 1.2)
try 5: mse = 4.8538e-01 / delta = +4.4586e-01 (nv= 3/fuzzy= 1.2)
try 6: mse = 4.8522e-01 / delta = +4.4570e-01 (nv= 3/fuzzy= 1.2)
try 7: mse = 4.8547e-01 / delta = +4.4595e-01 (nv= 3/fuzzy= 1.2)
try 8: mse = 4.8547e-01 / delta = +4.4594e-01 (nv= 3/fuzzy= 1.2)
try 9: mse = 4.8537e-01 / delta = +4.4585e-01 (nv= 3/fuzzy= 1.2)
try 10: mse = 4.8490e-01 / delta = +4.4538e-01 (nv= 3/fuzzy= 1.2)
time = 0.6712 s

Best model: nv= 3 / fuzzy=1.20 / mse = 3.9521e-02

```

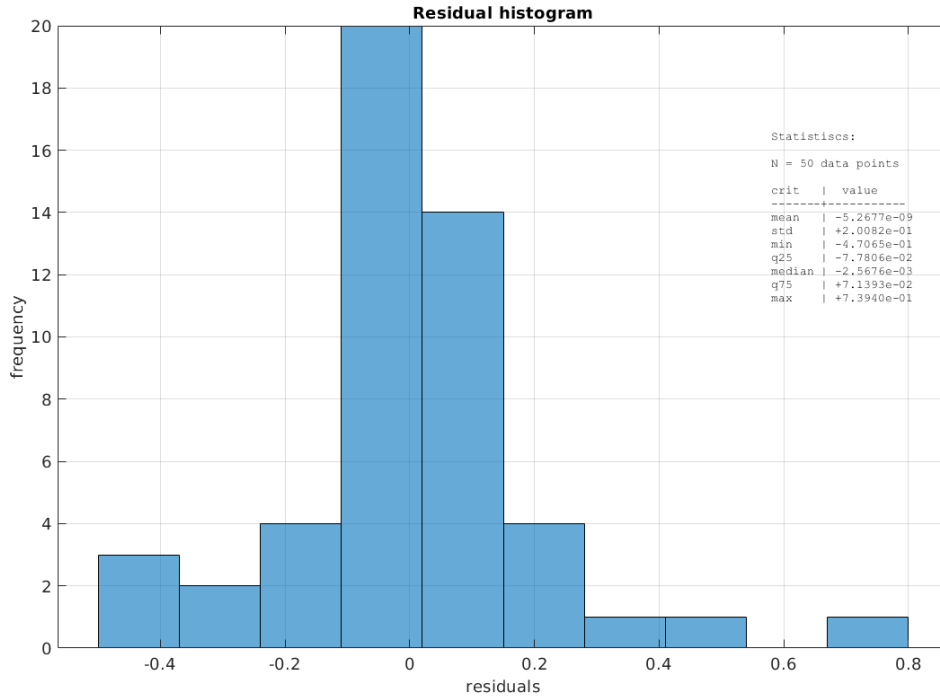


Predict the model output  $y_{\text{pred}}$  for input  $u$ :

```
y_pred = model.predict( u, y );
```

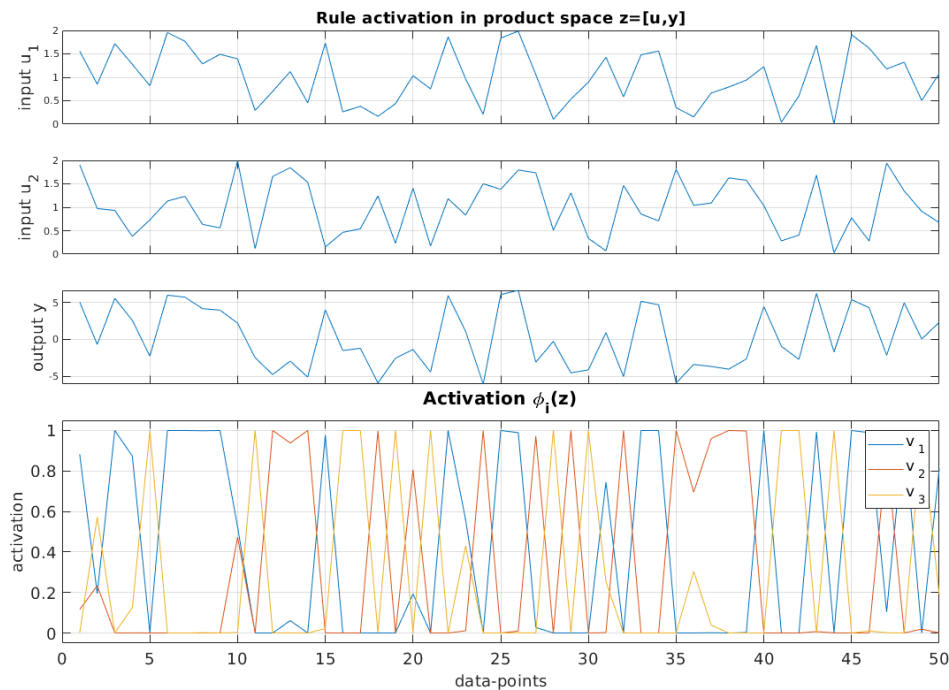
Plot a residual histogram:

```
hr = plotResidualHist( y, y_pred, 'figure', 3 );
```



Plot the rule activation and input/output data:

```
ha = plotRuleActivation( u,y,model, 'figure', 4 );
```



## 7 Retrieve the parameters of the final TS model

Show the TS model parameters:

```
disp( model )
```

```
TS-Model: Type=Static
Name: 'undefined'
Type: 'TSModel'
Date: '10-Mar-2021 11:53:33'
Comments:
  'created by TSM_static_auto'
Structural parameters: nu = 2, ny = 1, nv = 3
Identification data: N=50
Initial model estimation:
  Clustering: FCM, nue=1.2 norm=Euclidian
Estimation of local models:
  Initialization of local models: global
  Optimization of model parameters: MF&LM
```

Show the cluster centers  $v$  ( $n_v$  rows and  $n_u$  columns):

```
v = getCluster( model )
```

```
v =
```

```
    1.4815    0.9923
    0.5135    1.4882
    0.5059    0.5262
```

Show the local model matrices  $B_i$  and  $c_i$  ( $n_v$  rows and  $n_u$  columns):

```
[~,B,c] = getLM( model )
```

```
B =
```

```
    1.9182    1.2177
    3.8744   -1.5237
   -4.0956    4.1970
```

```
c =
```

```
    0.8801
   -4.6761
   -1.9318
```