# Takagi-Sugeno Model Identification Toolbox

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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  $u_j, j = 1, \ldots, n_u$  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  $v_i, i = 1, \ldots, n_v$ .

## 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 Euclidean 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

Number of clusters  $n_v = \text{number of local models } (n_v > 1)$ 

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

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

## 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: 'Euclidean' or 'Mahalanobis', default = 'Euclidean'
Par.Norm = 'Euclidean';
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 progess: (0=none, 1=info, 2=detailed)
Par.Debug = 2;
```

### 6 Estimation of Static TS model parameters

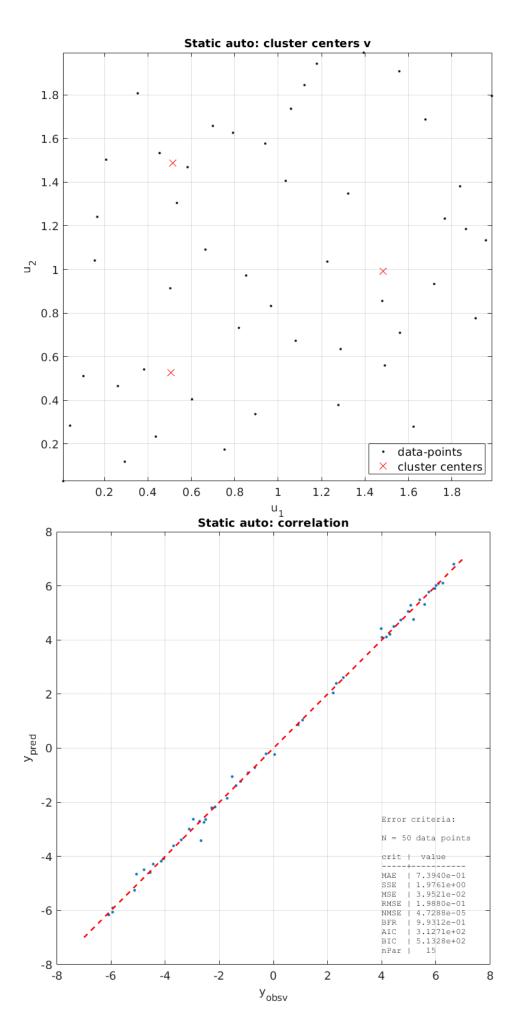
Estimate the TS model with plot of clustering and correlation:

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

```
Iteration: nv= 2 / fuzzy=1.05
try 1: mse = 1.4342e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
try 2: mse = 1.4334e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
try 3: mse = 1.4293e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
try 4: mse = 1.4403e+00 / delta = +1.0967e-02 (nv= 2/fuzzy=1.05)
try 5: mse = 1.4348e+00 / delta = +5.4669e-03 (nv= 2/fuzzy=1.05)
try 6: mse = 1.4335e+00 / delta = +4.1426e-03 (nv= 2/fuzzy=1.05)
try 7: mse = 1.4292e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
     8: mse = 1.4378e+00 / delta = +8.6055e-03 (nv= 2/fuzzy=1.05)
try 9: mse = 1.4336e+00 / delta = +4.3855e-03 (nv= 2/fuzzy=1.05)
try 10: mse = 1.4253e+00 / delta = +0.0000e+00 (nv= 2/fuzzy=1.05)
time = 0.252519 s
Iteration: nv= 2 / fuzzy=1.20
try 1: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 2: mse = 1.4835e+00 / delta = +5.8163e-02 (nv= 2/fuzzy=1.05)
try 3: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 4: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 5: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 6: mse = 1.4835e+00 / delta = +5.8163e-02 (nv= 2/fuzzy=1.05)
try 7: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 8: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 9: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
try 10: mse = 1.4839e+00 / delta = +5.8539e-02 (nv= 2/fuzzy=1.05)
time = 0.059033 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 = +0.0000e+00 (nv= 2/fuzzy=
                                                               2)
try 3: mse = 9.9954e-01 / delta = +1.2729e-10 (nv= 2/fuzzy=
try 4: mse = 9.9954e-01 / delta = +1.2596e-10 (nv= 2/fuzzy=
try 5: mse = 9.9954e-01 / delta = +9.5490e-12 (nv= 2/fuzzy=
try 6: mse = 9.9954e-01 / delta = +0.0000e+00 (nv= 2/fuzzy=
try 7: mse = 9.9954e-01 / delta = +1.1290e-11 (nv= 2/fuzzy=
try 8: mse = 9.9954e-01 / delta = +7.5993e-12 (nv= 2/fuzzy=
                                                               2)
try 9: mse = 9.9954e-01 / delta = +1.3306e-11 (nv= 2/fuzzy=
                                                               2)
try 10: mse = 9.9954e-01 / delta = +1.2980e-10 (nv= 2/fuzzy=
time = 0.051883 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 = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 3: mse = 7.5872e-02 / delta = +0.0000e+00 (nv= 3/fuzzy=1.05)
try 4: mse = 7.6107e-02 / delta = +2.3485e-04 (nv= 3/fuzzy=1.05)
try 5: mse = 7.5872e-02 / delta = +0.0000e+00 (nv= 3/fuzzy=1.05)
try 6: mse = 7.5872e-02 / delta = +1.7555e-14 (nv= 3/fuzzy=1.05)
try 7: mse = 7.5872e-02 / delta = +3.0593e-11 (nv= 3/fuzzy=1.05)
try 8: mse = 7.5872e-02 / delta = +3.0633e-11 (nv= 3/fuzzy=1.05)
try 9: mse = 7.5872e-02 / delta = +3.0633e-11 (nv= 3/fuzzy=1.05)
try 10: mse = 7.5872e-02 / delta = +8.0908e-15 (nv= 3/fuzzy=1.05)
time = 0.103849 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 = +2.1647e-13 (nv= 3/fuzzy= 1.2)
try 3: mse = 3.9521e-02 / delta = +1.6694e-13 (nv= 3/fuzzy= 1.2)
try 4: mse = 3.9521e-02 / delta = +1.7848e-13 (nv= 3/fuzzy= 1.2)
try 5: mse = 3.9521e-02 / delta = +2.7951e-13 (nv= 3/fuzzy= 1.2)
```

```
try 6: mse = 3.9521e-02 / delta = +4.7176e-12 (nv= 3/fuzzy= 1.2)
try 7: mse = 3.9521e-02 / delta = +4.8312e-12 (nv= 3/fuzzy= 1.2)
try 8: mse = 3.9521e-02 / delta = +1.3433e-13 (nv= 3/fuzzy= 1.2)
try 9: mse = 3.9521e-02 / delta = +4.8571e-12 (nv= 3/fuzzy= 1.2)
try 10: mse = 3.9521e-02 / delta = +4.9304e-12 (nv= 3/fuzzy= 1.2)
time = 0.034243 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= <math>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 = 0.087966 s
Iteration: nv= 4 / fuzzy=1.05
try 1: mse = 8.3754e-02 / delta = +4.4233e-02 (nv= 3/fuzzy= 1.2)
try 2: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 3: mse = 1.5795e-01 / delta = +1.1843e-01 (nv= 3/fuzzy= 1.2)
try 4: mse = 8.0152e-02 / delta = +4.0631e-02 (nv= 3/fuzzy= 1.2)
try 5: mse = 7.9655e-02 / delta = +4.0134e-02 (nv= 3/fuzzy= 1.2)
try 6: mse = 1.5795e-01 / delta = +1.1843e-01 (nv= 3/fuzzy= 1.2)
try 7: mse = 2.1947e-01 / delta = +1.7995e-01 (nv= 3/fuzzy= 1.2)
try 8: mse = 1.5795e-01 / delta = +1.1843e-01 (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 = 0.276368 s
Iteration: nv= 4 / fuzzy=1.20
try 1: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 2: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 3: mse = 3.9728e-02 / delta = +2.0712e-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 = +2.0712e-04 (nv= 3/fuzzy= 1.2)
try 9: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
try 10: mse = 3.9728e-02 / delta = +2.0706e-04 (nv= 3/fuzzy= 1.2)
time = 0.199125 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 = 4.8522e-01 / delta = +4.4570e-01 (nv= 3/fuzzy= 1.2)
try 3: mse = 4.8493e-01 / delta = +4.4541e-01 (nv= 3/fuzzy= 1.2)
try 4: mse = 4.8538e-01 / delta = +4.4586e-01 (nv= 3/fuzzy= 1.2)
try 5: mse = 4.8522e-01 / delta = +4.4570e-01 (nv= 3/fuzzy= 1.2)
try 6: mse = 4.8538e-01 / delta = +4.4586e-01 (nv= 3/fuzzy= 1.2)
try 7: mse = 4.8538e-01 / delta = +4.4586e-01 (nv= 3/fuzzy= 1.2)
try 8: mse = 4.8492e-01 / delta = +4.4540e-01 (nv= 3/fuzzy= 1.2)
try 9: mse = 4.8522e-01 / delta = +4.4570e-01 (nv= 3/fuzzy= 1.2)
try 10: mse = 4.8553e-01 / delta = +4.4601e-01 (nv= 3/fuzzy= 1.2)
time = 0.58829 s
```

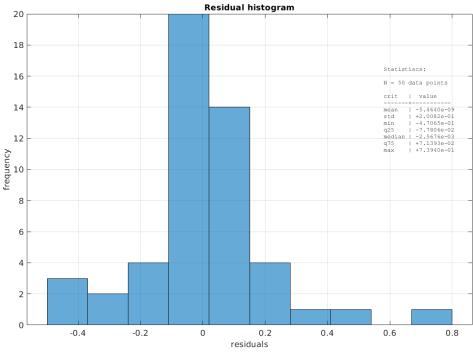
Best model: nv= 3 / fuzzy=1.20 / mse = 3.9521e-02



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

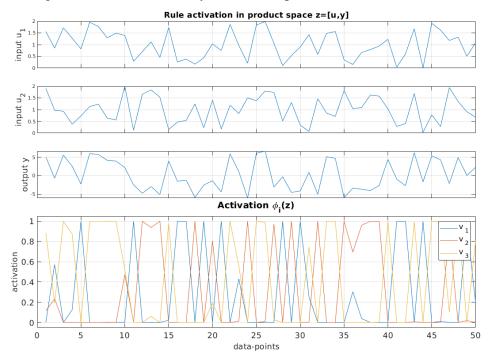
Plot a residual histogram:

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: '29-Mar-2021 15:42:12'
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=Euclidean in product space
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 )
```

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

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

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

```
B =
  -4.0956
          4.1970
          -1.5237
   3.8744
   1.9182
          1.2177
C =
  -1.9318
  -4.6761
   0.8801
```