

Calculation of estimates of measurand for example 3A

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1 Introduction

This document describes the information that must be provided by the user to define the sensor responses to the measurand in the example contained in MATLAB script `DataFusionSoftware_3A.m`.

In particular, section 2 contains a list of all the numerical input that the user is expected to provide. For users who are interested, section 3 lists previously calculated information that is used when determining the sensor responses, while sections 4 and 5 provide full mathematical descriptions of how the sensor response values and estimates of the measurand (and associated uncertainties) are calculated, respectively, by the function `LinearCalibrationLag.m`.

2 User input

The responses of the sensors to the measurand are defined by the following information provided by the user:

Sampling and quantization

- f_2 , the sampling frequency [cell B10],
- (optional) n_B , the number of bits for quantization [cell B11],
- (optional) s , the saturation value [cell B12].

It is assumed that $f_2 < f_1$, i.e., the sampling frequency for the sensors is less than the sampling frequency used when calculating the ‘true’ measurand (section 3).

Behaviour of sensors

- n_S , the number of sensors [cell B9 – filled in automatically from information in column C],
- A_{l_S} , $u_r(A_{l_S})$, $l_S = 1, \dots, n_S$, the estimates of the offset parameter and their associated relative standard uncertainties [columns D and G],
- B_{l_S} , $u_r(B_{l_S})$, $l_S = 1, \dots, n_S$, the estimates of the gain parameter and their associated relative standard uncertainties [columns E and H],
- L_{l_S} , $u_r(L_{l_S})$, $l_S = 1, \dots, n_S$, the estimates of the time lag and their associated relative standard uncertainties [columns F and J],
- $\text{cov}_r(A_{l_S}, B_{l_S})$, $l_S = 1, \dots, n_S$, the relative covariances associated with the estimates of the offset and gain parameters [column I],
- δ_{l_S} , $l_S = 1, \dots, n_S$, the levels of additive noise in the sensor output [column K].

Behaviour of faulty sensors

- n_F , the number of faulty sensors [cell N9 – filled in automatically from information in column O],
- I_{F,l_F} , $l_F = 1, \dots, n_F$, the indices of the faulty sensors [column O],
- C_{l_F} , $u_r(C_{l_F})$, $l_F = 1, \dots, n_F$, the estimates of the offset parameter and their associated relative standard uncertainties [columns P and S],
- D_{l_F} , $u_r(D_{l_F})$, $l_F = 1, \dots, n_F$, the estimates of the gain parameter and their associated relative standard uncertainties [columns Q and T],
- M_{l_F} , $u_r(M_{l_F})$, $l_F = 1, \dots, n_F$, the estimates of the time lag and their associated relative standard uncertainties [columns R and V],
- $\text{cov}_r(C_{l_F}, D_{l_F})$, $l_F = 1, \dots, n_F$, the relative covariances associated with the estimates of the offset and gain parameters [column U],
- T_{F,l_F}^1 , $l_F = 1, \dots, n_F$, the times at which sensors begin behaving faultily [column W],
- T_{F,l_F}^2 , $l_F = 1, \dots, n_F$, the times at which sensors end behaving faultily [column X].

Missing data

- n_M , the number of sensors for which packets of data are missing [cell AA9 – filled in automatically from information in column AB],
- n_D , the number of data points in each packet [cell AA10],
- I_{M,l_M} , $l_M = 1, \dots, n_M$, the indices of the sensors for which packets of data are missing [column AB],

- p_{M,l_M} , $l_M = 1, \dots, n_M$, the proportion (expressed as a percentage) of the data packets that are missing within the assigned time interval [column AC],
- T_{M,l_M}^1 , $l_M = 1, \dots, n_M$, the times at which sensors begin missing packets of data [column AD],
- T_{M,l_M}^2 , $l_M = 1, \dots, n_M$, the times at which sensors cease missing packets of data [column AE].

3 Additional input

The following information, previously calculated, is used:

- t_{1,i_1} , $i_1 = 1, \dots, m_1$, the times at which the ‘true’ values of the measurand are calculated (corresponding to the sampling frequency f_1),
- y_{1,i_1} , $i_1 = 1, \dots, m_1$, the ‘true’ values of the measurand.

4 Sensor response values

The array $\tilde{\mathbf{V}}_2$ of sensor response values is given by

$$\tilde{\mathbf{V}}_2 = [\tilde{\mathbf{v}}_{2,1} \quad \dots \quad \tilde{\mathbf{v}}_{2,n_S}],$$

where

$$\tilde{\mathbf{v}}_{2,l_S} = \begin{bmatrix} \tilde{v}_{2,l_S,1} \\ \vdots \\ \tilde{v}_{2,l_S,m_2} \end{bmatrix}$$

is the vector of response values for sensor l_S and is obtained as follows:

1. Draw the time lag $L_{l_S}^*$ from the rectangular distribution

$$\text{R} \left(L_{l_S} - \frac{\sqrt{3}u_r(L_{l_S})L_{l_S}}{100}, L_{l_S} + \frac{\sqrt{3}u_r(L_{l_S})L_{l_S}}{100} \right).$$

$L_{l_S}^*$ defines a number $q_{l_S}^*$ of values of the sampled ‘true’ measurand, where $q_{l_S}^*$ is the smallest integer satisfying

$$q_{l_S}^* \geq f_1 L_{l_S}^*.$$

Generate the vector \mathbf{y}_{1,l_S}^L of time-shifted ‘true’ measurand values given by

$$y_{1,l_S,i_1}^L = \begin{cases} \text{NaN}, & 1 \leq i_1 \leq q_{l_S}^*, \\ y_{1,i_1-q_{l_S}^*}, & q_{l_S}^* + 1 \leq i_1 \leq m_1, \end{cases}$$

where NaN denotes ‘Not a Number’.

Evaluate the vector of sensor responses

$$\mathbf{v}_{1,l_S} = A_{l_S}^{(2)*} + B_{l_S}^{(2)*} \mathbf{y}_{1,l_S}^L + \mathbf{r}_{l_S},$$

corresponding to the times t_{1,i_1} , $i_1 = 1, \dots, m_1$, where

$$\begin{bmatrix} A_{l_S}^{(2)*} \\ B_{l_S}^{(2)*} \end{bmatrix} \sim N \left(\begin{bmatrix} A_{l_S} \\ B_{l_S} \end{bmatrix}, \boldsymbol{\Sigma}_{l_S} \right),$$

with

$$\boldsymbol{\Sigma}_{l_S} = \begin{bmatrix} \left(\frac{u_r(A_{l_S})A_{l_S}}{100} \right)^2 & \frac{\text{cov}_r(A_{l_S}, B_{l_S})A_{l_S}B_{l_S}}{100} \\ \frac{\text{cov}_r(A_{l_S}, B_{l_S})A_{l_S}B_{l_S}}{100} & \left(\frac{u_r(B_{l_S})B_{l_S}}{100} \right)^2 \end{bmatrix},$$

and

$$\mathbf{r}_{l_S} = \begin{bmatrix} r_{l_S,1} \\ \vdots \\ r_{l_S,m_1} \end{bmatrix},$$

with

$$r_{l_S,i_1} \sim N(0, (\delta_{l_S})^2), \quad i_1 = 1, \dots, m_1.$$

2. For sensors that are faulty, the response values within the time intervals $[T_{F,l_F}^1, T_{F,l_F}^2]$ are obtained similarly to those in step 1, but using the parameters C_{l_F} , $u_r(C_{l_F})$, D_{l_F} , $u_r(D_{l_F})$, $\text{cov}_r(C_{l_F}, D_{l_F})$, M_{l_F} and $u_r(M_{l_F})$, $l_F = 1, \dots, n_F$.
3. Determine the times t_{2,i_2} , $i_2 = 1, \dots, m_2$, at which the sensor response values are to be evaluated.
4. Evaluate the sensor response values v_{2,l_S,i_2} corresponding to the times t_{2,i_2} , $i_2 = 1, \dots, m_2$, by applying linear interpolation to the sensor responses v_{1,l_S,i_1} corresponding to the times t_{1,i_1} , $i_1 = 1, \dots, m_1$.

5. The sensor response values v_{2,l_S,i_2} , $i_2 = 1, \dots, m_2$, are then quantized according to the values of n_B and s (if present) to give values

$$\tilde{v}_{2,l_S,i_2}, \quad i_2 = 1, \dots, m_2.$$

If no values have been provided for n_B and s , then

$$\tilde{v}_{2,l_S,i_2} = v_{2,l_S,i_2}, \quad i_2 = 1, \dots, m_2.$$

6. For sensors that have missing data, the proportions p_{M,l_M} of data packets (a data packet is defined to be a group of data points that are sequential in time) chosen randomly within the time intervals $[T_{M,l_M}^1, T_{M,l_M}^2]$ have their values set to NaN, $l_M = 1, \dots, n_M$.

5 Measurand estimates and associated uncertainties

Let q_{l_S} be the smallest integer satisfying

$$q_{l_S} \geq f_2 L_{l_S}.$$

The array \mathbf{Y}_2 of measurand estimates is given by

$$\mathbf{Y}_2 = \begin{bmatrix} y_{2,1,1} & \cdots & y_{2,n_S,1} \\ \vdots & \ddots & \vdots \\ y_{2,1,m_2} & \cdots & y_{2,n_S,m_2} \end{bmatrix},$$

where

$$y_{2,l_S,i_2} = \begin{cases} \text{NaN}, & 1 \leq i_2 \leq q_{l_S}, \\ \frac{\tilde{v}_{2,l_S,i_2} - A_{l_S}}{B_{l_S}}, & q_{l_S} + 1 \leq i_2 \leq m_2. \end{cases}$$

The array \mathbf{U}_2 of standard uncertainties associated with the measurand estimates is given by

$$\mathbf{U}_2 = \begin{bmatrix} u(y_{2,1,1}) & \cdots & u(y_{2,n_S,1}) \\ \vdots & \ddots & \vdots \\ u(y_{2,1,m_2}) & \cdots & u(y_{2,n_S,m_2}) \end{bmatrix},$$

where

$$u(y_{2,l_S,i_2}) = \text{NaN}, \quad 1 \leq i_2 \leq q_{l_S},$$

and

$$u^2(y_{2,l_S,i_2}) = \left(\frac{1}{B_{l_S}}\right)^2 \left(\frac{u_r(A_{l_S})A_{l_S}}{100}\right)^2 + \left(\frac{y_{2,l_S,i_2}}{B_{l_S}}\right)^2 \left(\frac{u_r(B_{l_S})B_{l_S}}{100}\right)^2 + 2\left(\frac{y_{2,l_S,i_2}}{(B_{l_S})^2}\right) \left(\frac{\text{cov}_r(A_{l_S}, B_{l_S})A_{l_S}B_{l_S}}{100}\right) + \left(\frac{\delta_{l_S}}{B_{l_S}}\right)^2, \quad q_{l_S} + 1 \leq i_2 \leq m_2.$$