

# Calculation of estimates of subsidiary measurand for examples 1A and 1B

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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 subsidiary measurand in the examples contained in MATLAB scripts `DataFusionSoftware_1A.m` and `DataFusionSoftware_1B.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 subsidiary measurand (and associated uncertainties) are calculated, respectively, by the function `LinearCalibration.m`.

## 2 User input

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

### Sampling and quantization

- $f_2^{(2)}$ , the sampling frequency (in Hz) [cell B10],
- (optional)  $n_B^{(2)}$ , the number of bits for quantization [cell B11],
- (optional)  $s^{(2)}$ , the saturation value [cell B12].

It is assumed that  $f_2^{(2)} < f_1^{(2)}$ , i.e., the sampling frequency for the sensors is less than the sampling frequency used when calculating the ‘true’ subsidiary measurand (section 3).

**Behaviour of sensors**

- $n_S^{(2)}$ , the number of sensors [cell B9 – filled in automatically from information in column C],
- $A_{l_S}^{(2)}$ ,  $u_r(A_{l_S}^{(2)})$ ,  $l_S = 1, \dots, n_S^{(2)}$ , the estimates of the offset parameter and their associated relative standard uncertainties [columns D and F],
- $B_{l_S}^{(2)}$ ,  $u_r(B_{l_S}^{(2)})$ ,  $l_S = 1, \dots, n_S^{(2)}$ , the estimates of the gain parameter and their associated relative standard uncertainties [columns E and G],
- $\text{cov}_r(A_{l_S}^{(2)}, B_{l_S}^{(2)})$ ,  $l_S = 1, \dots, n_S^{(2)}$ , the relative covariances associated with the estimates of the offset and gain parameters [column H],
- $\delta_{l_S}^{(2)}$ ,  $l_S = 1, \dots, n_S^{(2)}$ , the levels of additive noise in the sensor output [column I].

**Behaviour of faulty sensors**

- $n_F^{(2)}$ , the number of faulty sensors [cell L9 – filled in automatically from information in column M],
- $I_{F,l_F}^{(2)}$ ,  $l_F = 1, \dots, n_F^{(2)}$ , the indices of the faulty sensors [column M],
- $C_{l_F}^{(2)}$ ,  $u_r(C_{l_F}^{(2)})$ ,  $l_F = 1, \dots, n_F^{(2)}$ , the estimates of the offset parameter and their associated relative standard uncertainties [columns N and P],
- $D_{l_F}^{(2)}$ ,  $u_r(D_{l_F}^{(2)})$ ,  $l_F = 1, \dots, n_F^{(2)}$ , the estimates of the gain parameter and their associated relative standard uncertainties [columns O and Q],
- $\text{cov}_r(C_{l_F}^{(2)}, D_{l_F}^{(2)})$ ,  $l_F = 1, \dots, n_F^{(2)}$ , the relative covariances associated with the estimates of the offset and gain parameters [column R],
- $T_{F,l_F}^{(2),1}$ ,  $l_F = 1, \dots, n_F^{(2)}$ , the times at which sensors begin behaving faultily [column S],
- $T_{F,l_F}^{(2),2}$ ,  $l_F = 1, \dots, n_F^{(2)}$ , the times at which sensors end behaving faultily [column T].

**Missing data**

- $n_M^{(2)}$ , the number of sensors for which packets of data are missing [cell W9 – filled in automatically from information in column X],
- $n_D^{(2)}$ , the number of data points in each packet [cell W10],
- $I_{M,l_M}^{(2)}$ ,  $l_M = 1, \dots, n_M^{(2)}$ , the indices of the sensors for which packets of data are missing [column X],

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

### 3 Additional input

The following information, previously calculated, is used:

- $t_{1,i_1}^{(2)}$ ,  $i_1 = 1, \dots, m_1^{(2)}$ , the times at which the ‘true’ values of the subsidiary measurand are calculated (corresponding to the sampling frequency  $f_1^{(2)}$ ),
- $y_{1,i_1}^{(2)}$ ,  $i_1 = 1, \dots, m_1^{(2)}$ , the ‘true’ values of the subsidiary measurand, stored in vector  $\mathbf{y}_1^{(2)}$ .

### 4 Sensor response values

The array  $\tilde{\mathbf{V}}_2^{(2)}$  of sensor response values is given by

$$\tilde{\mathbf{V}}_2^{(2)} = \begin{bmatrix} \tilde{\mathbf{v}}_{2,1}^{(2)} & \dots & \tilde{\mathbf{v}}_{2,n_S^{(2)}}^{(2)} \end{bmatrix},$$

where

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

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

1. Evaluate the vector of sensor responses

$$\mathbf{v}_{1,l_S}^{(2)} = A_{l_S}^{(2)*} + B_{l_S}^{(2)*} \mathbf{y}_1^{(2)} + \mathbf{r}_{l_S}^{(2)},$$

corresponding to the times  $t_{1,i_1}^{(2)}$ ,  $i_1 = 1, \dots, m_1^{(2)}$ , where

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

with

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

and

$$\mathbf{r}_{l_S}^{(2)} = \begin{bmatrix} r_{l_S,1}^{(2)} \\ \vdots \\ r_{l_S,m_1^{(2)}}^{(2)} \end{bmatrix},$$

with

$$r_{l_S,i_2}^{(2)} \sim N\left(0, \left(\delta_{l_S}^{(2)}\right)^2\right), \quad i_2 = 1, \dots, m_1^{(2)}.$$

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

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

If no values have been provided for  $n_B^{(2)}$  and  $s^{(2)}$ , then

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

6. For sensors that have missing data, the proportions  $p_{M,l_M}^{(2)}$  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  $\left[T_{M,l_M}^{(2),1}, T_{M,l_M}^{(2),2}\right]$  have their values set to Not-a-Number (NaN),  $l_M = 1, \dots, n_M^{(2)}$ .

## 5 Subsidiary measurand estimates and associated uncertainties

The array  $\mathbf{Y}_2^{(2)}$  of subsidiary measurand estimates is given by

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

where

$$y_{2,l_S,i_2}^{(2)} = \frac{\tilde{v}_{2,l_S,i_2}^{(2)} - A_{l_S}^{(2)}}{B_{l_S}^{(2)}}.$$

The array  $\mathbf{U}_2^{(2)}$  of standard uncertainties associated with the subsidiary measurand estimates is given by

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

where

$$\begin{aligned} u^2\left(y_{2,l_S,i_2}^{(2)}\right) &= \left(\frac{1}{B_{l_S}^{(2)}}\right)^2 \left(\frac{u_r\left(A_{l_S}^{(2)}\right) A_{l_S}^{(2)}}{100}\right)^2 + \left(\frac{y_{2,l_S,i_2}^{(2)}}{B_{l_S}^{(2)}}\right)^2 \left(\frac{u_r\left(B_{l_S}^{(2)}\right) B_{l_S}^{(2)}}{100}\right)^2 + \\ &2 \left(\frac{y_{2,l_S,i_2}^{(2)}}{\left(B_{l_S}^{(2)}\right)^2}\right) \left(\frac{\text{cov}_r\left(A_{l_S}^{(2)}, B_{l_S}^{(2)}\right) A_{l_S}^{(2)} B_{l_S}^{(2)}}{100}\right) + \left(\frac{\delta_{l_S}^{(2)}}{B_{l_S}^{(2)}}\right)^2. \end{aligned}$$