



Tutorial 8

Analysis of Error Factors for Measurement Data and Inverse Techniques Application to Temperature Measurements and Heat Flux Estimations

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INSA de Lyon - CNRS - UCBL

14th June 2011





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- 2 Method 1 : Determination of an Estimator Uncertainty
- 3 Method 2 : The Monte Carlo Method
- 4 Application 0 : Presentation of Study Case
- 5 Application 1 : Temperature Estimation Uncertainties
- 6 Application 2 : Inverse Technique Uncertainties



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- 4 Application 0 : Presentation of Study Case
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- 6 Application 2 : Inverse Technique Uncertainties



**Centre de
Thermique de
Lyon**

M0 : Foreword

Bibliography

Definitions

Scope

M1 : Estimator
Uncertainty

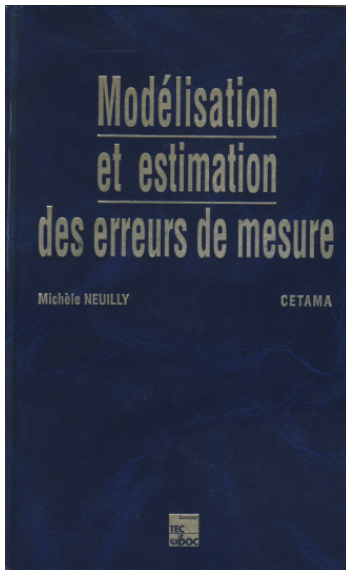
M2 : Monte Carlo
Method

A0 : Presentation
Study Case

A1 : Temperature
Uncertainties

A2 : IT
Uncertainties

Bibliography





Bibliography

M0 : Foreword

Bibliography

Definitions

Scope

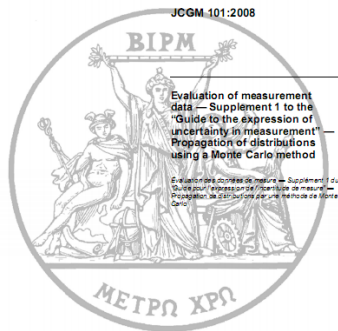
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A2 : IT
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First edition 2008

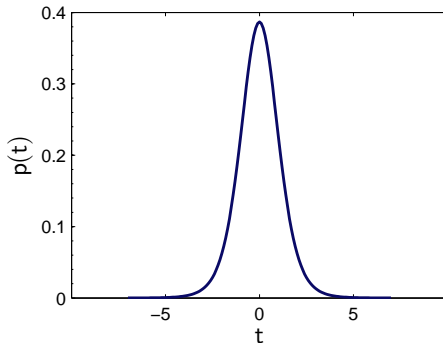
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Definitions : Statistics

t-Distribution Random Variable

$$\Delta(E, s^2, \nu, N)$$



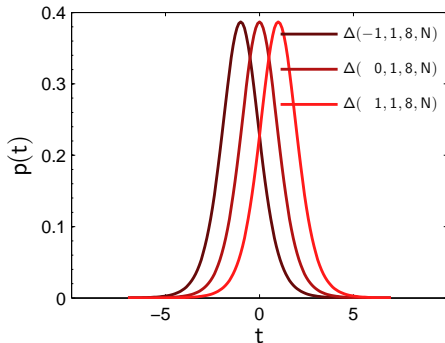


Definitions : Statistics

t-Distribution Random Variable

$$\Delta(E, s^2, \nu, N)$$

Mathematical Expectation

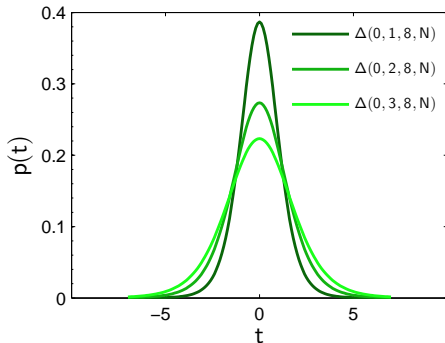




Definitions : Statistics

t-Distribution Random Variable

$$\Delta(E, s^2, \nu, N) \quad \text{Variance}$$



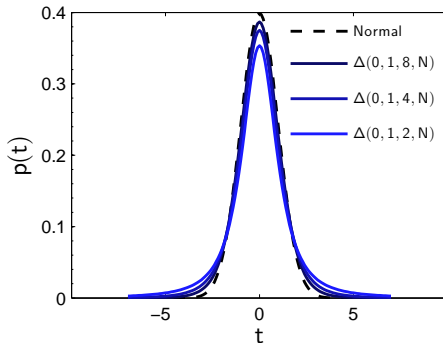


Definitions : Statistics

t-Distribution Random Variable

$$\Delta(E, s^2, \nu, N)$$

Degree of Freedom





Definitions : Statistics

t-Distribution Random Variable

$$\Delta(E, s^2, \nu, N)$$

Number of associated elements



Definitions : Statistics

t-Distribution Random Variable

$$\Delta(E, s^2, \nu, N)$$

Interval of confidence :

$$\Delta^{95\%} = E \pm t_{\nu}^{0.975} \sqrt{s^2}$$



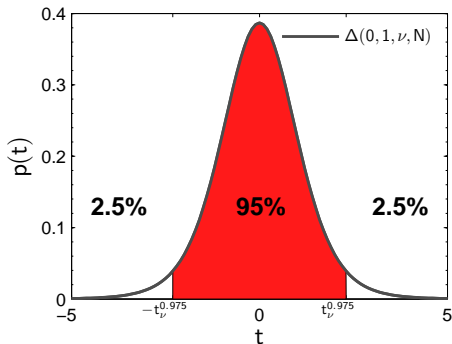
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Definitions

Estimator of Y_{true}

Measurement Device

(Thermocouple, Balance...)

Data Treatment

(Averaging)



Definitions

Estimator of Y_{true}

Measurement Device

(Thermocouple, Balance...)

Raw Data

Y_m^p

Data Treatment

(Averaging)

Estimation

$$\bar{Y}_m = \frac{\sum_{p=1}^{N_m} Y_m^p}{N_m}$$



Definitions

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(Thermocouple, Balance...)

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$$Y_m^p$$

Data Treatment

(Averaging)

Estimation

$$\bar{Y}_m = \frac{\sum_{p=1}^{N_m} Y_m^p}{N_m}$$

Raw Data Total Error

$$Y_m^p = Y_{\text{true}} + \varepsilon_{Y_m}^p$$

Estimation Total Error

$$\bar{Y}_m = Y_{\text{true}} + \varepsilon_{\bar{Y}_m}$$



Definitions

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Measurement Device

(Thermocouple, Balance...)

Raw Data

$$Y_m^p$$

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$$Y_m^p = Y_{\text{true}} + \varepsilon_{Y_m}^p$$

$$\varepsilon_{Y_m}^p \in \mathcal{N}_Y(\mathbf{0}, \sigma_Y)$$

Data Treatment

(Averaging)

Estimation

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Estimation Total Error

$$\bar{Y}_m = Y_{\text{true}} + \varepsilon_{\bar{Y}_m}$$

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Definitions

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(Thermocouple, Balance...)

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$$Y_m^p$$

Raw Data Total Error

$$Y_m^p = Y_{\text{true}} + \varepsilon_{Y_m}^p$$

$$\varepsilon_{Y_m}^p \in \mathcal{N}_Y(0, \sigma_Y)$$

Raw Data Total Error
Random Variable

$$\Delta_Y(0, s_Y^2, \nu_Y, N_Y)$$

Data Treatment

(Averaging)

Estimation

$$\bar{Y}_m = \frac{\sum_{p=1}^{N_m} Y_m^p}{N_m}$$

Estimation Total Error

$$\bar{Y}_m = Y_{\text{true}} + \varepsilon_{\bar{Y}_m}$$

$$\varepsilon_{\bar{Y}_m} \in \mathcal{N}_{\bar{Y}_m}(0, \sigma_{\bar{Y}_m})$$

Estimation Total Error
Random Variable

$$\Delta_{\bar{Y}_m}(0, s_{\bar{Y}_m}^2, \nu_{\bar{Y}_m}, N_{\bar{Y}_m})$$



Definitions

Estimator Uncertainty

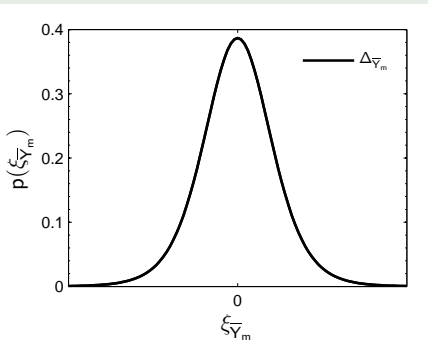
$Y_{\text{true}} = \overline{Y}_m \pm \Delta_{\overline{Y}_m}^{95\%}$ is true with a level of confidence 95%



Definitions

Estimator Uncertainty

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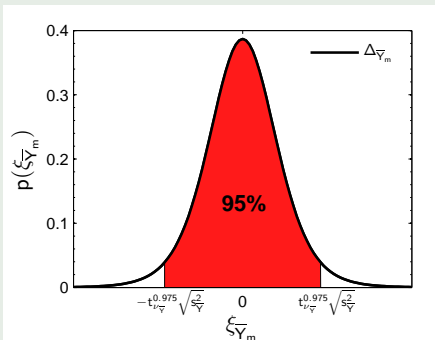




Definitions

Estimator Uncertainty

$Y_{\text{true}} = \bar{Y}_m \pm \Delta_{\bar{Y}_m}^{95\%}$ is true with a level of confidence 95%



$$\Delta_{\bar{Y}_m}^{95\%} = t_{\nu_{\bar{Y}_m}}^{0.975} \sqrt{s_{\bar{Y}_m}^2}$$



Definitions

Error Factors f_q

Physical Phenomena which deteriorate the quality of the measurement. They must be independent !

Fixed Error Factors

Temperature, Pollutant Concentration, Air speed, measurement tool...

$$f_1 \dots f_{M-1}$$

Fluctuating Error Factor

Electromagnetic noise...

$$f_M$$



Definitions

Error Factors f_q

Physical Phenomena which deteriorate the quality of the measurement. They must be independent !

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$$f_1 \dots f_{M-1}$$

Fluctuating Error Factor

Electromagnetic noise...

$$f_M$$

Raw Data Total Error

$$Y_m^p = Y_{\text{true}} + \varepsilon_{Y_m}^p$$

Estimation Total Error

$$\bar{Y}_m = Y_{\text{true}} + \varepsilon_{\bar{Y}_m}$$

Raw Data Associated Errors

$$Y_m^p = Y_{\text{true}} + \varepsilon_{Y_m, f_1}^p + \dots + \varepsilon_{Y_m, f_M}^p$$

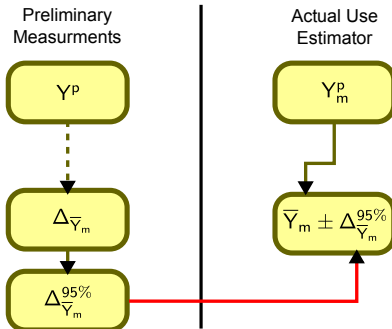
Estimation Associated Errors

$$\bar{Y}_m = Y_{\text{true}} + \varepsilon_{\bar{Y}_m, f_1}^p + \dots + \varepsilon_{\bar{Y}_m, f_M}^p$$



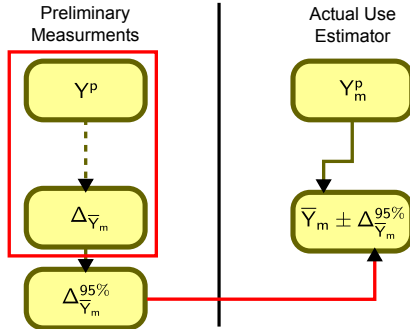
Scope

- Conservation of $\Delta_{\bar{Y}_m}$



Scope

- Conservation of $\Delta_{\bar{Y}_m}$



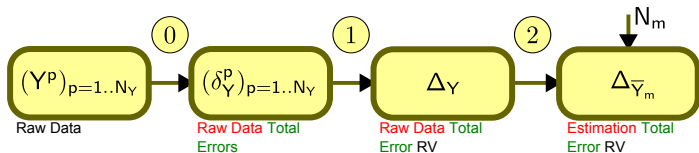


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 - Error Factor Analysis
 - Extra Calculation
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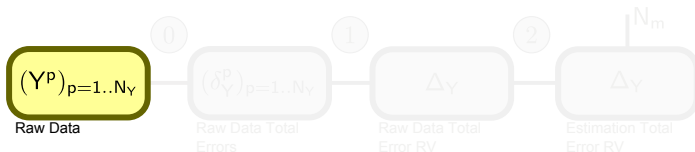


No Error Factor Analysis





No Error Factor Analysis

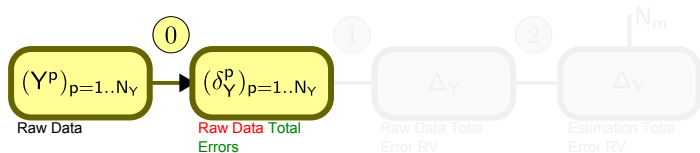


Population of raw data

- Measurement of the **same true value** Y_{true}
- Y_{true} not necessarily known.

$$Y^p = Y_{\text{true}} + \varepsilon_Y^p$$

No Error Factor Analysis



Step 0 : Approximation of the raw data errors

- Approximation of ε_Y^p

$$\delta_Y^p = Y^p - \bar{Y} = \varepsilon_Y^p - \bar{\varepsilon}_Y^p$$

No Error Factor Analysis

M0 : Foreword

M1 : Estimator
Uncertainty

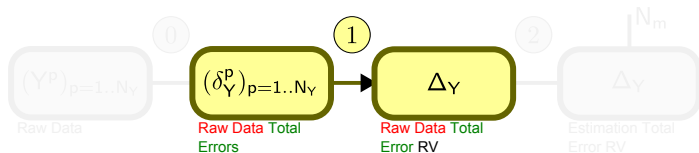
No Error Factor
Analysis
Error Factor Analysis
Extra Calc

M2 : Monte Carlo
Method

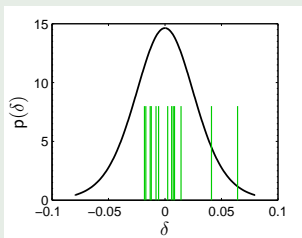
A0 : Presentation
Study Case

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A2 : IT
Uncertainties



Step 1 : Determination of Δ_Y



$$N_Y = N_Y$$

$$\nu_Y = N_Y - 1$$

$$s_Y^2 = \frac{\sum_{p=1}^{N_Y} (\delta_Y^p)^2}{\nu_Y}$$

No Error Factor Analysis

M0 : Foreword

M1 : Estimator
Uncertainty

No Error Factor
Analysis

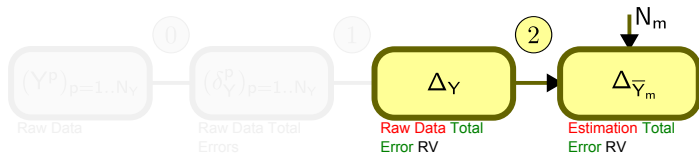
Error Factor Analysis
Extra Calc

M2 : Monte Carlo
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Step 3 : Determination of $\Delta_{\bar{Y}_m}$

$$\frac{\varepsilon^1 + .. + \varepsilon^{N_m}}{N_m}$$

$$N_{\bar{Y}_m} = N_Y$$

$$\nu_{\bar{Y}_m} = \nu_Y$$

$$s_{\bar{Y}_m}^2 = \frac{s_Y^2}{N_m} \quad \text{Decreasing}$$



No Error Factor Analysis

Advantages

- Quick Method
- Fluctuation of the Estimation Value

Drawbacks

- No Bias
- No Interpretation of the Error Structure

Error Factor Analysis

M0 : Foreword

M1 : Estimator
Uncertainty

No Error Factor
Analysis

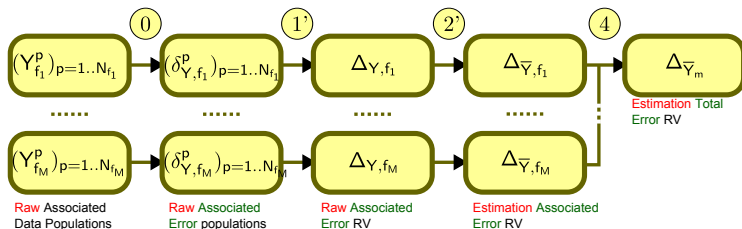
Error Factor Analysis
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Error Factor Analysis



Population of raw data

- f_k are **blocked** by controlling environment

$$Y_{f_M}^1 = Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \varepsilon_{Y,f_M}^1$$

$$Y_{f_M}^2 = Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \varepsilon_{Y,f_M}^2$$

.....

$$Y_{f_q}^{N_{f_M}} = Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \varepsilon_{Y,f_M}^{N_{f_M}}$$



Error Factor Analysis



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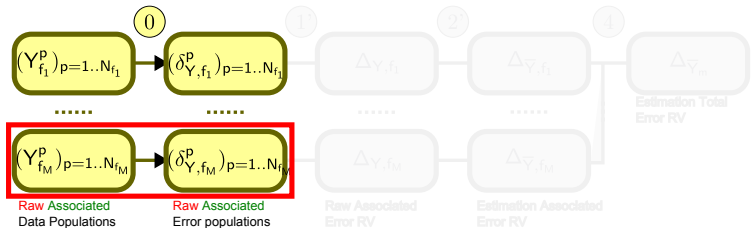
$$Y_{f_M}^2 = Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \varepsilon_{Y,f_M}^2$$

.....

$$Y_{f_q}^{N_{f_M}} = Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \varepsilon_{Y,f_M}^{N_{f_M}}$$

$$Y_{f_M} = Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}$$

Error Factor Analysis



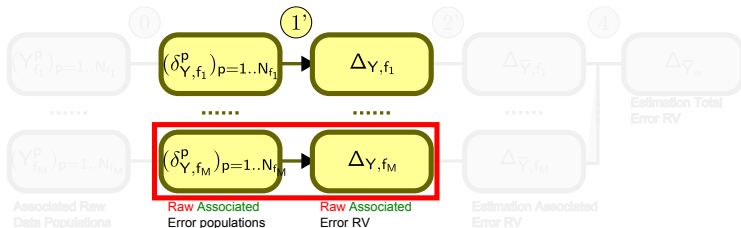
Step 0 : Approximation of the raw associated errors

- Approximation of ε_{Y,f_M}^p

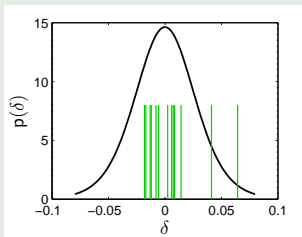
$$\delta_{Y,f_M}^p = Y_{f_M}^p - \bar{Y}_{f_M} = \varepsilon_{Y,f_M}^p - \bar{\varepsilon}_{Y,f_M}$$



Error Factor Analysis



Step 1 : Determination of Δ_{Y, f_M}



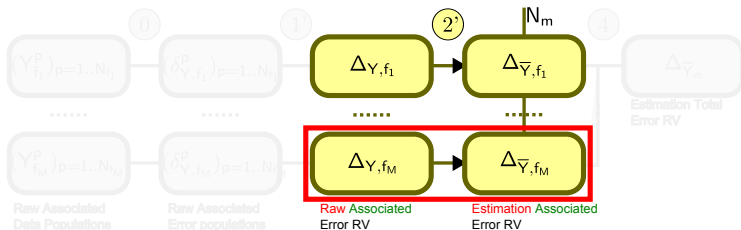
$$N_{Y, f_M} = N_{f_M}$$

$$\nu_{Y, f_M} = N_{f_M} - 1$$

$$S^2_{Y, f_M} = \frac{\sum_{p=1}^{N_{f_M}} (\delta^p_{Y, f_M})^2}{\nu_{Y, f_M}}$$



Error Factor Analysis



Step 2 : Determination of $\Delta_{\bar{Y},f_M}$

$$\frac{\varepsilon^1 + .. + \varepsilon^{N_m}}{N_m}$$

$$N_{\bar{Y},f_M} = N_{Y,f_M}$$

$$\nu_{\bar{Y},f_M} = \nu_{Y,f_M}$$

$$s_{\bar{Y},f_M}^2 = \frac{s_{Y,f_M}^2}{N_m} \text{ Decreasing}$$



Error Factor Analysis



Population of raw data

- f_M blocked by averaging

$$\begin{aligned}
 Y_{f_q}^1 &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q}^1 + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}^1 \\
 Y_{f_q}^2 &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q}^2 + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}^2 \\
 &\dots\dots\dots \\
 Y_{f_q}^{N_{f_q}} &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q}^{N_{f_q}} + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}^{N_{f_q}}
 \end{aligned}$$



Error Factor Analysis

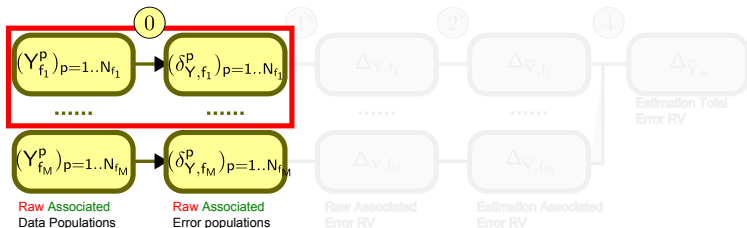


Population of raw data

- f_M blocked by averaging

$$\begin{aligned}
 Y_{f_q}^1 &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q}^1 + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}^1 \\
 Y_{f_q}^2 &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q}^2 + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}^2 \\
 &\dots\dots\dots \\
 Y_{f_q}^{N_{f_q}} &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \varepsilon_{Y,f_q}^{N_{f_q}} + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}^{N_{f_q}} \\
 \hline
 Y_{f_q} &= Y_{\text{true}} + \varepsilon_{Y,f_1} + \dots + \varepsilon_{Y,f_{q-1}} + \bar{\varepsilon}_{Y,f_q} + \varepsilon_{Y,f_{q+1}} + \dots + \bar{\varepsilon}_{Y,f_M}
 \end{aligned}$$

Error Factor Analysis



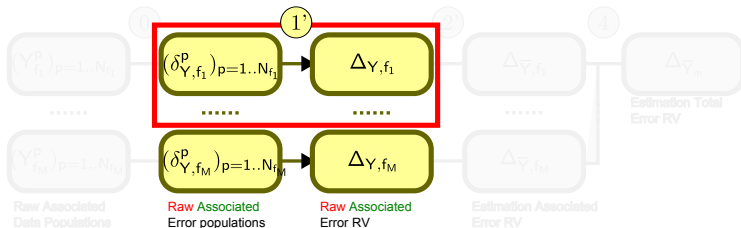
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- Approximation of ε_{Y,f_q}^p

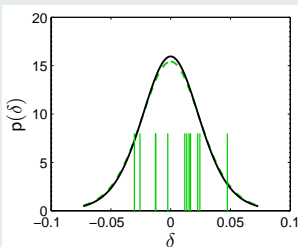
$$\delta_{Y,f_q}^p = Y_{f_q}^p - \bar{Y}_{f_q} = (\varepsilon_{Y,f_q}^p - \bar{\varepsilon}_{Y,f_q}) + (\bar{\varepsilon}_{Y,f_M} - \bar{\bar{\varepsilon}}_{Y,f_M})$$



Error Factor Analysis



Step 1 : Determination of Δ_{Y,f_q}



$$N_{Y,f_q} = N_{f_q}$$

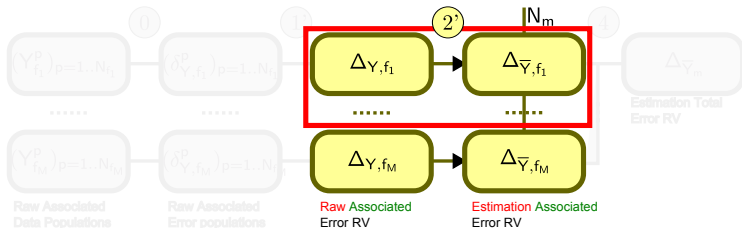
$$\nu_{Y,f_q} = N_{f_q} - 1$$

$$s_{Y,f_q}^2 = \frac{\sum_{p=1}^{N_{f_q}} (\delta_{Y,f_q}^p)^2}{\nu_{Y,f_q}} - \frac{s_{Y,f_M}^2}{N_{f_q}}$$

Compensation



Error Factor Analysis



Step 2 : Determination of $\Delta \bar{Y}_{f_q}$

$$\frac{\varepsilon^1 + .. + \varepsilon^{N_m}}{N_m}$$

$$N_{\bar{Y},f_q} = N_{Y,f_q}$$

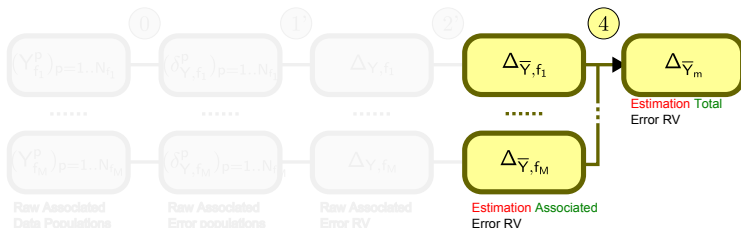
$$\nu_{\bar{Y},f_q} = \nu_{Y,f_q}$$

$$s_{\bar{Y},f_q}^2 = \frac{s_{Y,f_q}^2}{N_m} \quad \text{Fluctuating Error}$$

$$s_{\bar{Y},f_q}^2 = s_{Y,f_q}^2 \quad \text{Bias}$$



Error Factor Analysis



Step 4 : Determination of $\Delta \bar{Y}_m$

$$\varepsilon_{f_1} + \dots + \varepsilon_{f_M}$$

$$N_{\bar{Y}_m} = \sum_q N_{\bar{Y}, f_q}$$

$$\nu_{\bar{Y}_m} = \left| \frac{\left(\sum_{q=1}^{N_f} \left[\frac{s_{\bar{Y}, f_q}^2}{N_{\bar{Y}, f_q}} \right] \right)^2}{\sum_{q=1}^{N_f} \left[\left(\frac{s_{\bar{Y}, f_q}^2}{N_{\bar{Y}, f_q}} \right)^2 \frac{1}{\nu_{\bar{Y}, f_q}} \right]} \right|$$

$$s_{\bar{Y}_m}^2 = \sum_q s_{\bar{Y}, f_q}^2 \quad \text{Sum Variances}$$



Error Factor Analysis

Advantages

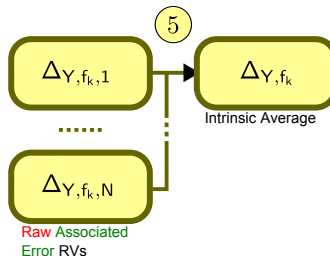
- Description of the Error Structure
- Bias Taken Into Account

Drawbacks

- Choice of the Predominant Error Factors



Extra Calculation



Step 5 : Intrinsic average



- Same Error Factor
- Different Y_{true}

$$N_{Y,f_k} = \sum_i N_{Y,f_k,i}$$

$$\nu_{Y,f_k} = \sum_i \nu_{Y,f_k,i}$$

$$s_{Y,f_k}^2 = \frac{\sum_i \nu_{Y,f_k,i} \times s_{Y,f_k,i}^2}{\sum \nu_{Y,f_k,i}}$$

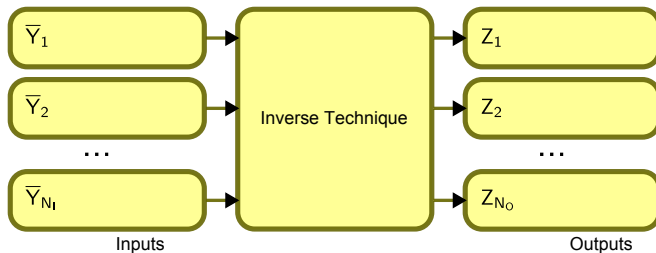


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 - Generation of Noised Input Populations
 - Error Factor Analysis
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The Monte Carlo Method

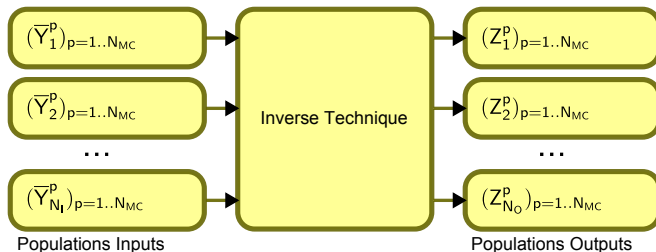


Inverse Technique

- N_I Estimation Inputs
- N_O Outputs



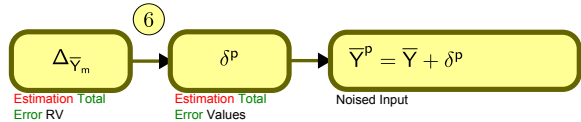
The Monte Carlo Method



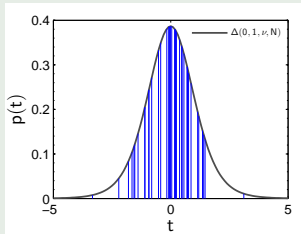
The Monte Carlo Method

- Noised population for the inputs
- Noised outputs

Generation of Noised Input Populations



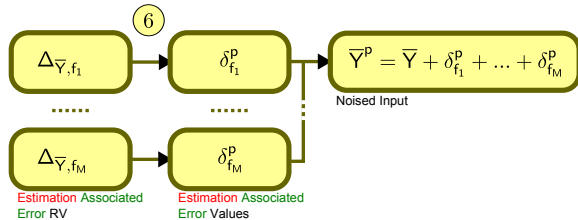
Generation from $\Delta \bar{Y}_m$



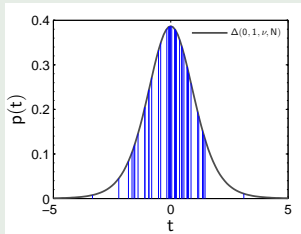
$$\delta^P = t_{\text{rand}}^{\nu_{\bar{Y}_m}} \sqrt{s_{\bar{Y}_m}^2}$$

Command `trand` on
MATLAB™

Generation of Noised Input Populations



Generation from $\Delta\bar{Y}_m$

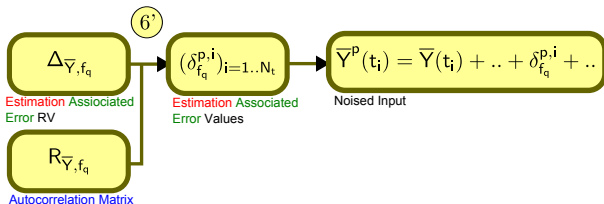


$$\delta_{f_q}^p = t_{\text{rand}}^{\nu_{\bar{Y}, f_q}} \sqrt{s_{\bar{Y}, f_q}^2}$$

Command `trand` on
MATLAB™



Generation of Noised Input Populations



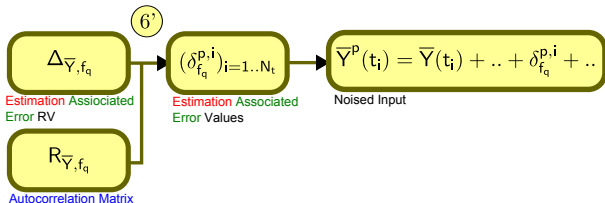
Same estimator \bar{Y} , different times t_i

Dependency between :

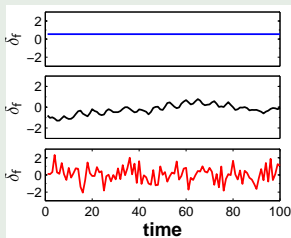
- $\delta(t_i)$
- $\delta(t_j)$ with $j < i$



Generation of Noised Input Populations



Same estimator \bar{Y} , different times t_i

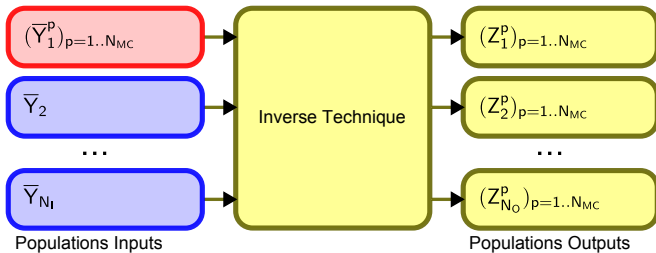


$$(\delta^i) = \text{mvtrnd}(R, \nu) \sqrt{s^2}$$

- Fully dependent : $R_{i,j} = 1$
- Autocorrelated : $R_{i,j} \neq 0$
- Independent : $R_{i,j} = 0$

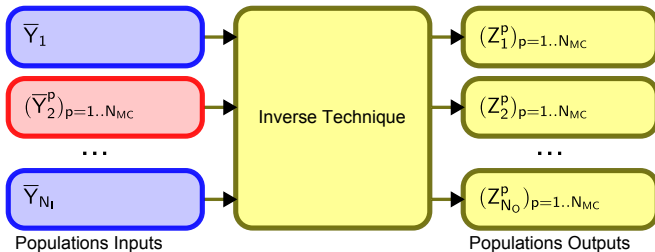


Error Factor Analysis



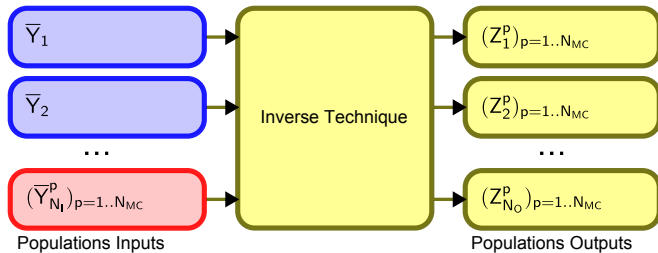


Error Factor Analysis





Error Factor Analysis



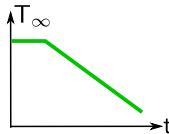
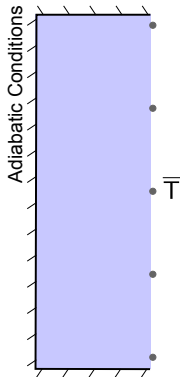


Content of the section

- 1 Method 0 : Foreword, Definitions
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- 3 Method 2 : The Monte Carlo Method
- 4 Application 0 : Presentation of Study Case**
 - The Inverse Technique
- 5 Application 1 : Temperature Estimation Uncertainties
- 6 Application 2 : Inverse Technique Uncertainties

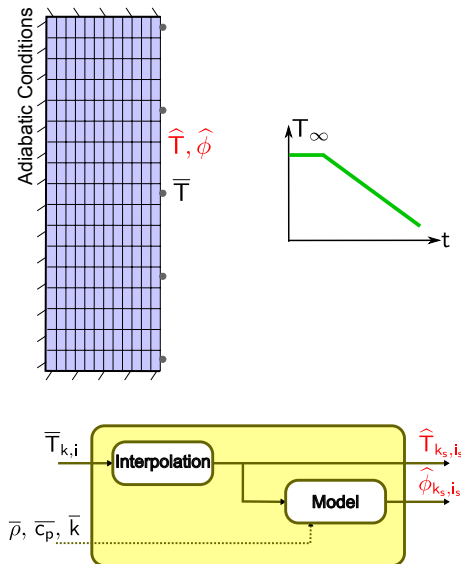


The Inverse Technique





The Inverse Technique





The Inverse Technique

M0 : Foreword

M1 : Estimator
Uncertainty

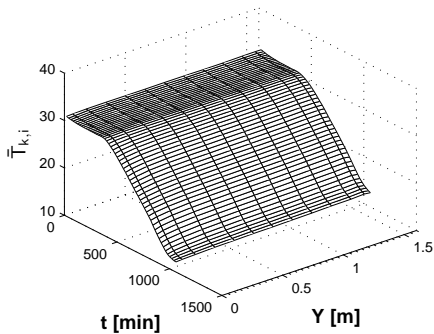
M2 : Monte Carlo
Method

A0 : Presentation
Study Case

The Inverse Technique

A1 : Temperature
Uncertainties

A2 : IT
Uncertainties





Centre de
Thermique de
Lyon

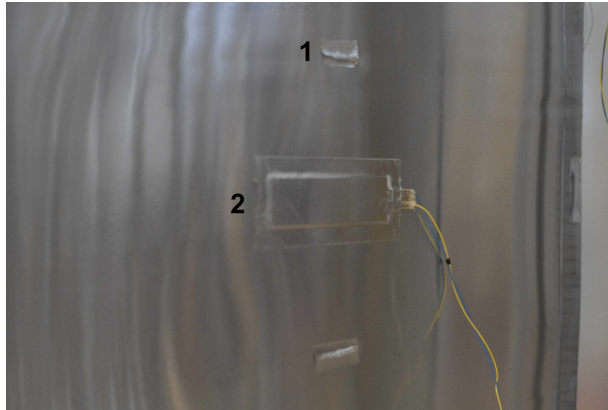
The Inverse Technique

M0 : Foreword
M1 : Estimator
Uncertainty
M2 : Monte Carlo
Method
A0 : Presentation
Study Case
The Inverse Technique
A1 : Temperature
Uncertainties
A2 : IT
Uncertainties



The Inverse Technique

- M0 : Foreword
- M1 : Estimator
Uncertainty
- M2 : Monte Carlo
Method
- A0 : Presentation
Study Case
 - The Inverse Technique**
- A1 : Temperature
Uncertainties
- A2 : IT
Uncertainties



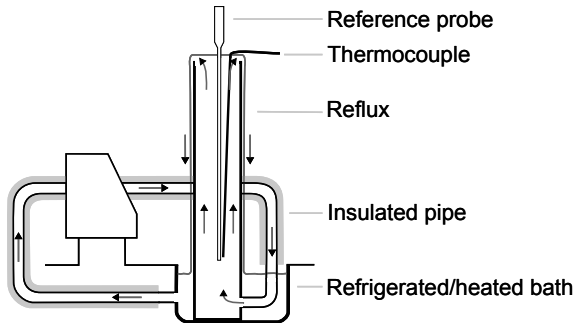


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 - Calibration of the Thermocouples
 - The error factors
 - Results
- 6 Application 2 : Inverse Technique Uncertainties



The Calibration of the Thermocouples



Experimental Setup



The Calibration of the Thermocouples

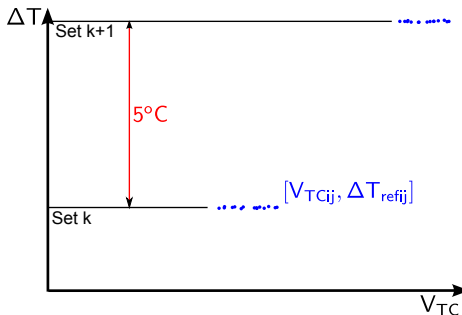


Data Acquisition

- ΔT : Temperature difference between the two junctions
- V_{TC} : Voltage between the two junctions



The Calibration of the Thermocouples

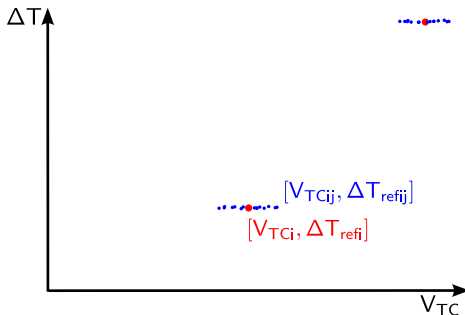


Raw data

- $N_T = 13$ Sets of acquisition
- $N_A = 15$ Acquisitions per set



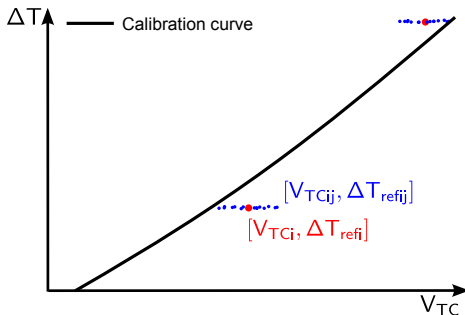
The Calibration of the Thermocouples



Averaging the data



The Calibration of the Thermocouples



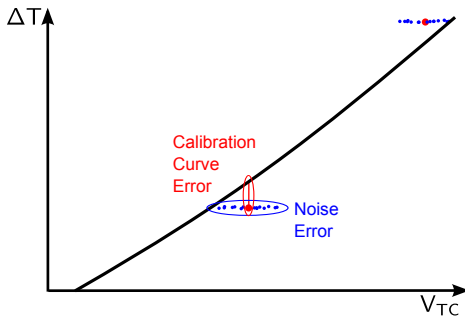
Calibration Curve

- obtained with $[V_{TC,i}, \Delta T_{ref,i}]$

$$\Delta T_c = c_2 V_{TC}^2 + c_1 V_{TC} + c_0$$



The Calibration of the Thermocouples

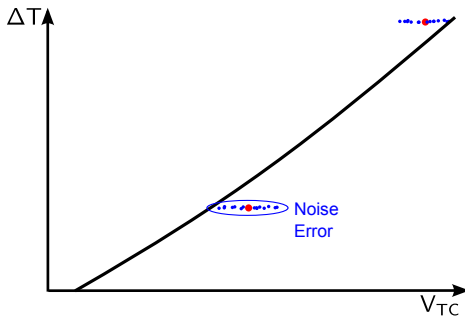


The error factors

- Noise Error
- Calibration Curve Error



The Noise Error

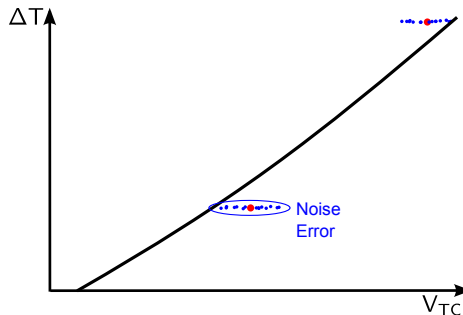


Description

- Due to the electromagnetic noise
- Fast fluctuating error



The Noise Error



Approximation of the noise error

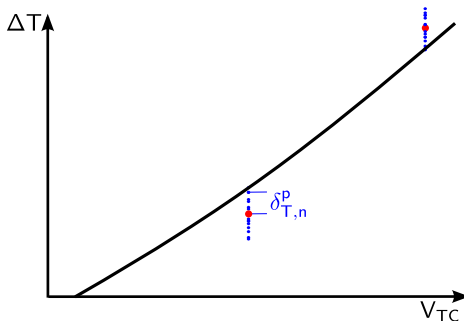
$$V_{TC,i,1} = V_{\text{true}}^i + \epsilon_{V,c}^i + \epsilon_{V,n}^{i,1}$$

$$V_{TC,i,2} = V_{\text{true}}^i + \epsilon_{V,c}^i + \epsilon_{V,n}^{i,2}$$

.....

$$\frac{V_{TC,i,N_A} = V_{\text{true}}^i + \epsilon_{V,c}^i + \epsilon_{V,n}^{i,N_A}}{V_{TC,i} = V_{\text{true}}^i + \epsilon_{V,c}^i + \bar{\epsilon}_{V,n}^i}$$

The Noise Error



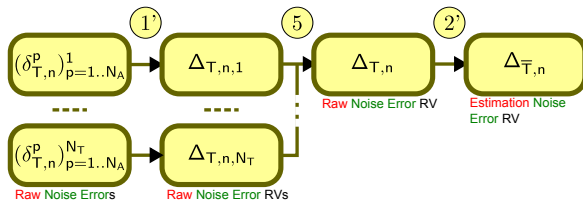
Approximation of the noise error

- Turning a voltage error into a temperature error

$$\delta_{T,n}^p = \Delta_c(V_{TC,i,1} - \bar{V}_{TC,i,j}) = \varepsilon_{T,n}^{i,p} - \bar{\varepsilon}_{T,n}^i$$



The Noise Error

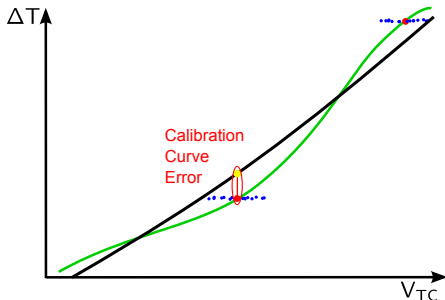


Calculation steps

- One noise error random variable per set of acquisition
- Step 5 : Intrinsic average
- Step 2' : **Fluctuating Error** : the Averaging decreases the Variance



The Calibration Curve Error

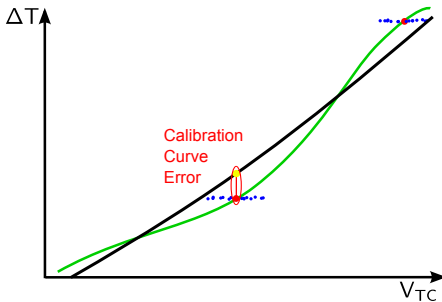


Description

- Discrepancy between the curve and the real behavior of the Thermocouple
- Bias during a set of measurement



The Calibration Curve Error



Approximation of the calibration curve error

$$\Delta_c(\bar{V}_{TC,1}) = T_{\text{true}}^1 + \varepsilon_{T,c}^1 + \bar{\varepsilon}_{T,n}^1$$

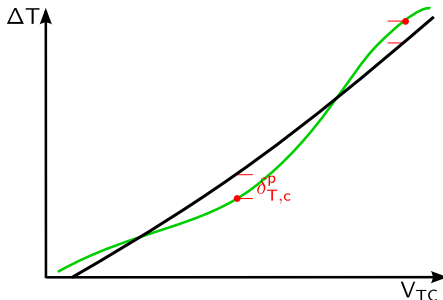
$$\Delta_c(\bar{V}_{TC,2}) = T_{\text{true}}^2 + \varepsilon_{T,c}^2 + \bar{\varepsilon}_{T,n}^2$$

.....

$$\Delta_c(\bar{V}_{TC,N_T}) = T_{\text{true}}^{N_T} + \varepsilon_{T,c}^{N_T} + \bar{\varepsilon}_{T,n}^{N_T}$$



The Calibration Curve Error



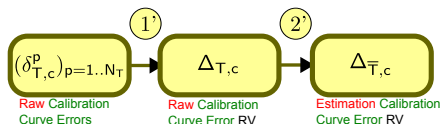
Approximation of the errors

- We suppose $T_{\text{true}}^i \approx \Delta T_{\text{ref},i}$
- 3 Parameters c_1, c_2, c_3 to estimate the errors :

$$\delta_{T,c}^p = \Delta T_c(\bar{V}_{TC,i}) - \Delta T_{\text{ref},i}$$



The Calibration Curve Error

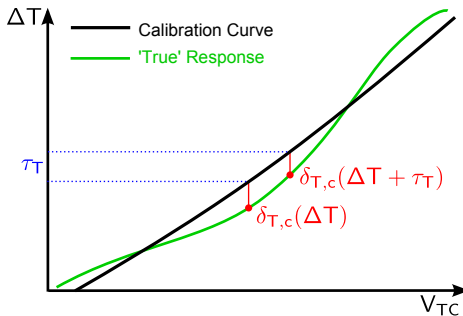


Calculation steps

- Step 1' : $\nu_{T,c} = N_T - 3$, because of 3 parameters
- Step 1' : **Variance Correction**
- Step 2' : **Bias** : the Averaging does not modify the Variance



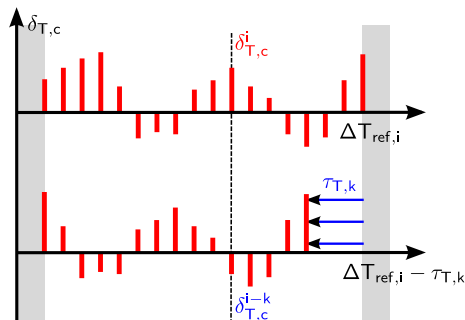
The Calibration Curve Error



Autocorrelation



The Calibration Curve Error



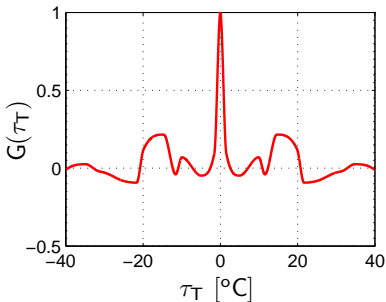
Autocorrelation

- $\tau_{T,k} = k \times 5^\circ\text{C}$ Temperature delay
- Autocorrelation :

$$G_{T,c}(\tau_{T,k}) = \frac{\sum_{i=1}^{N_A} \delta_{T,c}^i \delta_{T,c}^{i-k}}{\sum_{i=1}^{N_A} (\delta_{T,c}^i)^2}$$



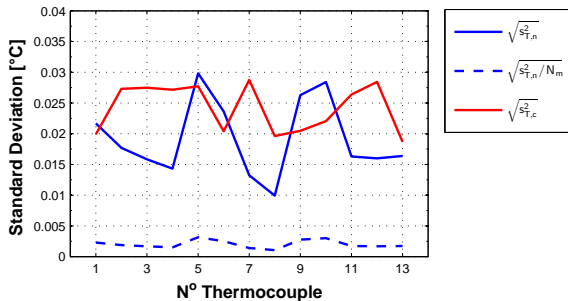
The Calibration Curve Error



Autocorrelation



Temperature Estimation Uncertainties

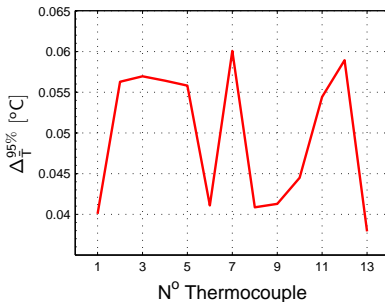


Standard Deviations

- $N_m = 90$



Temperature Estimation Uncertainties



Uncertainties

$$T = T_{SF} + \Delta T_c(\overline{V_{TC}})$$

- Error on cold junction neglected
- Average : $\Delta \bar{T}_m^{95\%} = 0.05^{\circ}\text{C}$

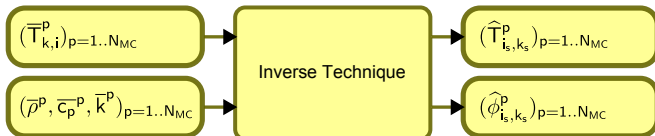


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 - Generation of the Noised Input Populations
 - Temperature Uncertainty
 - Heat Flux Uncertainty



Generation of the Noised Input Populations



M0 : Foreword

M1 : Estimator
Uncertainty

M2 : Monte Carlo
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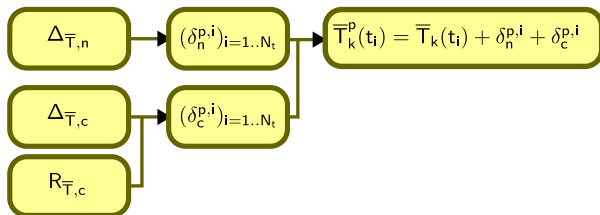
NIP Generation

Temperature
Uncertainty

Heat Flux Uncertainty



Generation of the Noised Input Populations

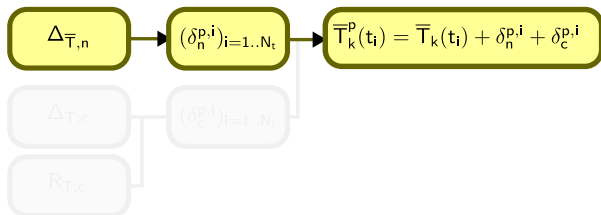


Noised Temperature

$$\bar{T}_{k,i} = \bar{T}_k(t_i)$$



Generation of the Noised Input Populations

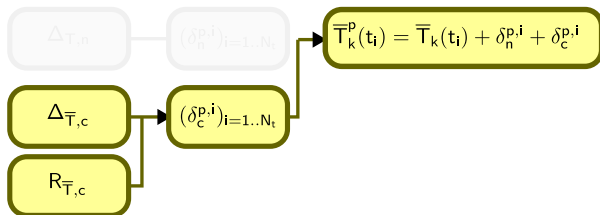


Noise Errors

- Independent Errors



Generation of the Noised Input Populations



Calibration Curve Errors

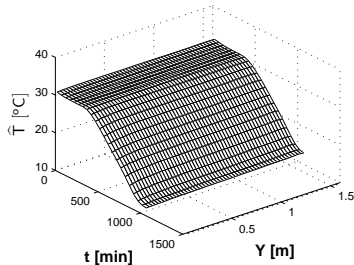
- Autocorrelated Errors

$$R_{i,j} = G_T(\bar{T}(t_i) - \bar{T}(t_j))$$

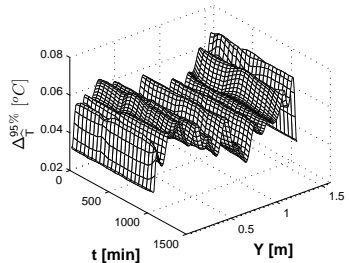


Temperature Uncertainty

IM Mean Temperature



IM Temperature Uncertainty

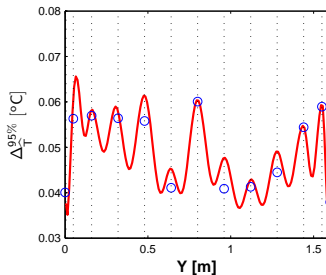


Temperature Uncertainty

- $\Delta_{\hat{T}}^{95\%}$ constant with time



Temperature Uncertainty



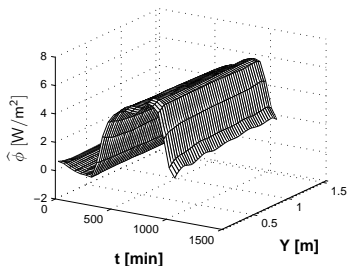
Time-Averaged Temperature Uncertainty

- **Local Maximums** at Thermocouple locations

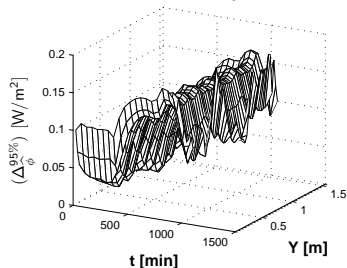


Heat Flux Uncertainty

IM Mean Heat Flux



Uncertainty

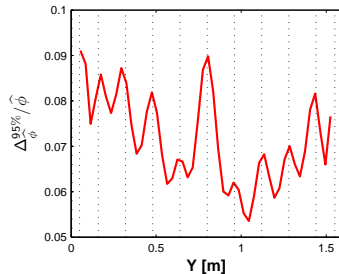
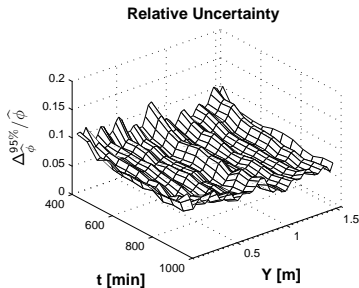


Heat Flux Uncertainty

- $\Delta_{\hat{\phi}}^{95\%}$ proportional to $\hat{\phi}$



Heat Flux Uncertainty



Relative Heat Flux Uncertainty

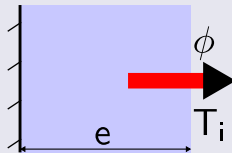
- $\Delta_{\hat{\phi}}^{95\%}/\hat{\phi}$ almost **constant with time**



Heat Flux Uncertainty : Error Factors

Interpretation

$$\phi \approx e \rho c_p \frac{T_i - T_{i-1}}{\Delta t}$$



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Temperature
Uncertainty

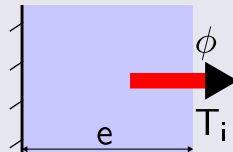
Heat Flux Uncertainty



Heat Flux Uncertainty : Error Factors

Interpretation

$$\phi \approx e \rho c_p \frac{T_i - T_{i-1}}{\Delta t}$$



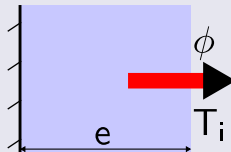
$$\phi + \varepsilon_\phi \approx e(\rho + \varepsilon_\rho)(c_p + \varepsilon_{c_p}) \frac{(T_i + \varepsilon_T^i) - (T_{i-1} + \varepsilon_T^{i-1})}{\Delta t}$$



Heat Flux Uncertainty : Error Factors

Interpretation

$$\phi \approx e \rho c_p \frac{T_i - T_{i-1}}{\Delta t}$$



$$\phi + \varepsilon_\phi \approx e(\rho + \varepsilon_\rho)(c_p + \varepsilon_{c_p}) \frac{(T_i + \varepsilon_T^i) - (T_{i-1} + \varepsilon_T^{i-1})}{\Delta t}$$

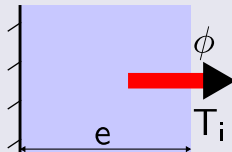
$$\phi + \varepsilon_\phi \approx \phi + \frac{\varepsilon_\rho}{\rho} \phi + \frac{\varepsilon_{c_p}}{c_p} \phi + (\varepsilon_T^i - \varepsilon_T^{i-1}) \frac{e \rho c_p}{\Delta t}$$



Heat Flux Uncertainty : Error Factors

Interpretation

$$\phi \approx e \rho c_p \frac{T_i - T_{i-1}}{\Delta t}$$



$$\phi + \varepsilon_\phi \approx e(\rho + \varepsilon_\rho)(c_p + \varepsilon_{c_p}) \frac{(T_i + \varepsilon_T^i) - (T_{i-1} + \varepsilon_T^{i-1})}{\Delta t}$$

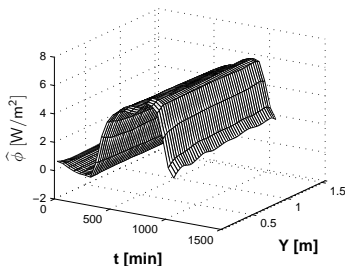
$$\phi + \varepsilon_\phi \approx \phi + \frac{\varepsilon_\rho}{\rho} \phi + \frac{\varepsilon_{c_p}}{c_p} \phi + (\varepsilon_T^i - \varepsilon_T^{i-1}) \frac{e \rho c_p}{\Delta t}$$

$$\varepsilon_\phi \approx \left[\frac{\varepsilon_\rho}{\rho} + \frac{\varepsilon_{c_p}}{c_p} \right] \phi + (\varepsilon_T^i - \varepsilon_T^{i-1}) \frac{e \rho c_p}{\Delta t}$$

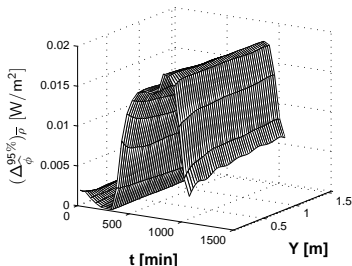


Heat Flux Uncertainty : Error Factors

IM Mean Heat Flux



Uncertainty associated to ρ



Heat Flux Uncertainty associated to ρ

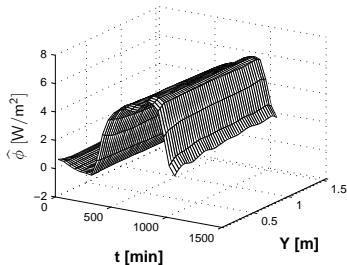
- proportional to $\hat{\phi}$

$$\varepsilon_{\phi} \approx \frac{\varepsilon_{\rho}}{\rho} \phi \propto \phi$$

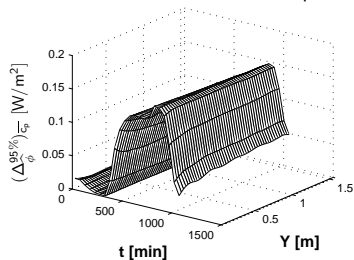


Heat Flux Uncertainty : Error Factors

IM Mean Heat Flux



Uncertainty associated to c_p



Heat Flux Uncertainty associated to c_p

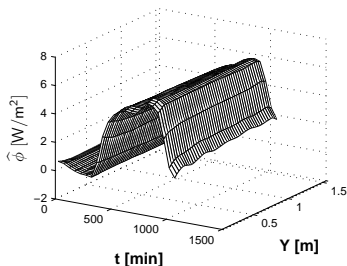
- proportional to $\hat{\phi}$

$$\epsilon_{\phi} \approx \frac{\epsilon_{c_p}}{c_p} \phi \propto \phi$$

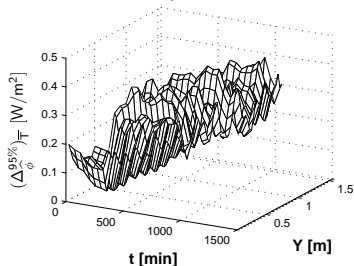


Heat Flux Uncertainty : Error Factors

IM Mean Heat Flux



Uncertainty associated to T



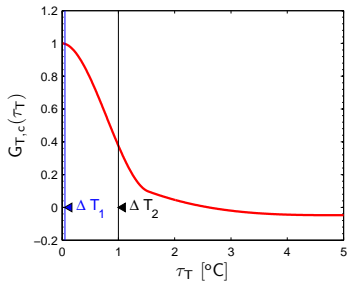
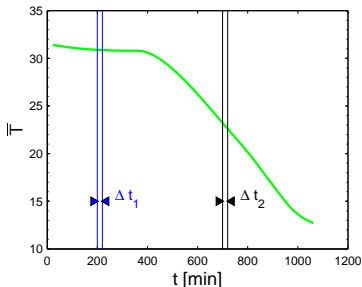
Heat Flux Uncertainty associated to T

- proportional to $\hat{\phi}$

$$\varepsilon_{\phi} \approx (\varepsilon_T^i - \varepsilon_T^{i-1}) \frac{e\rho C_p}{\Delta t} \propto \phi$$



Heat Flux Uncertainty : Error Factors

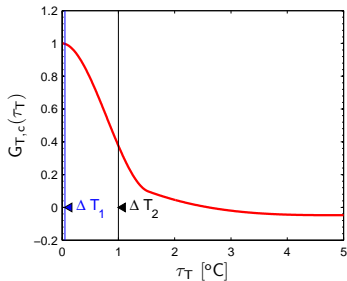
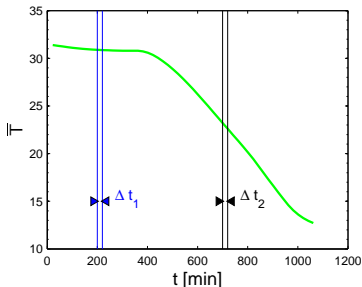


Orders of Magnitude

$$\Delta T_1 < \Delta T_2$$



Heat Flux Uncertainty : Error Factors

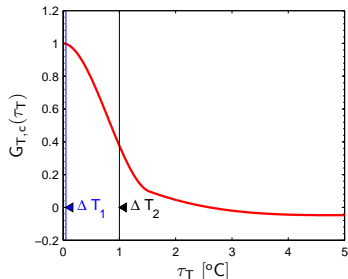
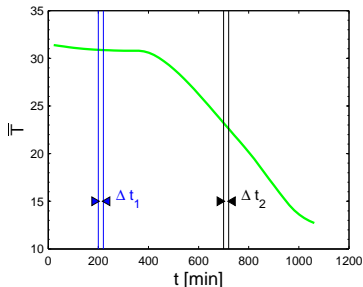


Orders of Magnitude

$$\Delta T_1 < \Delta T_2$$

$$\phi_1 < \phi_2$$

Heat Flux Uncertainty : Error Factors



Orders of Magnitude

$$\Delta T_1 < \Delta T_2$$

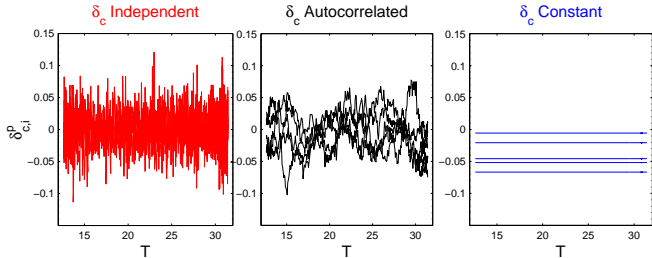
$$\phi_1 < \phi_2$$

$$G(\Delta T_1) > G(\Delta T_2)$$

$$(\varepsilon_T^i - \varepsilon_T^{i-1})_1 < (\varepsilon_T^i - \varepsilon_T^{i-1})_2$$

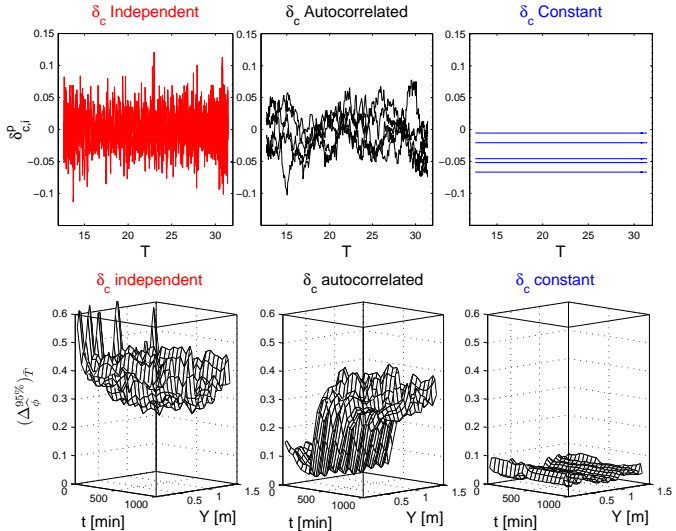
$$(\varepsilon_\phi)_1 < (\varepsilon_\phi)_2$$

Heat Flux Uncertainty : Error Factors



Heat Flux Uncertainty : Error Factors

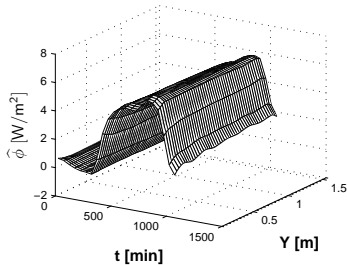
M0 : Foreword
M1 : Estimator
Uncertainty
M2 : Monte Carlo
Method
A0 : Presentation
Study Case
A1 : Temperature
Uncertainties
A2 : IT
Uncertainties
NIP Generation
Temperature
Uncertainty
Heat Flux Uncertainty



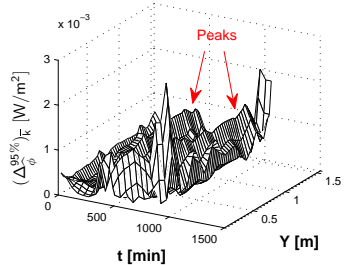


Heat Flux Uncertainty : Error Factors

IM Mean Heat Flux



Uncertainty associated to k

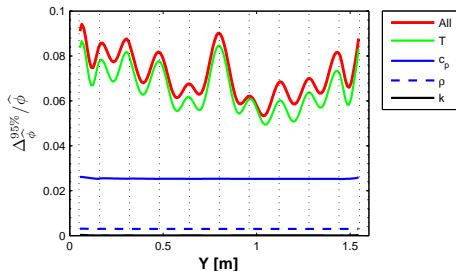


Heat Flux Uncertainty associated to k

- Peaks when $\hat{\phi}$ varies



Heat Flux Uncertainty : Error Factors



Orders of Magnitude

- Error in k negligible
- Error Mainly due to the Temperature Uncertainties



Centre de
Thermique de
Lyon

Thanks For Your Attention

