

# Workshop on Bayesian calibration using UQLab

WINDTRUE: WP2

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CWI

## UQLab and Aeromodule installation test

- ▶ UQLab
  - ▶ Set the Matlab layout to **Default**.
  - ▶ Type **uqlab** in command window.
  - ▶ Change the path in **config.m** file located in the **sensitivity\_analysis** folder. The path will be where your personal UQLab licence is stored.
  - ▶ Type **uqlab** in command window.
- ▶ Aeromodule
  - ▶ Run the **ECNAero** executable to check whether your Aeromodule licence is activated.  
Folder: sensitivity\_analysis/AEROModule/NM80\_calibrate

# Ingredients to perform Bayesian calibration

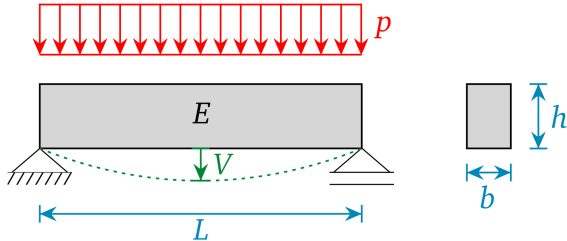


- ▶ Experimental data
- ▶ Forward model
- ▶ Likelihood/Discrepancy
- ▶ Prior

# Examples

- ▶ Simple beam model calibration
- ▶ ECN AeroModule calibration
- ▶ Exercise

## Simple beam model calibration



Calibration/surrogate\_beam.m

Calibration/uq\_SimplySupportedBeam.m

## Problem setup

### Ingredients for Bayesian calibration

- ▶ Data ( $\mathbf{y}$ ): {12.84, 13.12, 12.13, 12.19, 12.67} (mm)
- ▶ Forward model:  $V = \frac{5}{32} \frac{pL^4}{Ebh^3}$
- ▶ Likelihood:  $\pi(\mathbf{y}|\theta) : \sigma^2 = 10^{-6}$

- ▶ Prior distribution:

$\theta$	$\pi(\theta)$
b (m)	0.15
h (m)	0.3
L (m)	5
p (kN/m)	$\mathcal{N}(0.012, 0.0006)$
E (MPa)	$\mathcal{LN}(30000, 4500)$

# Problem setup

## Ingredients for Bayesian calibration

```
myData.y = [12.84; 13.12; 12.13; 12.19; 12.67]/1000; % (m)
myData.Name = 'Mid-span deflection';

ModelOpts.mFile = 'uq_SimplySupportedBeam';
ModelOpts.isVectorized = true;
myForwardModel = uq_createModel(ModelOpts);

DiscrepancyOpts.Parameters = 1e-6; % known discrepancy variance
BayesOpts.Discrepancy = DiscrepancyOpts;

PriorOpts.Marginals(1).Name = 'b'; % beam width
PriorOpts.Marginals(1).Type = 'Constant';
PriorOpts.Marginals(1).Parameters = {0.15}; % (m)

PriorOpts.Marginals(2).Name = 'h'; % beam height
PriorOpts.Marginals(2).Type = 'Constant';
PriorOpts.Marginals(2).Parameters = {0.3}; % (m)

PriorOpts.Marginals(3).Name = 'L'; % beam length
PriorOpts.Marginals(3).Type = 'Constant';
PriorOpts.Marginals(3).Parameters = 5; % (m)

PriorOpts.Marginals(4).Name = 'E'; % Young's modulus
PriorOpts.Marginals(4).Type = 'LogNormal';
PriorOpts.Marginals(4).Moments = {30000 4500}; % (MPa)

PriorOpts.Marginals(5).Name = 'p'; % uniform load
PriorOpts.Marginals(5).Type = 'Gaussian';
PriorOpts.Marginals(5).Moments = {0.012 0.012*0.05}; % (kN/m)

myPriorDist = uq_createInput(PriorOpts);
```

# Prior and posterior distribution

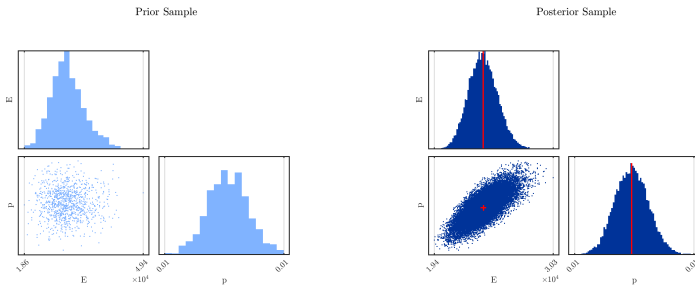


Figure: Prior and posterior samples.



# MAP estimate

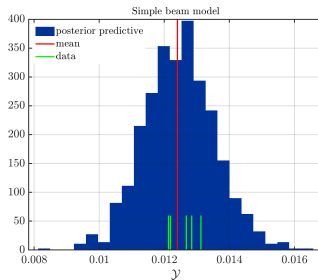
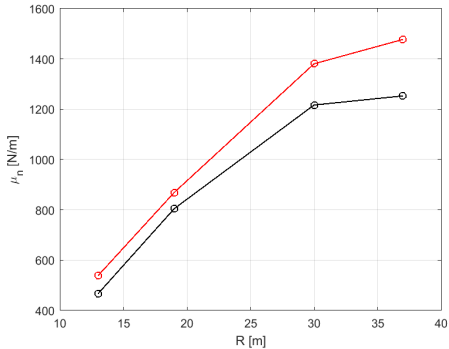


Figure: Bayesian estimate using posterior distribution against the experimental data.

MAP estimate:  $\mathbb{E}[\theta|y]$

p	E
$2.4 \times 10^4$	0.0012

# ECN AeroModule calibration



sensitivity\_analysis/cases/aero\_module/NM80\_calibrate.m

sensitivity\_analysis/cases/aero\_module/initialize\_calibration.m

sensitivity\_analysis/testCalibration.m

## Problem setup

### Ingredients for Bayesian calibration

- ▶ Axial force data  $(\mathbf{y}) = \{\mathbf{y}_1, \mathbf{y}_2, \mathbf{y}_3, \mathbf{y}_4\}$
- ▶ Forward model: ECN Aero-Module
- ▶ Likelihood:  $\pi(\mathbf{y}|\theta) = \prod_{i=1}^n \mathcal{N}[\mathbf{y}_i - f_i(\theta)]^2$

- ▶ Prior distribution:

$\theta$	$\pi(\theta)$
$P$	Constant
$C_L$	Uniform
$\sigma^2$	Uniform

# Problem setup

Marco's (ECN) script for reading the data in N/m

```
filename_exp = {'../../Experimental/WINDTRUE/raw.dat'};  
output_raw = read_exp_data(filename_exp, 2);  
% Because the model has different discrepancy options at different  
% radial locations,  
% the measurement data is stored in four different data structures:  
Data(1).y = mean(output_raw.Fy03); % [N/m]  
Data(1).Name = 'Fy03';  
Data(1).MOMap = 1; % Model Output Map 1
```

Name of Matlab file representing the model

```
Model.mHandle = @aero_module_calibration;  
% Optionally, one can pass parameters to model stored in the cell  
% array P  
P = getParameterAeroModule(turbineName);  
Model.Parameters = P;  
Model.isVectorized = false;
```

```
DiscrepancyPriorOpts1.Name = 'Prior of sigma 1';  
DiscrepancyPriorOpts1.Marginals(1).Name = 'Sigma1';  
DiscrepancyPriorOpts1.Marginals(1).Type = 'Uniform';  
DiscrepancyPriorOpts1.Marginals(1).Parameters =  
    [0.5*std(output_raw.Fy03), 1.5*std(output_raw.Fy03)];  
DiscrepancyPrior1 = uq_createInput(DiscrepancyPriorOpts1);
```

```
DiscrepancyOpts(1).Type = 'Gaussian';  
DiscrepancyOpts(1).Prior = DiscrepancyPrior1;
```

```
uncertain_params = {{'CL',1, 0.2}, {'CL',2, 0.2}, {'CL',3, 0.3},  
{'CL',4,0.3}};  
QoI = 'force'; % Force at different radial stations  
aero_module_outputfile = 'Bln_BEM.txt'; % Aero-Module data to be  
calibrated
```

# Problem setup

Switch for Bayesian analysis with the AeroModule or with the surrogate model

```
Bayes_full = 0; % 0: use surrogate model (PCE); 1: run full model for
Bayes (Computationally expensive!)
```

```
% If Bayes_full = 0, we need to specify options for loading a
surrogate model
```

```
Surrogate_model_type = 0; % 0: Uses a stored PCE surrogate model, 1:
create surrogate model
```

```
% Options for loading a surrogate model
```

```
Surrogate_model_filename = 'surrogate/PCE_60.mat'; % Specify the
surrogate model file to be used
```

```
% Options for creating a surrogate model
```

```
% These are used if Bayes_full = 0 and Surrogate_model_type = 1
```

```
MetaOpts.Type = 'Metamodel';
```

```
MetaOpts.MetaType = 'PCE';
```

```
MetaOpts.Method = 'LARS'; % Quadrature, OLS, LARS
```

```
MetaOpts.ExpDesign.Sampling = 'LHS';
```

```
MetaOpts.ExpDesign.NSamples = 60;
```

```
MetaOpts.Degree = 1:4;
```

```
MetaOpts.TruncOptions.qNorm = 0.75;
```

## MCMC parameters

```
Solver.Type = 'MCMC';
```

```
% MCMC algorithms available in UQLab
```

```
MH = 0; % Metropolis-Hastings
```

```
AM = 0; % Adaptive Metropolis
```

```
AIES = 1; % Affine invariant ensemble
```

```
HMC = 0; % Hamilton Monte Carlo
```

# Convergence diagnostics

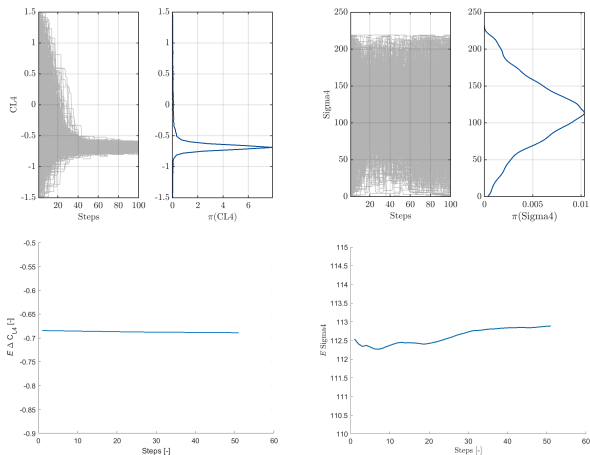


Figure: Trace plots and convergence for  $y_4$

# Prior and posterior distribution

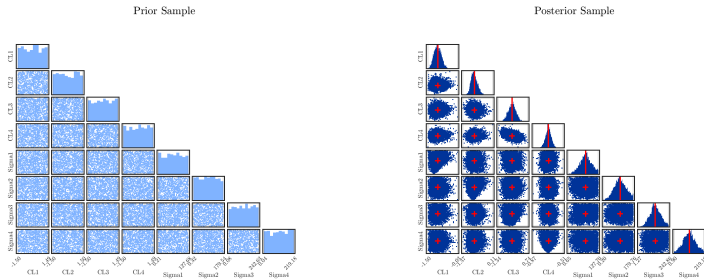


Figure: Prior and posterior samples.

# MAP estimate

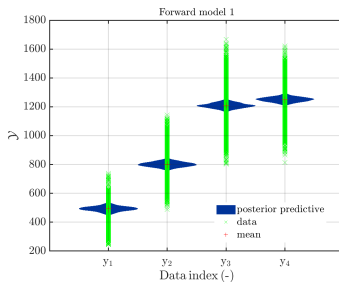


Figure: Violin plot showing distributions of Bayesian prediction against the DANAERO data.

MAP estimate:  $\mathbb{E}[\theta|y]$

$\Delta C_{L1}$	$\Delta C_{L2}$	$\Delta C_{L3}$	$\Delta C_{L4}$	$\sigma_1^2$	$\sigma_2^2$	$\sigma_3^2$	$\sigma_4^2$
-0.2025	-0.1604	-0.1087	-0.2067	73.7666	100.2012	138.4856	112.8861



# Calibrated polars

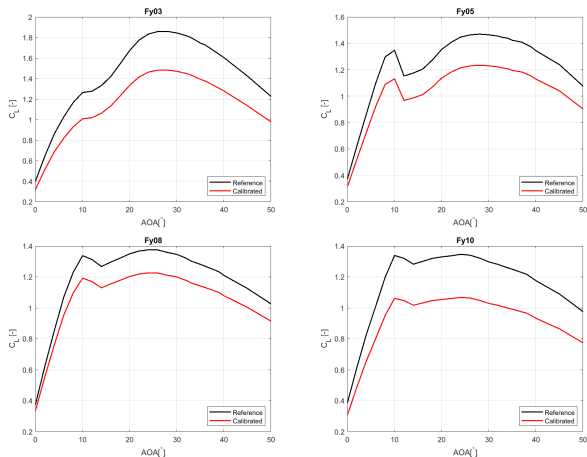
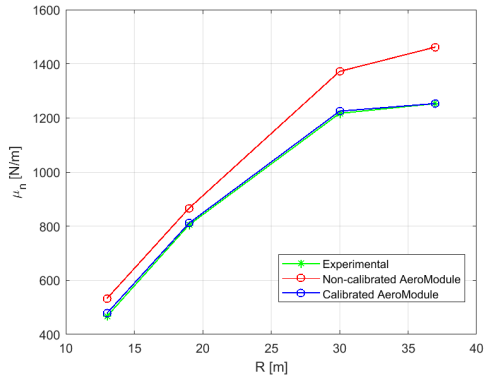


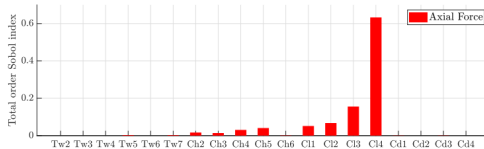
Figure:  $C_L$  polars comparison.

## QoI validation



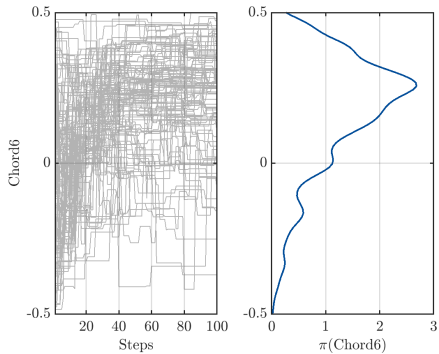
**Figure:** Comparison of axial force obtained using: experimental, non-calibrated Aero-Module run and calibrated Aero-Module run.

# Exercise



1. Calibrate Aero-Module using Cl and Ch as the uncertain parameters
2. Create a surrogate model using 50 samples and save the model in the surrogate folder

# Note



$$\text{Ch6 calibrated} = \text{Ch6 reference} (1 + (0.264 \times 0.2))$$

## Relevant cases?

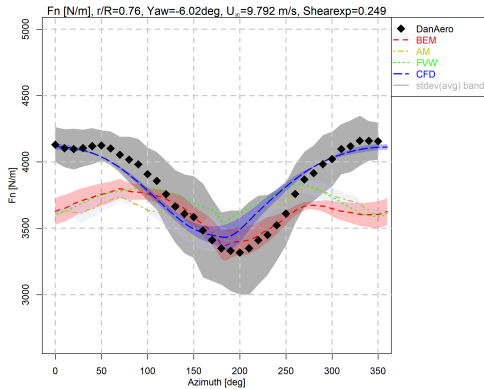


Figure: Comparison of axial force in the azimuthal direction.