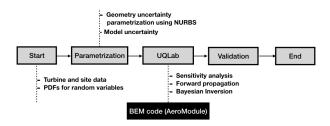
# Bayesian calibration applied to DANAERO data WINDTRUE: WP2

Vinit Dighe & Benjamin Sanderse March 16, 2020

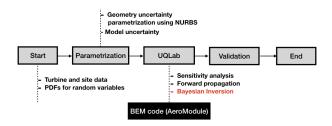


#### WINDTRUE workflow





#### WINDTRUE workflow





#### Results from WP1

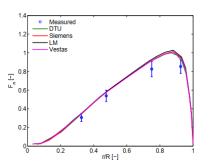


## Kumar et al. (2020)

The results highlight amongst others the importance of the lift coefficient, especially for the axial force prediction.



#### DANAERO test case



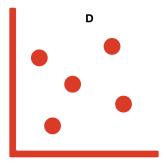
- ► D: Comparison of measurements and BEM computations of axial force in non-sheared inflow
- ► Forward model:

$$F_n^* = 0.5 \rho W^2 c (C_L \sin\beta + C_D \cos\beta)$$

 $\bullet \ \theta = [C_{L1}, \ C_{D1}, \ C_{L2}, \ C_{D2}, \ C_{L3}, \ C_{D3}, \ C_{L4}, \ C_{D4}]$ 

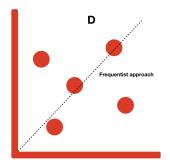


### Inferential statistics





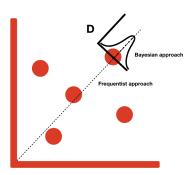
#### Inferential statistics



▶ Frequentist inference:  $\mathbb{E}(\theta)$ 



#### Inferential statistics

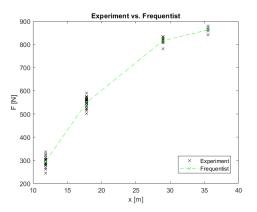


- ▶ Frequentist inference:  $\mathbb{E}(\theta)$
- ► Bayesian inference:

$$P(\theta|\mathsf{D}) = \frac{P(\mathsf{D}|\theta) \times P(\theta)}{P(\mathsf{D})}$$



### Frequentist inference

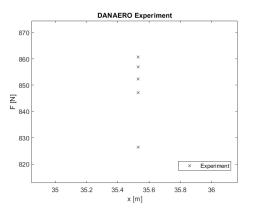


► Frequentist inference (Ordinary least-squares regression):

$$\theta_{OLS} = (X^T X)^{-1} X^T D,$$
 $X = X_{ij}$  is the design matrix



# Bayesian inference

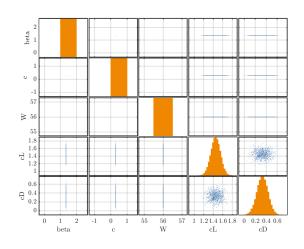


► Radial location: 35.53 m

▶ Number of data points (N) = 5 (Remember N> $\theta$ )

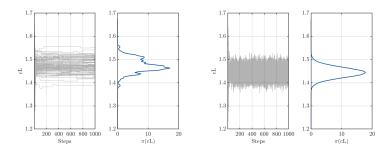


# Bayesian inference - Prior distribution





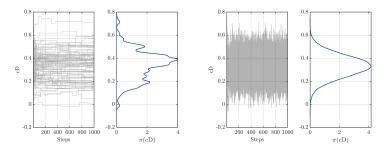
## Bayesian inference - MCMC algorithms



Figures showing trace plots for  $C_L$  using Metropolis-Hastings (left) and Affine-invariant ensemble sampling (right) algorithms.



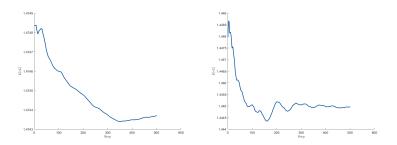
# Bayesian inference - MCMC algorithms



Figures showing trace plots for  $C_D$  using Metropolis-Hastings (left) and Affine-invariant ensemble sampling (right) algorithms.



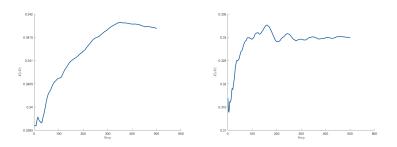
# Bayesian inference - Assessing convergence



Figures showing convergence plots for  $C_L$  using Metropolis-Hastings (left) and Affine-invariant ensemble sampling (right) algorithms.



## Bayesian inference - Assessing convergence



Figures showing convergence plots for  $C_D$  using Metropolis-Hastings (left) and Affine-invariant ensemble sampling (right) algorithms.



# Bayesian inference - Posterior distribution

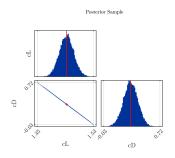
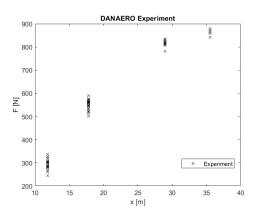


Figure showing posterior distribution obtained using Affine-invariant ensemble sampling algorithm.

	$C_{L4}$	$C_{D4}$
$\mathbb{E}( heta)$	1.4821	0.3421
MAP	1.4828	0.3391

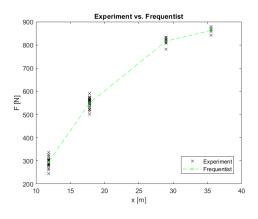


## Frequentist vs. Bayesian approach



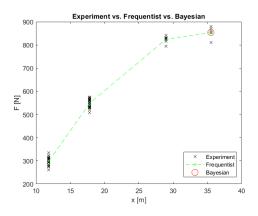


## Frequentist vs. Bayesian approach





## Frequentist vs. Bayesian approach





## Concluding remarks

- ► Bayesian calibration framework using UQLab
  - ► 4 different MCMC algorithms
  - ► Gelman-Rubin diagnostics: Assess convergence
  - ► Trace plots: Evolution of Markov chain
  - ► Acceptance rate: Quantitative indication of accepted samples
- ► DANAERO test case



#### Next steps

- ► Experimental data (D)
- ightharpoonup Choose relevant calibration parameters ( $\theta$ )
- ► BEM code for the forward model
- Choosing prior distribution for Bayesian calibration (Good engineering estimate)

