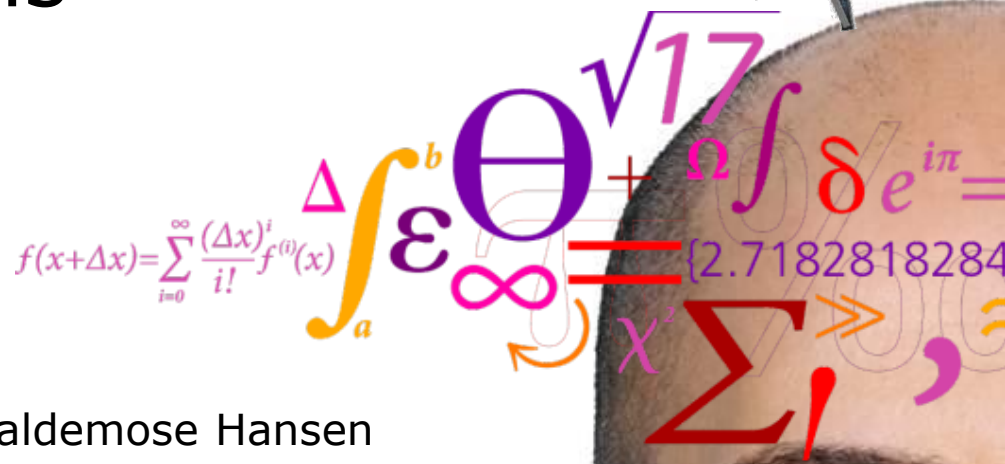


# Fuga

- Validating a wake model for offshore wind farms



$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$

$\int_a^b \varepsilon \Theta$

$\sqrt{17}$

$\Omega \int \delta e^{i\pi} =$

$[2.7182818284]$

$\chi'$

$\Sigma$

$\gg$

$!$

Søren Ott, Morten Nielsen & Kurt Shaldemose Hansen

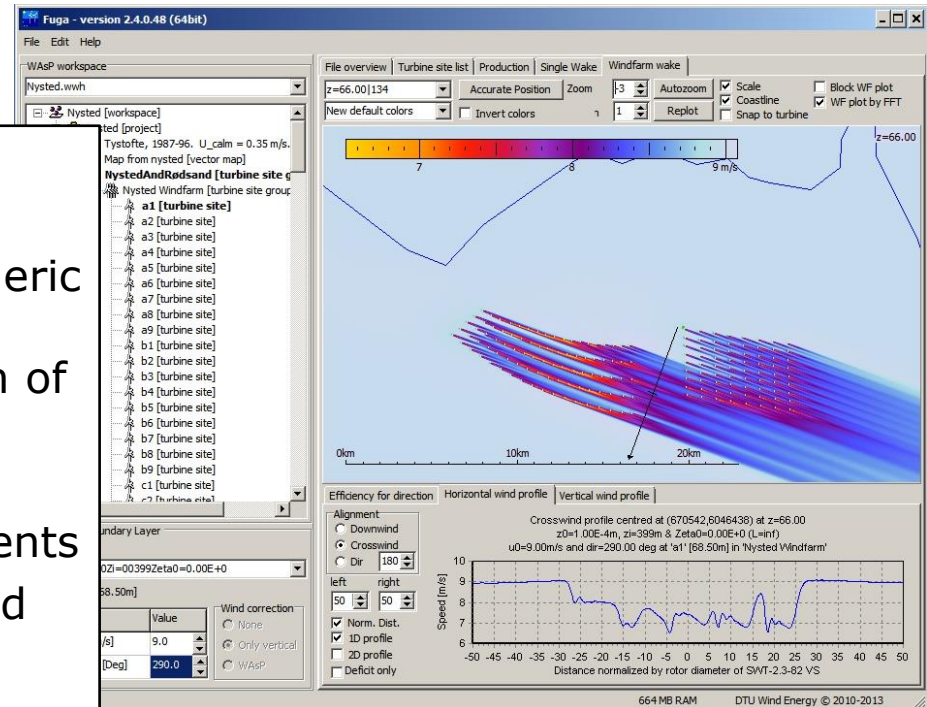
# Outline

- What is Fuga?
- Model validation: which assumptions are tested?
- Met data interpretation: Is it important?
- Simple models for the impact of large scales
- Illustrated with Fuga tests
- Conclusions



# Fuga – features\*

- Solves linearized RANS equations
- Latest version incorporates: atmospheric stability, meandering, effects of non-stationarity and spatial de-correlation of the flow field.
- No computational grid, no numerical diffusion, no spurious pressure gradients
- Integration with WAsP: import of wind climate and turbine data.
- Fast, mixed-spectral solver:
  - $10^6$  times faster than conventional RANS!
  - $10^8$  to  $10^{10}$  times faster than LES!

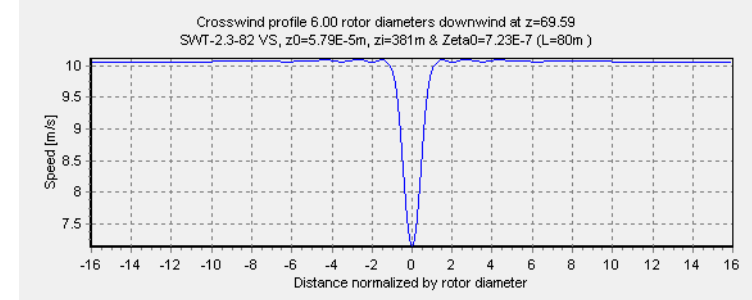
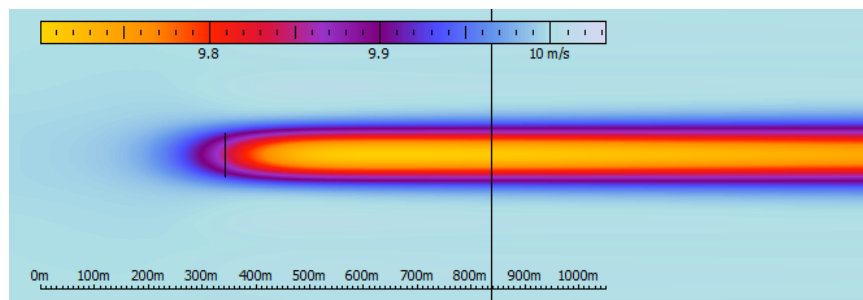
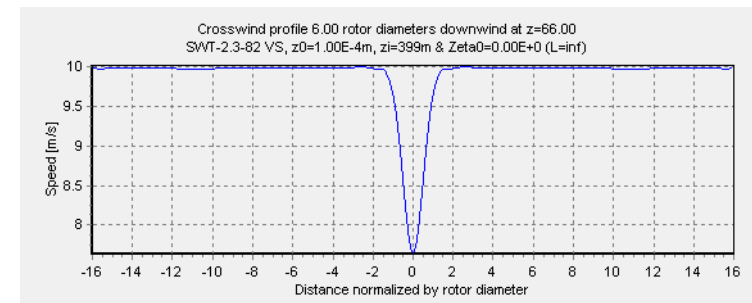
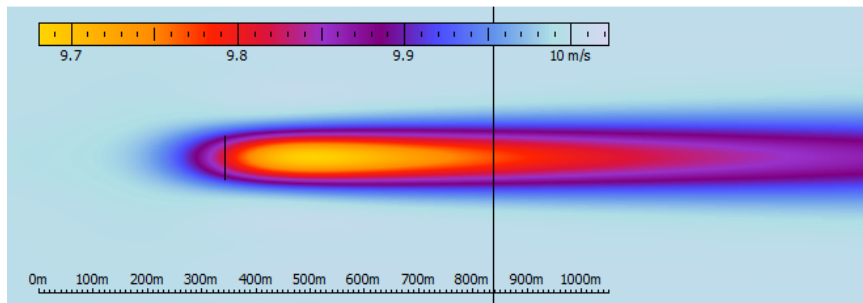
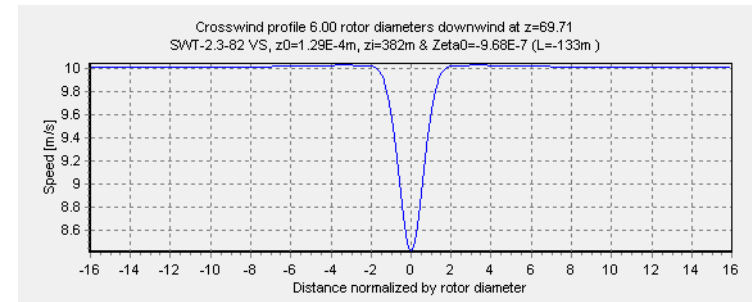
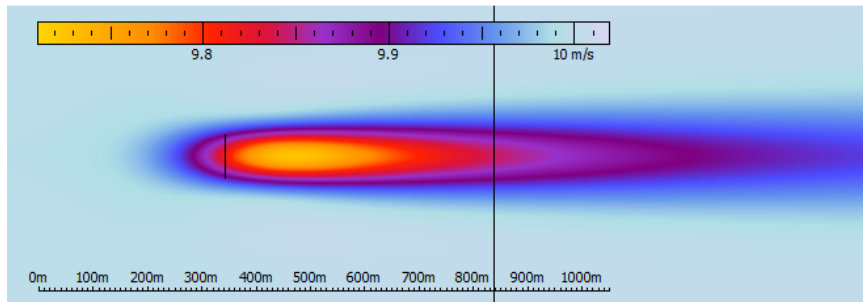


Fuga development  
partly sponsored by

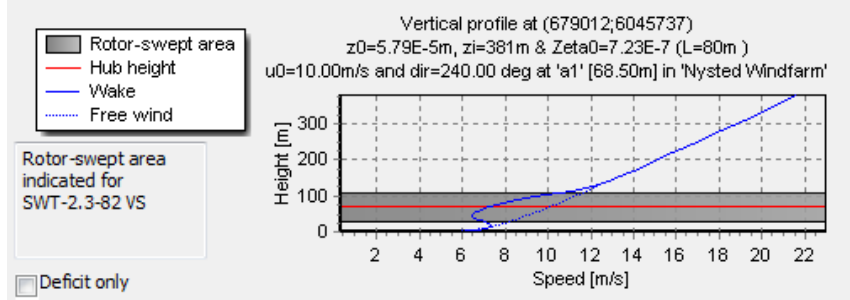
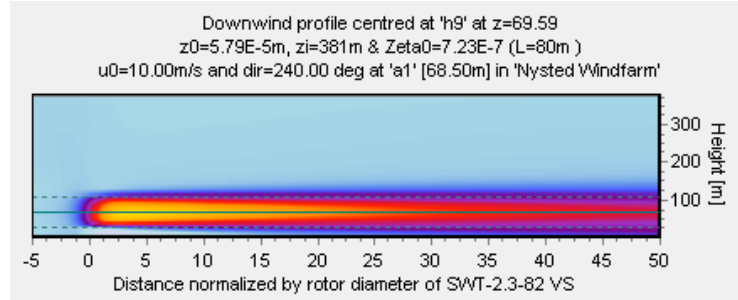
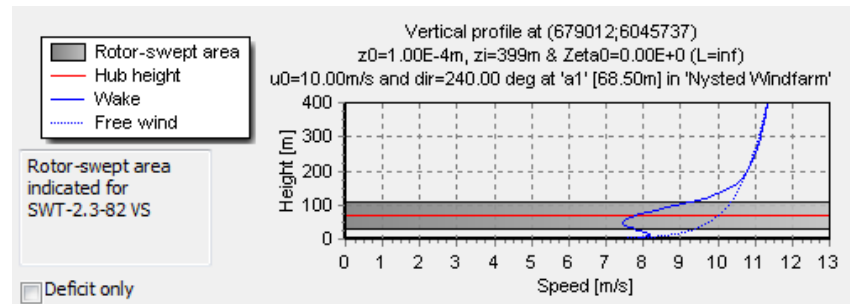
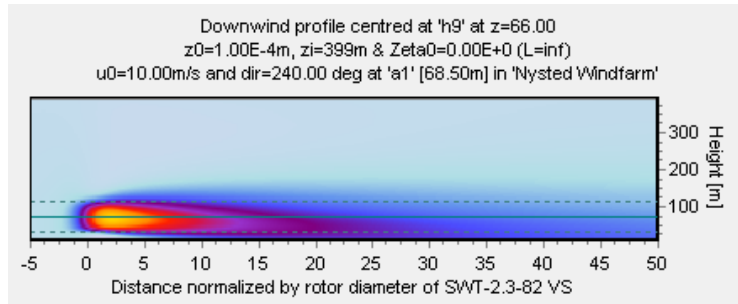
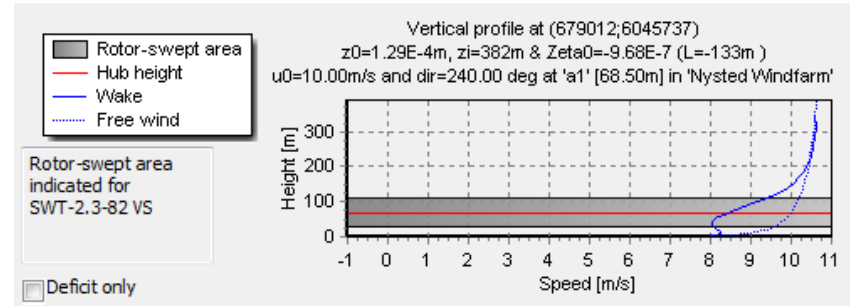
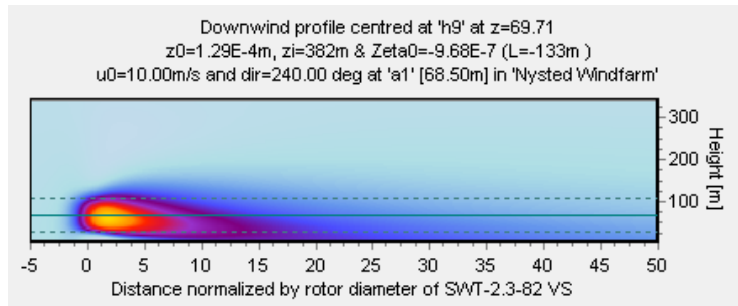


\*Søren Ott, Jacob Berg and Morten Nielsen: 'Linearised CFD Models for Wakes', Risoe-R-1772(EN), 2011

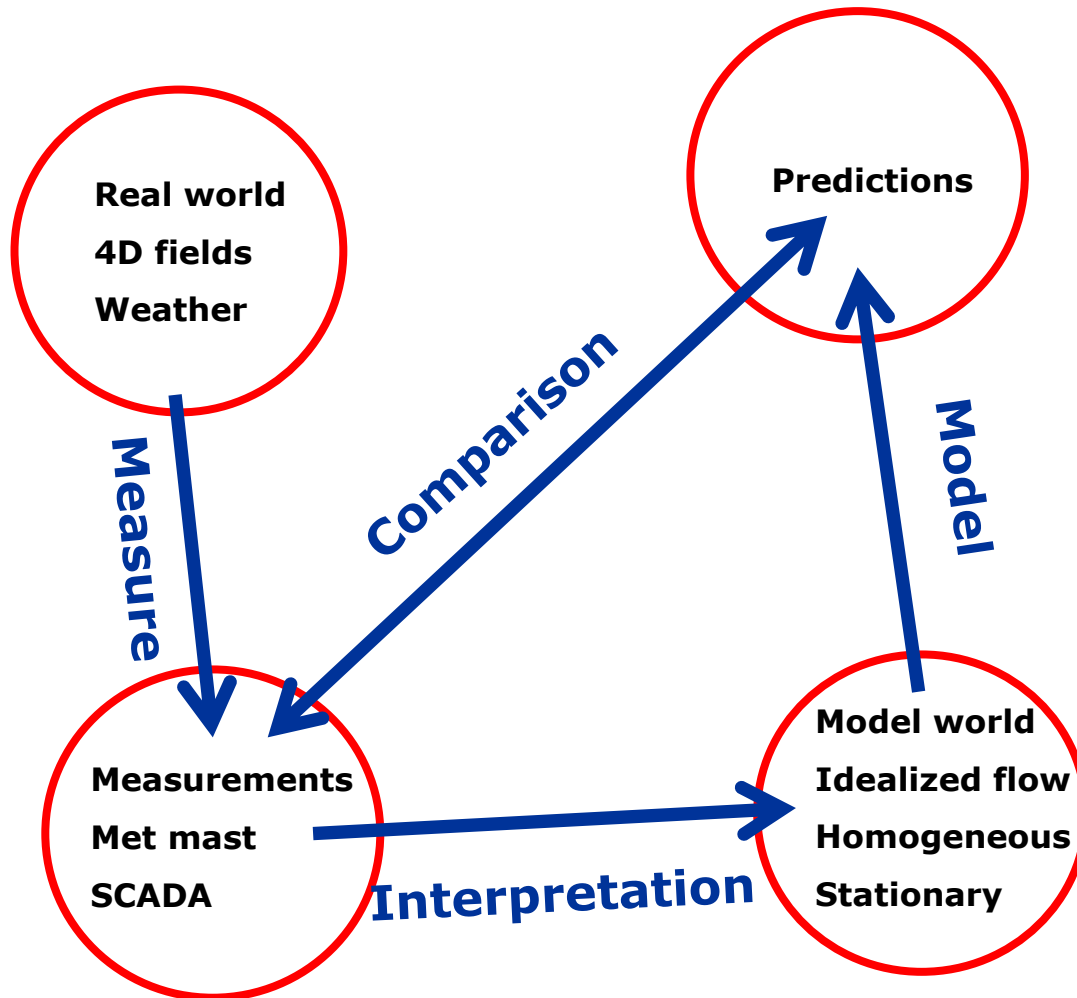
# Variable atmospheric stability – horizontal profiles



# Variable atmospheric stability – vertical profiles



# Model testing and assumptions



Two layers of assumptions:

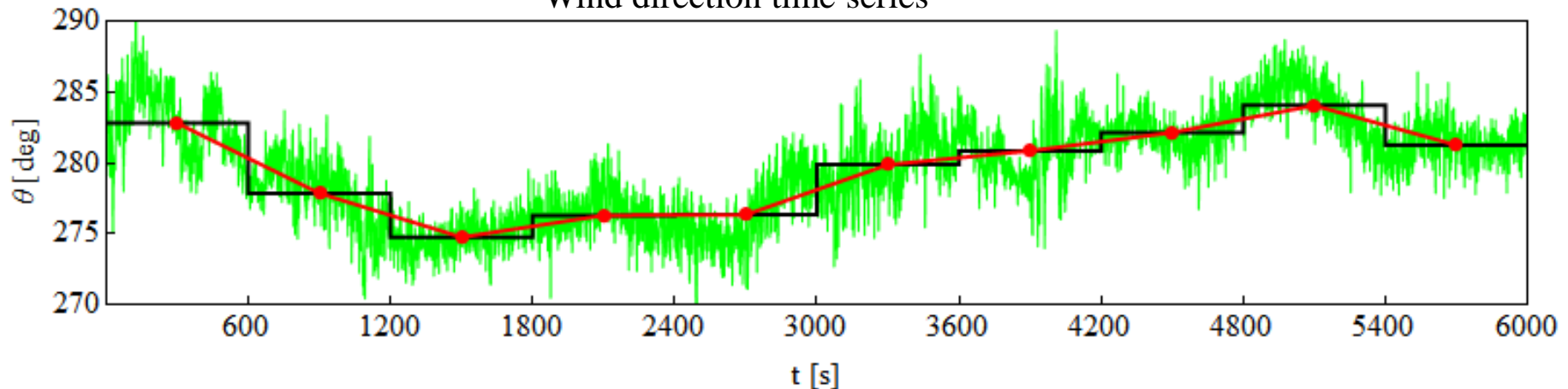
- 1) interpretation of data in terms of an idealized 'model world'
- 2) Flow model assumptions

Controlled experiments test only one assumption at a time!

Is data interpretation important?


# Two different interpretations of met data


Wind direction time series



 Wind direction measurements

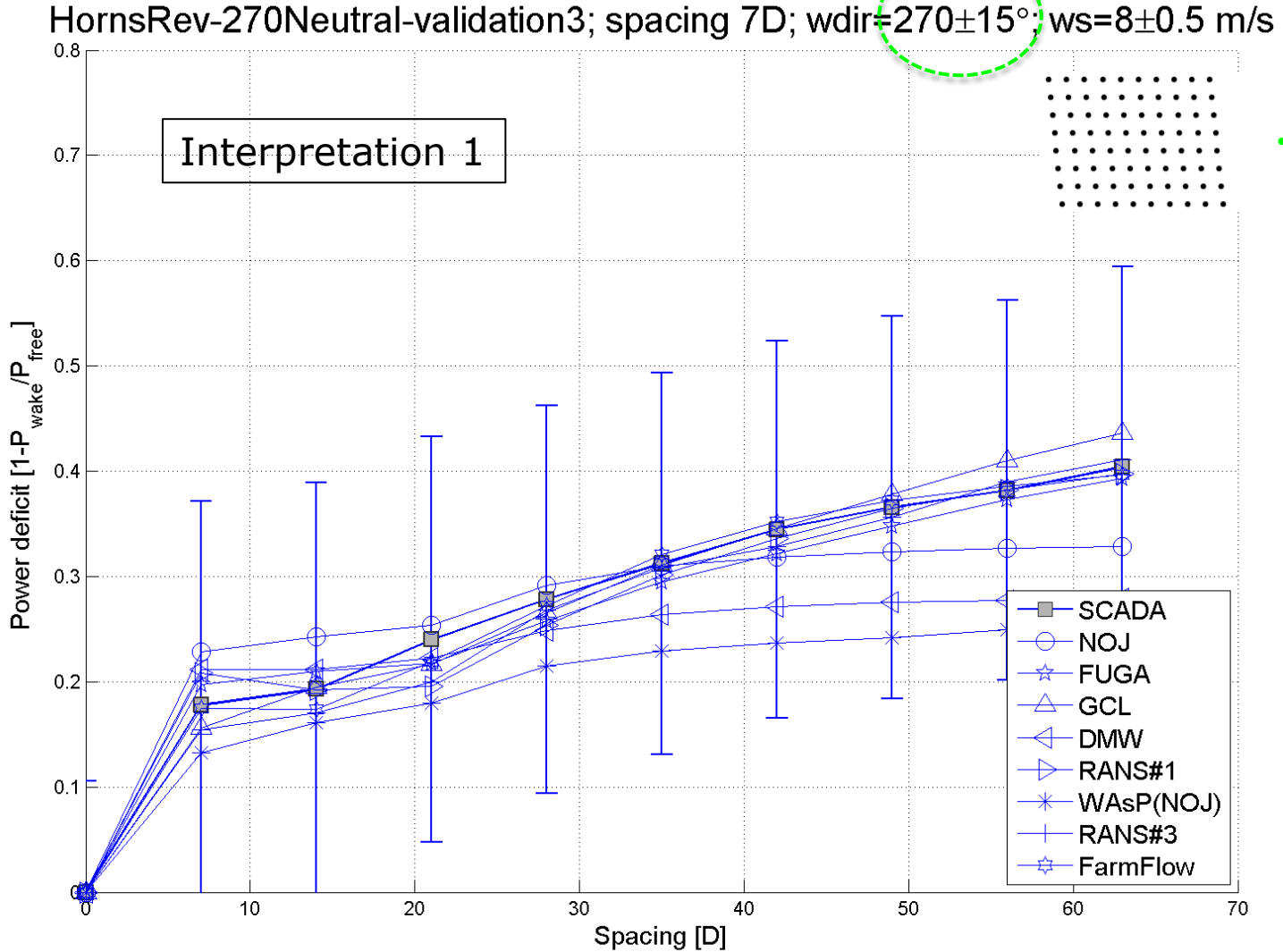
 Ten minutes average wind direction

 **Interpretation 1:** each ten minutes period can be regarded as a piece of a stationary time series with mean = ten minutes average. *We only need to simulate the measured mean direction.*

 **Interpretation 2:** the mean value changes following the red curve. The deviation from the red curve is statistically stationary. *We need to simulate a range of directions for each measure mean direction.*

# Flow cases, part 1- wide inflow sector

Thanks to Kurt S. Hansen for this slide

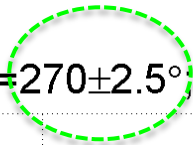




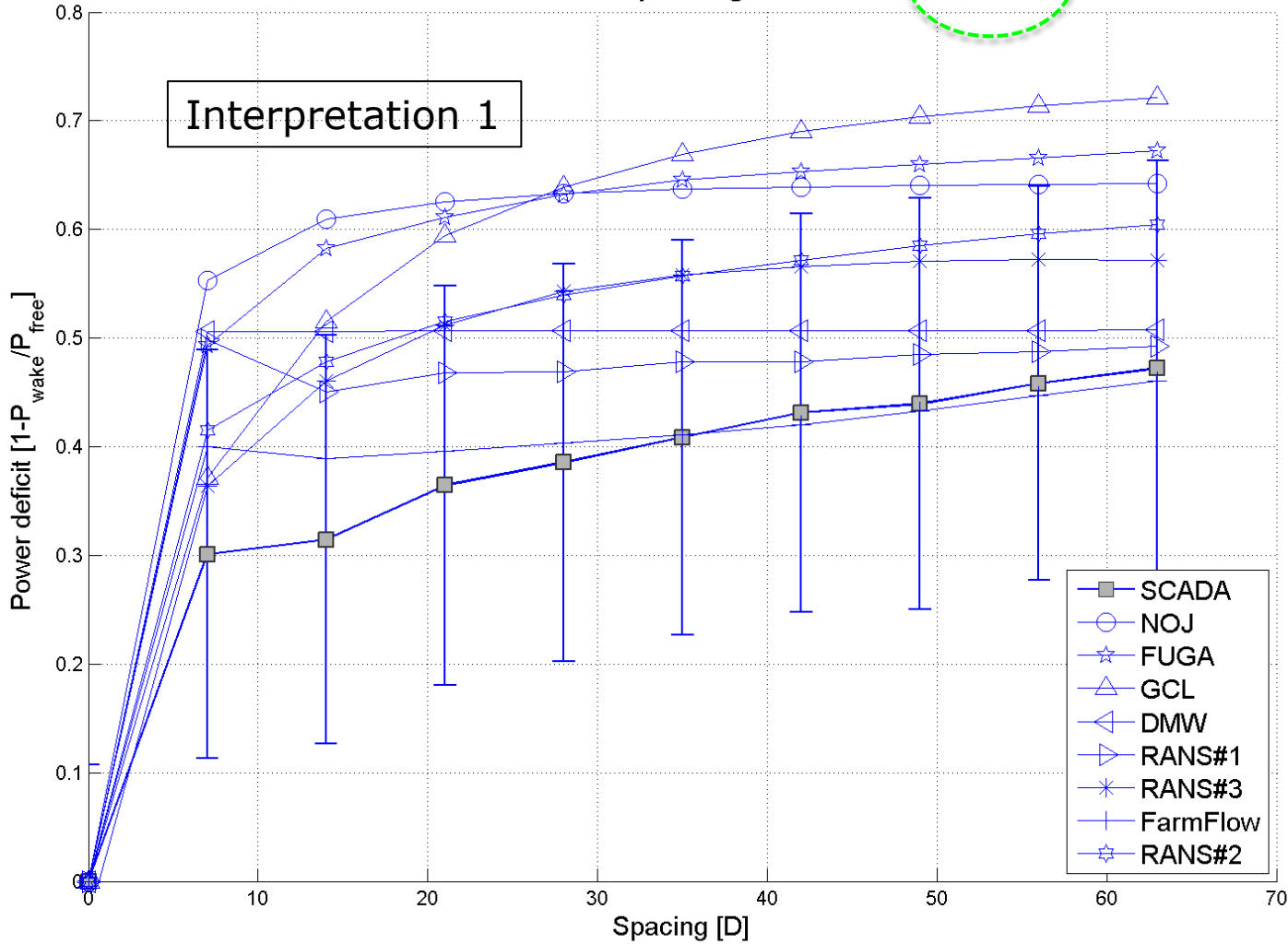
# Flow cases, part 1 – narrow inflow sector



HornsRev-270Neutral-validation1; spacing 7D; wdir=270±2.5° ws=8±0.5 m/s



Interpretation 1

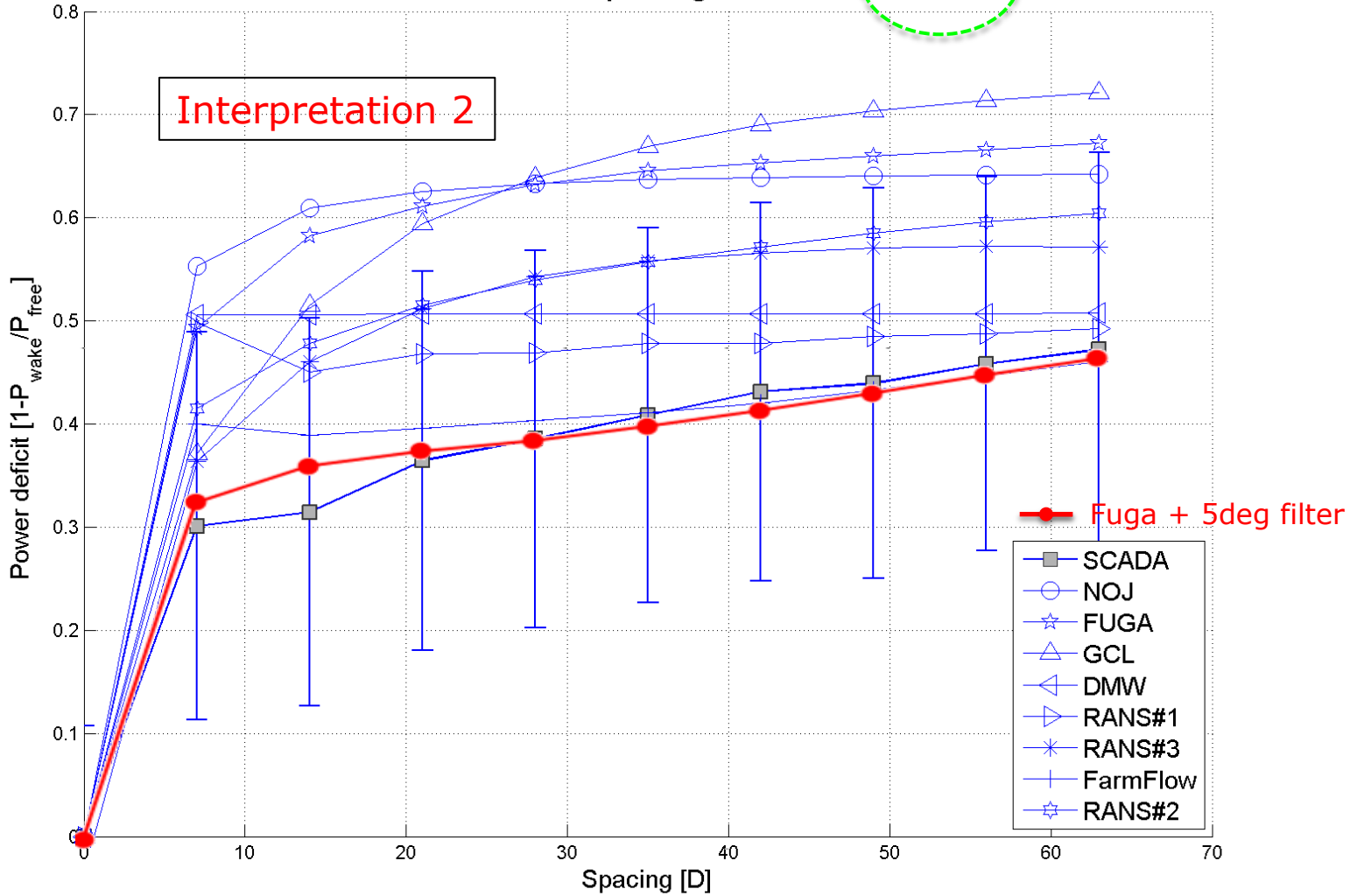


Thanks to Kurt S. Hansen for this slide

# Flow cases, part 1 – narrow inflow sector



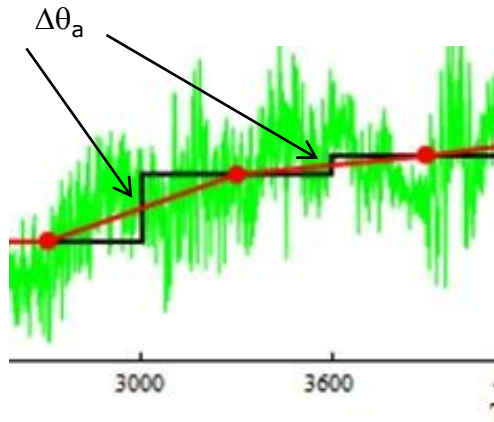
HornsRev-270Neutral-validation1; spacing 7D; wdir= $270 \pm 2.5^\circ$  ws= $8 \pm 0.5$  m/s



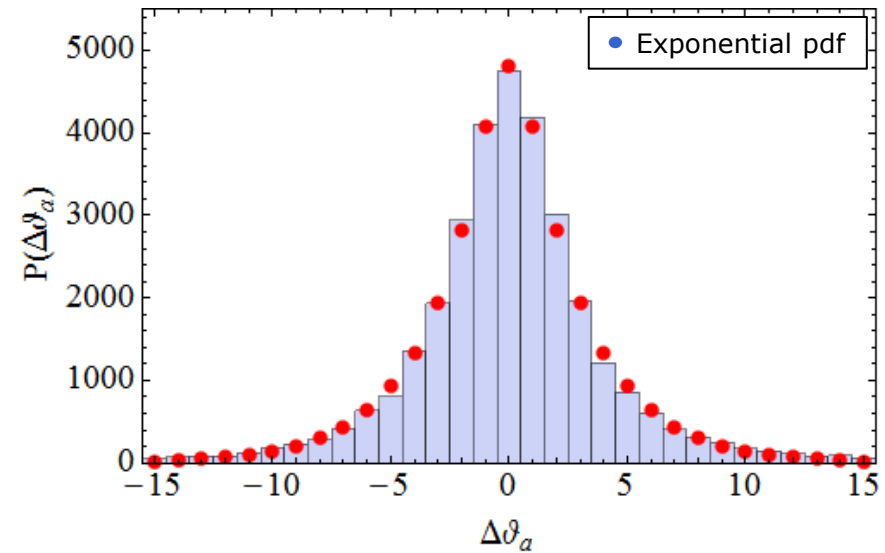
Thanks to Kurt S. Hansen for this slide

# Effect of 'drifting' wind direction can be estimated from data

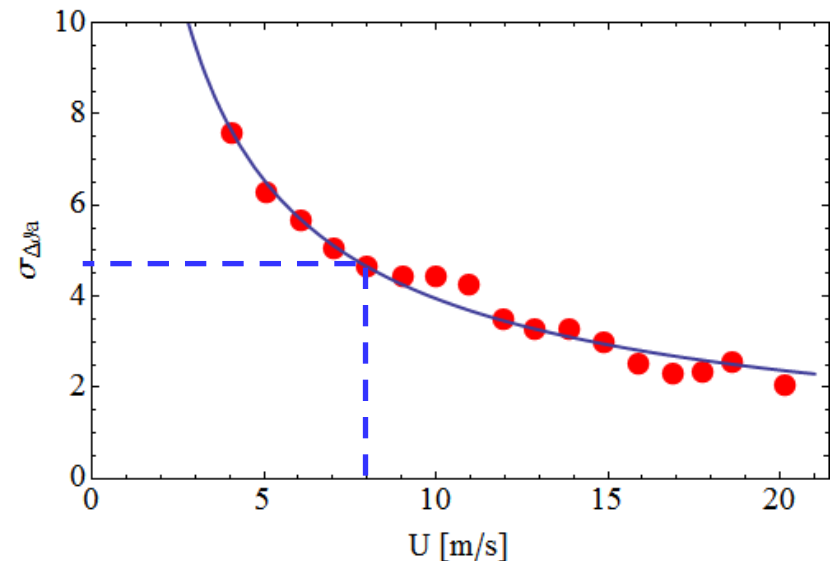
- $\Delta\theta_a$  = Difference of two consecutive ten minutes averages of the wind direction



- Width  $\sigma_{\Delta\theta_a} = \langle (\Delta\theta_a)^2 \rangle^{1/2}$
- $\sigma_{\Delta\theta_a}$  can be obtained from 10 minutes averages.



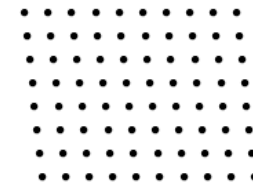
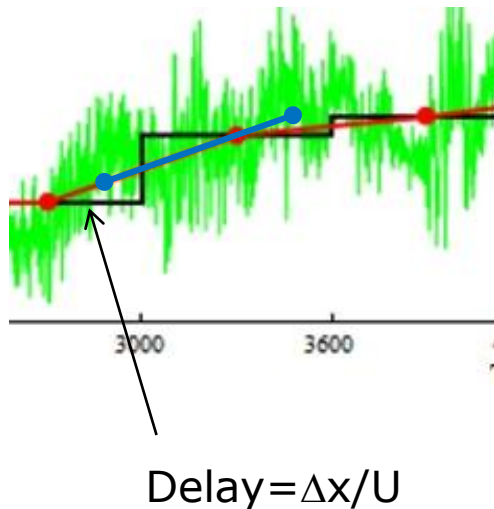
All stabilities



# Spatial de-correlations of wind direction

Problem: wind direction at met mast  $\neq$  wind directions at the turbines.

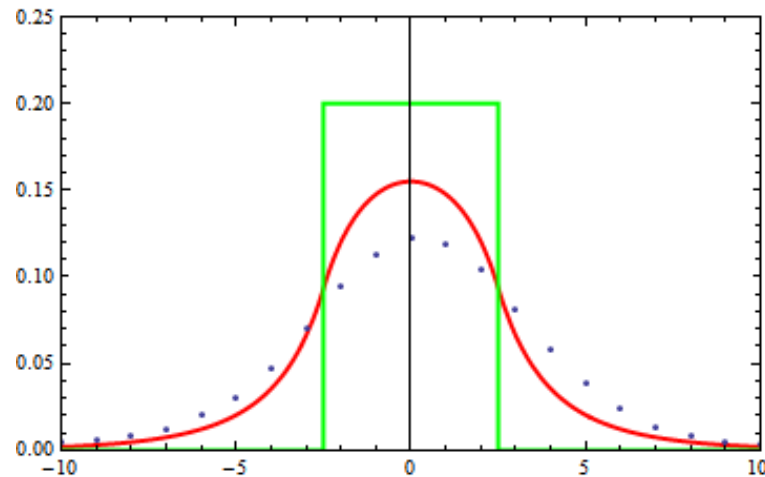
This uncertainty can be estimated using Taylor's hypothesis:



M6



M7

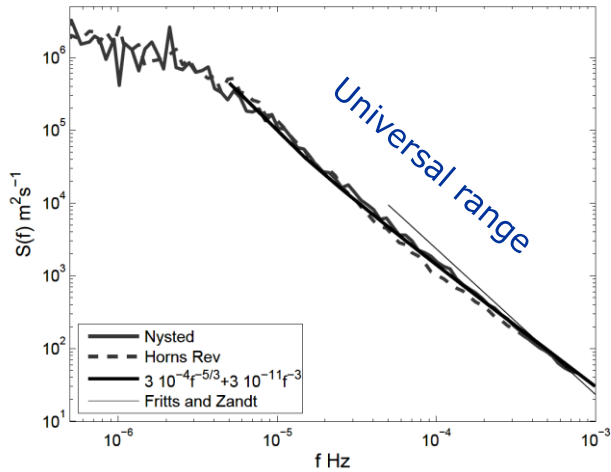


— Pdf at M7

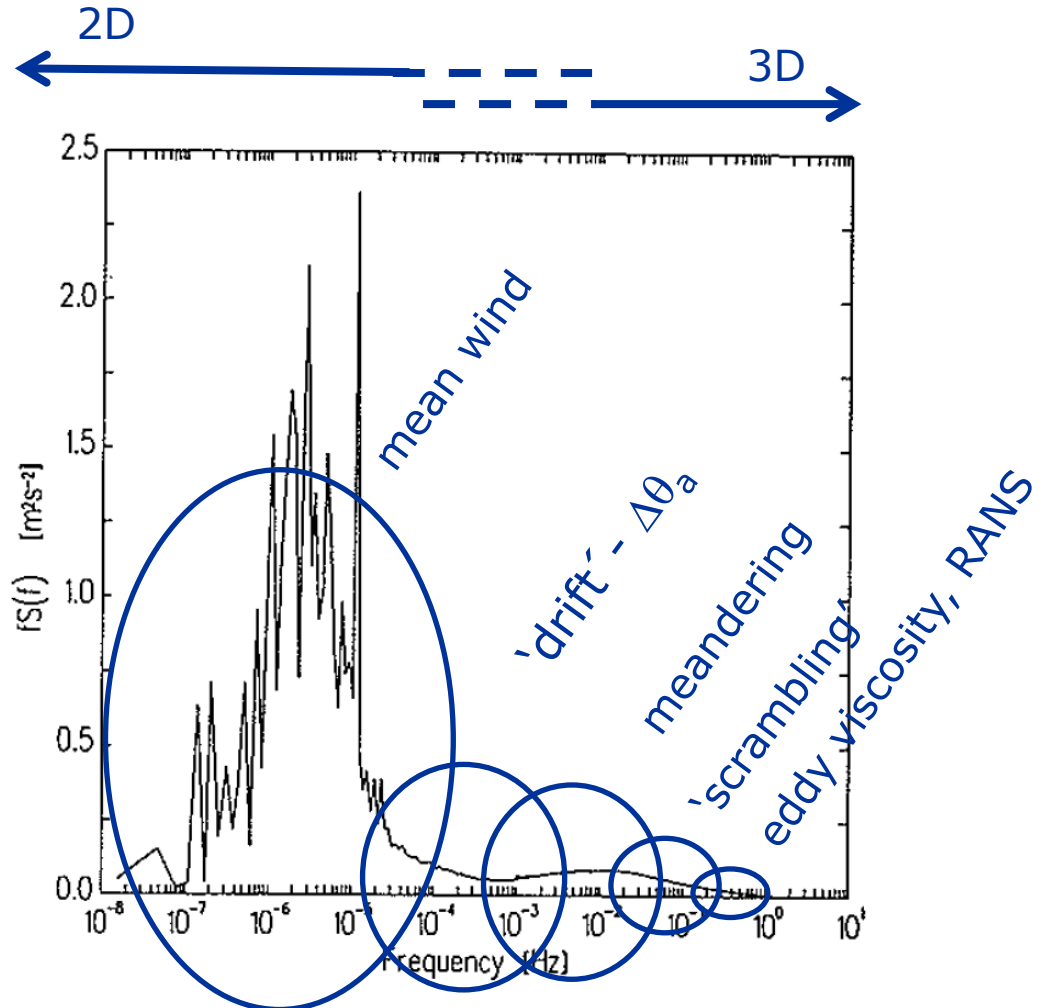
— Predicted pdf at M6

• • • Data

# Spectral regimes



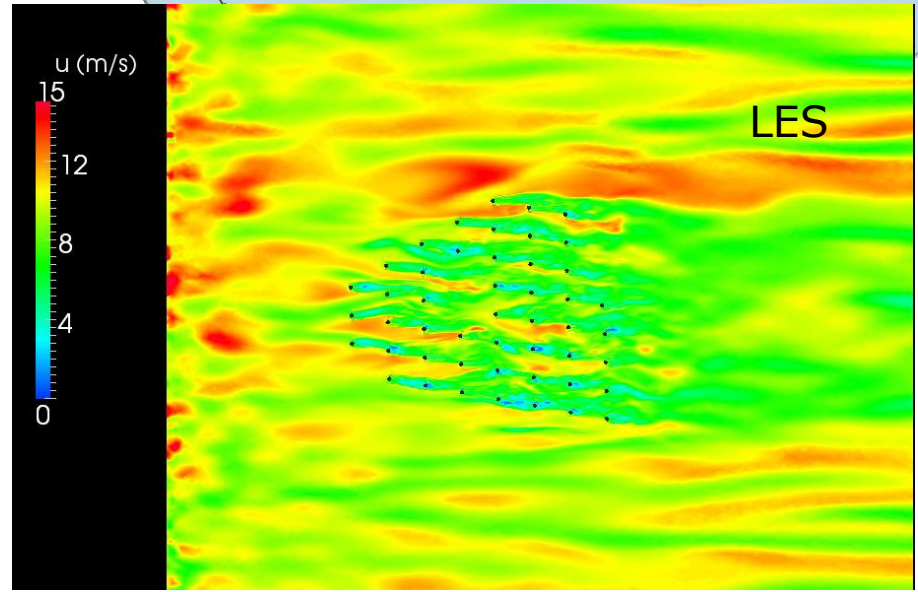
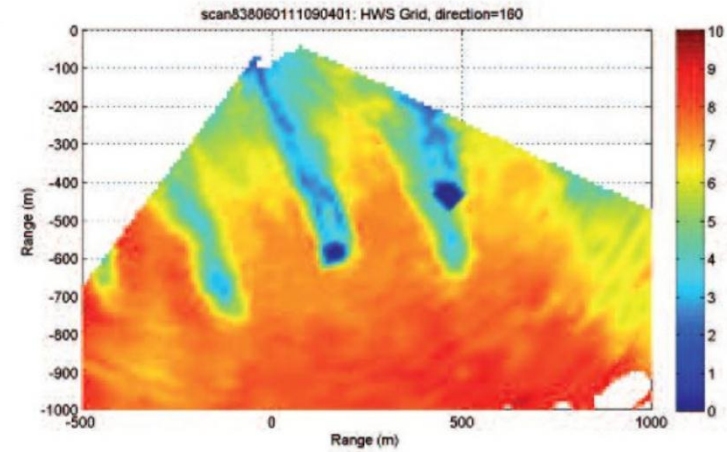
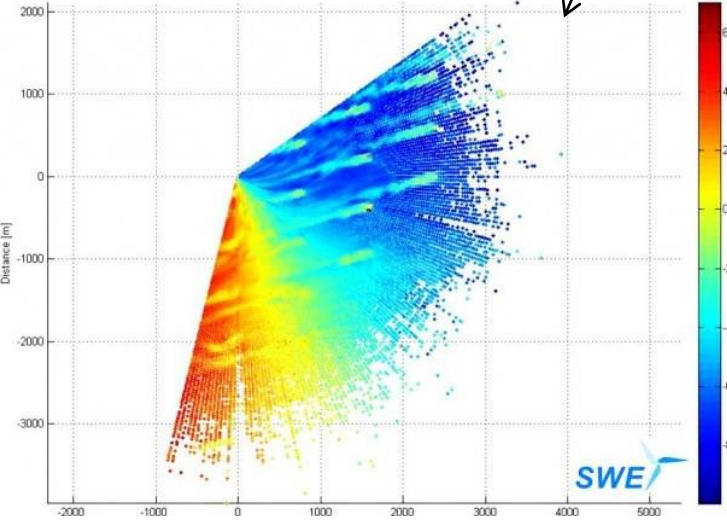
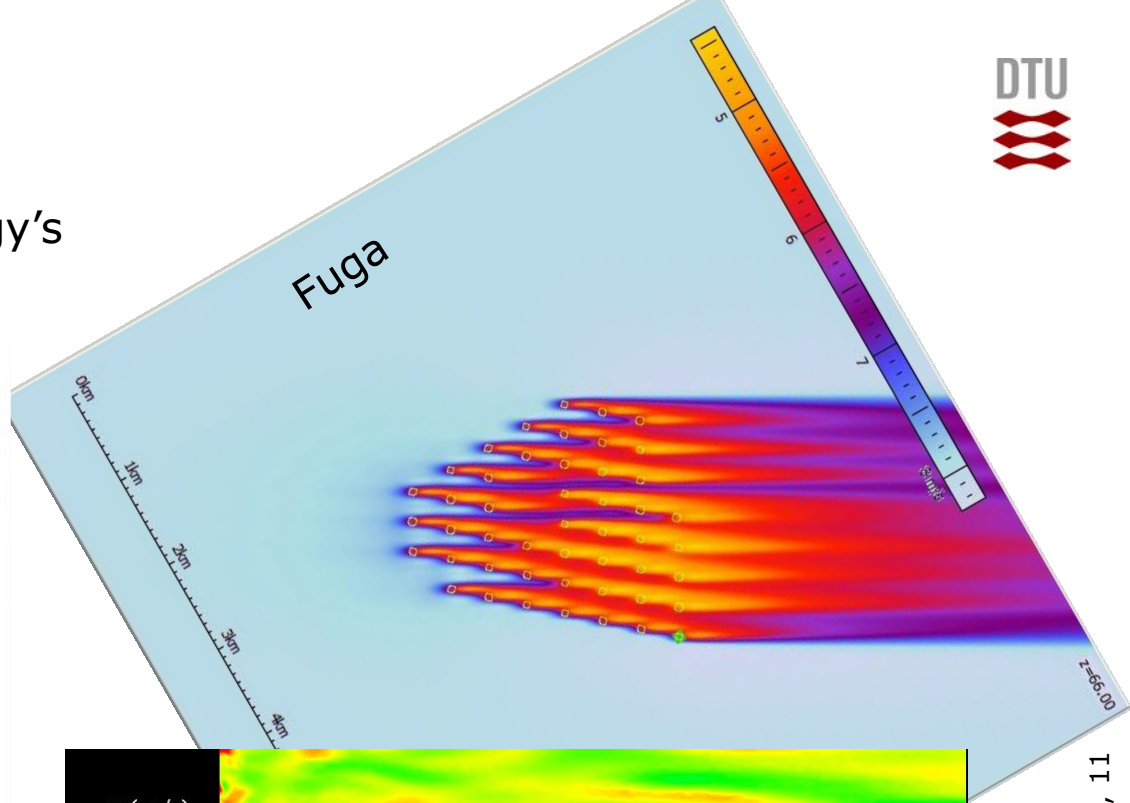
Larsén, Vincent & Larsen 2011



Courtney & Troen 1990

# Meandering

Measurements with SgurrEnergy's Galion Lidar.



Creech et al.(2013): High-resolution CFD modelling of Lilgrund Wind farm, RE&PQJ, 11

28 June 2013

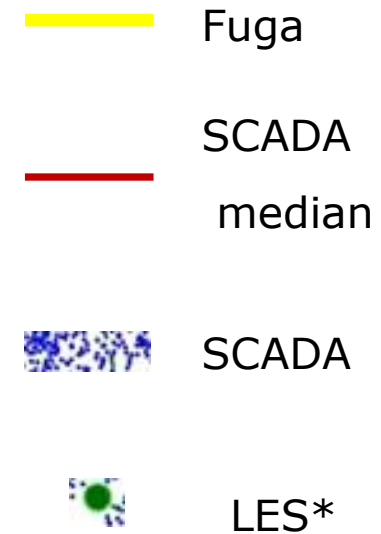
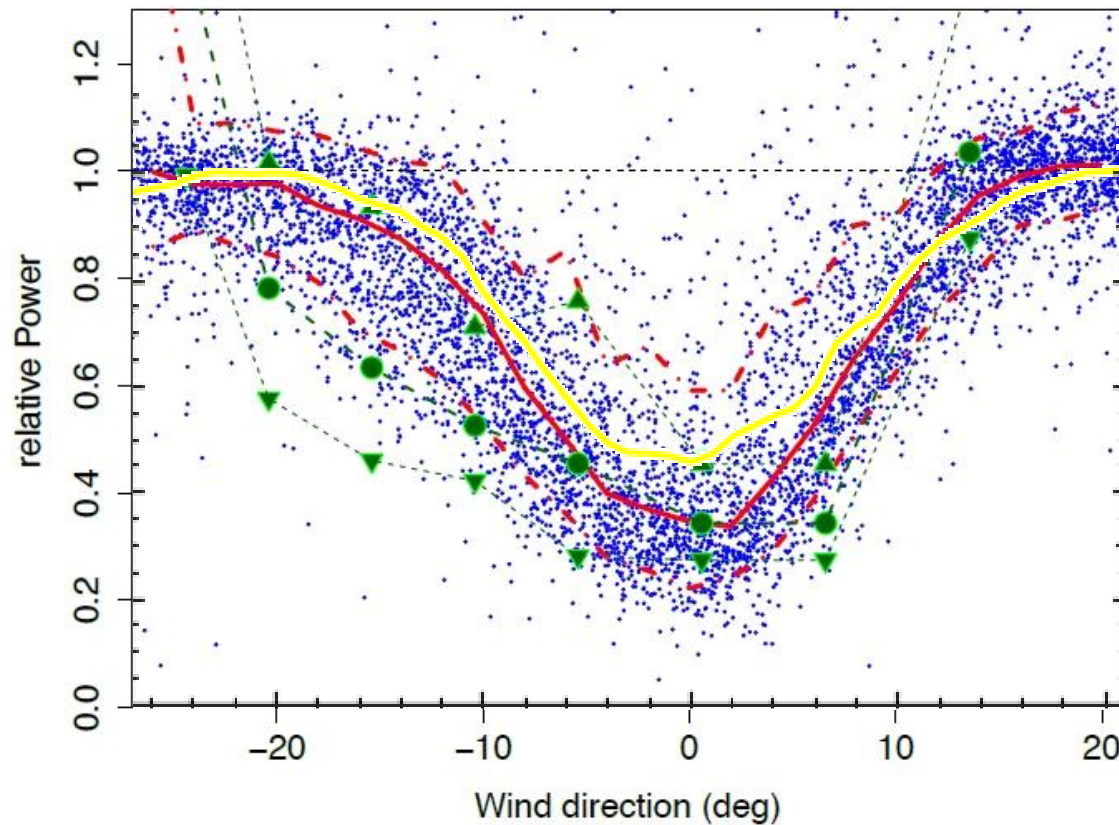
# Lillgrund wind farm



Photo by Kurt's wife



# D7 being shadowed by D8



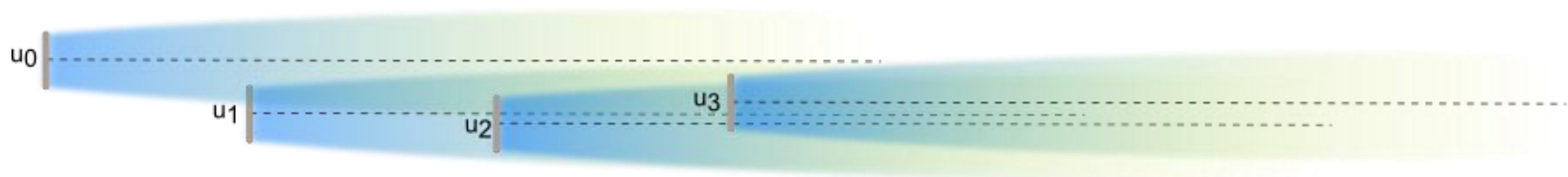
CPU time:  
 Fuga: < 1 minute  
 LES: 105000 hours

\*Creech et al.(2013): *High-resolution CFD modelling of Lillgrund Wind farm*, RE&PQJ, 11

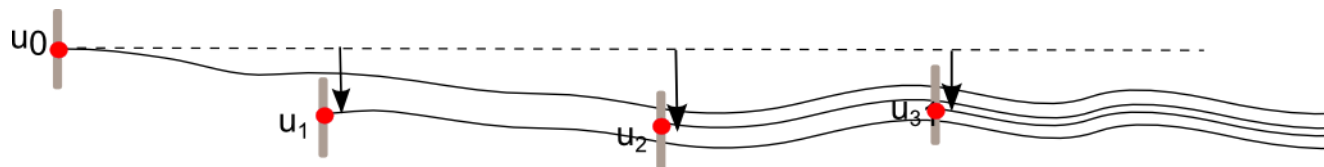


# Stochastic meandering

The model says:



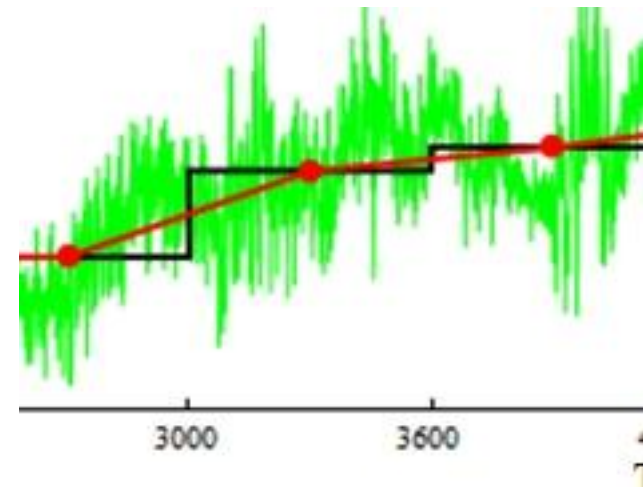
But something like this is more realistic:



# Meandering model

- 1) Make spectra of the difference between the green and the red curves.
- 2) Fit to the Mann model spectra\*
- 3) Make tracer particle diffusion in Mann turbulence field using the spectra
- 4) Compute Lagrangian velocity auto-correlation function
- 5) Fit a model to it

Simulating  $U-U$

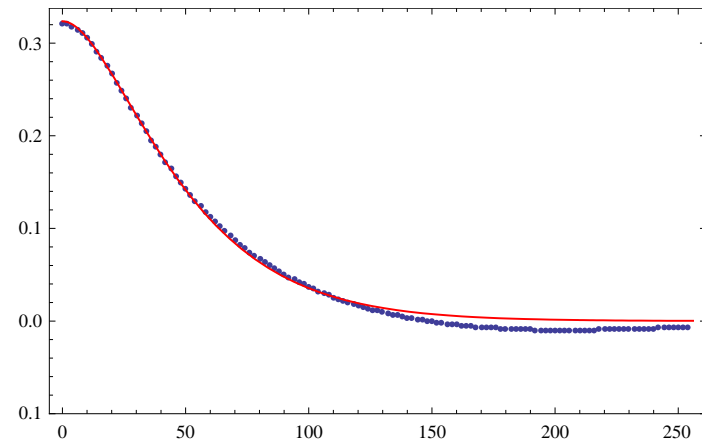


\* Peña, Gryning & Mann (2010): On the length-scale of the wind profile, *Q. J. R. Meteorol. Soc.* 136: 2119–2131

# A model for turbulent diffusion in Mann turbulence

Fitting the Lagrangian velocity auto-correlation function:

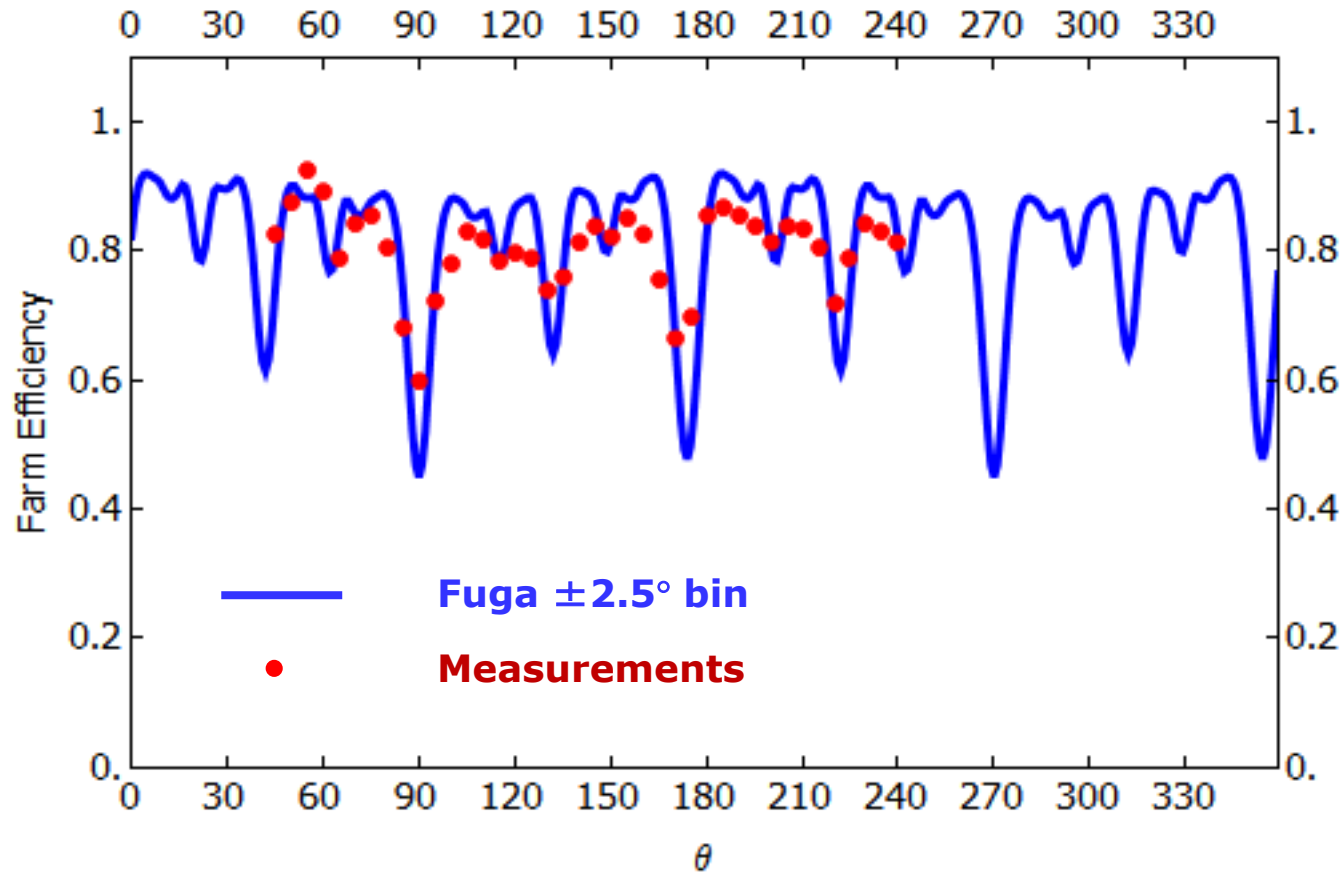
$$\langle v(t+\tau) v(t) \rangle = \sigma^2 (1 + \lambda \tau) e^{-\lambda \tau}$$



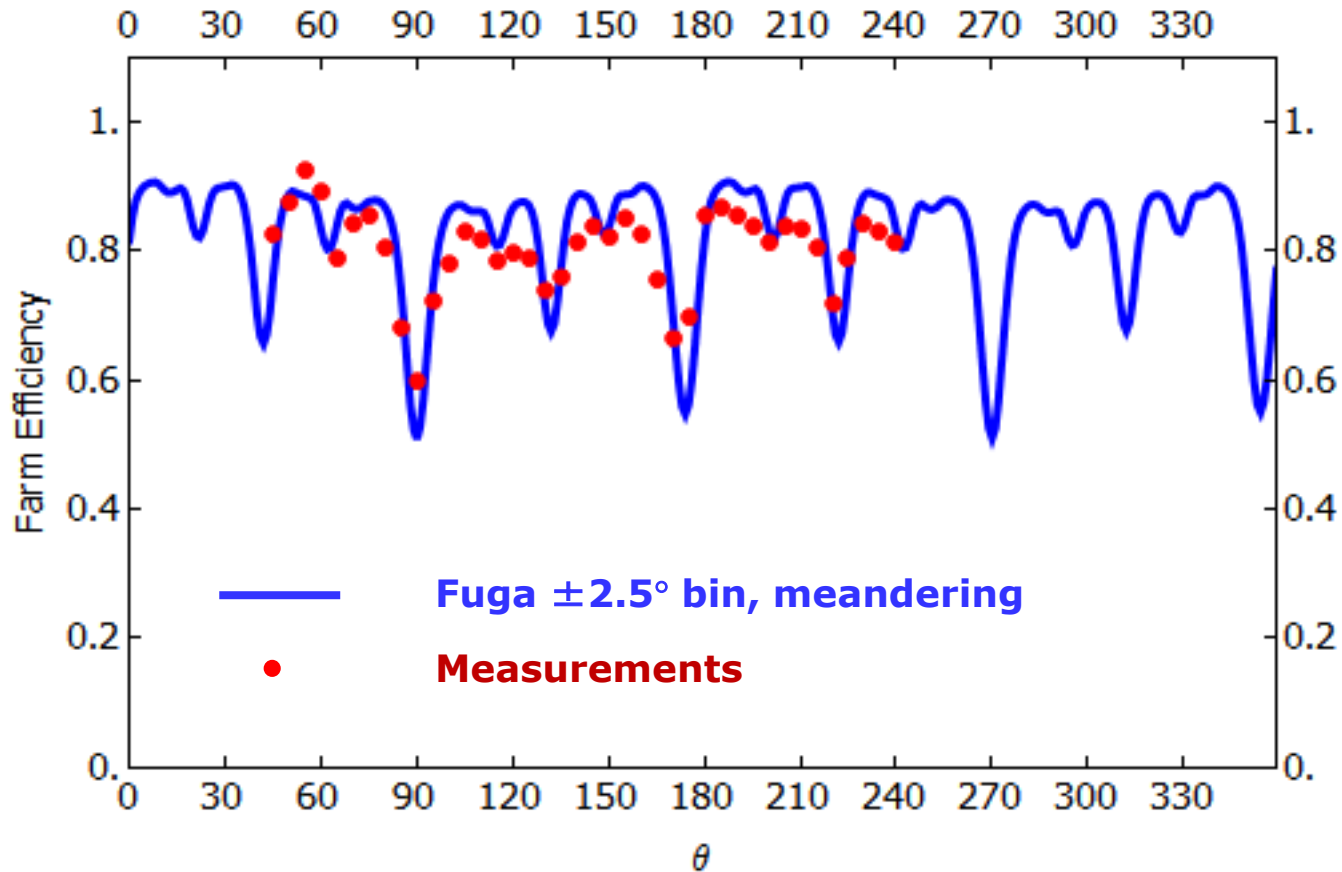
This model says: the acceleration is an Ornstein-Uhlenbeck process.

Analytical solution in terms of stochastic numbers.

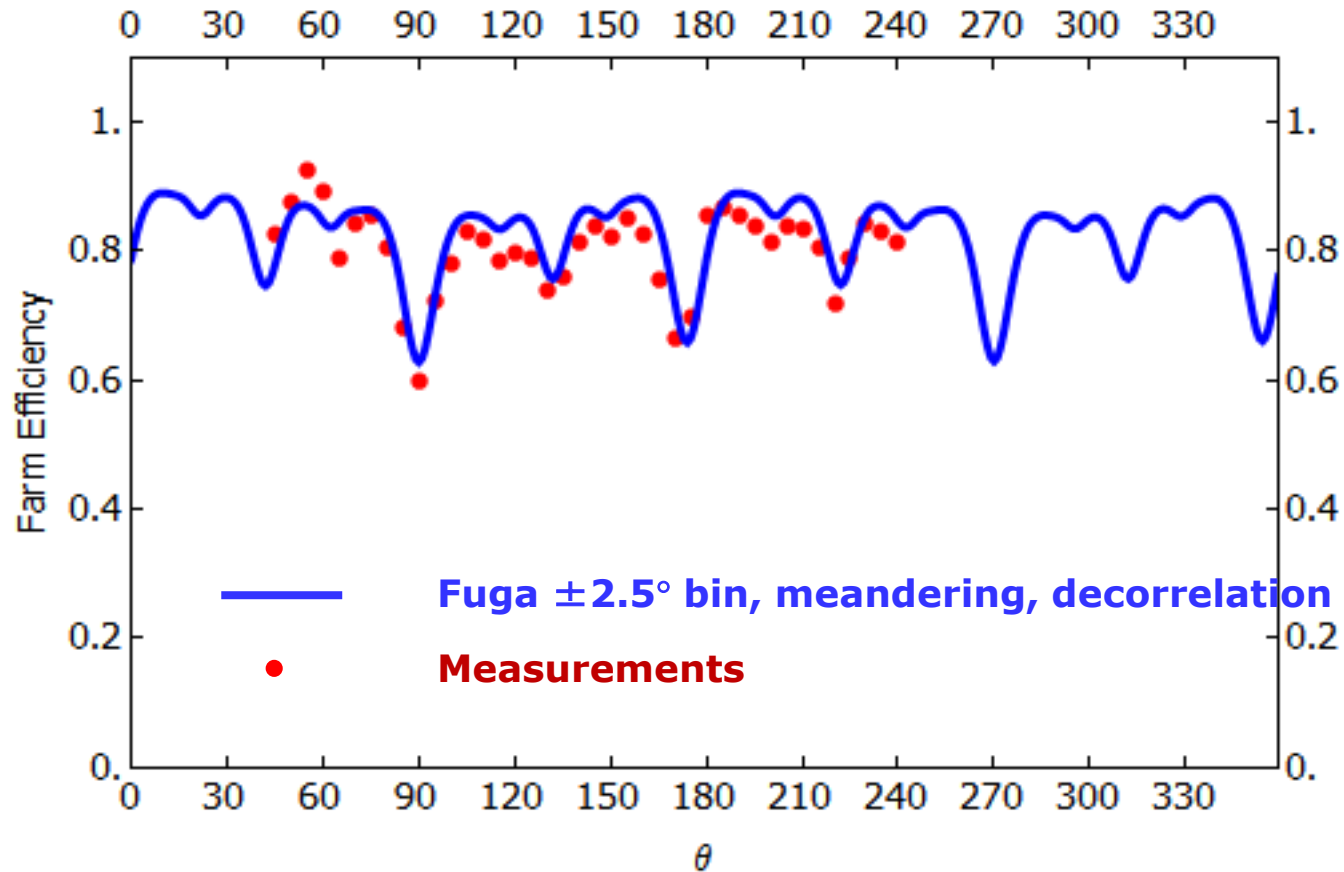
# Validation – Horns Rev 1



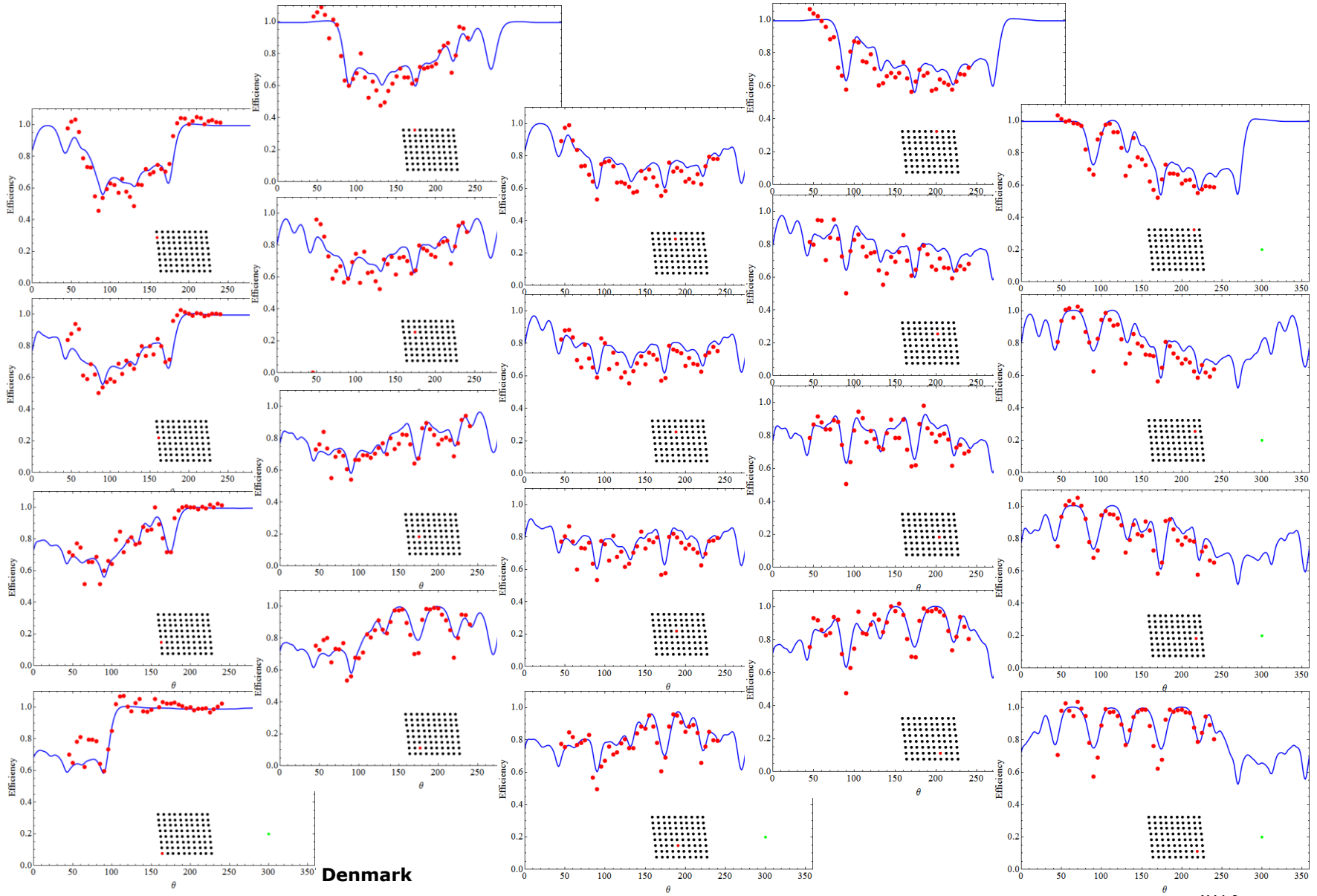
# Validation – Horns Rev 1



# Validation – Horns Rev 1



# Horns Rev 1. Efficiency for individual turbines



# Conclusions

- The interpretation of met data can have a profound impact on the results of validation exercises for offshore turbine wake models.
- The interpretation of met data can and should be backed up by on-site observations.
- Data for narrow bins around down-the-rows wind directions are the most uncertain.
- Sampling in very small wind direction bins introduces a large uncertainty because we don't really know what 'true' wind direction we should feed the models with.
- We need to know more about spatial and temporal wind field correlations across wind farms. The Doppler lidar is the key instrument.
- Fuga is fast and yields good results. It is a useful tool.