

Offshore Wind R&D 2015

Adwen

AN AREVA GAMESA COMPANY

Long-term experience at alpha ventus – Model and measurement based life time estimation

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Introduction

Adwen – an AREVA GAMESA company

Adwen has 126 wind turbines in operation in the German North Sea

- 6 alpha ventus
- 40 Trianel Wind Farm Borkum
- 80 Global Tech I

19.4% market share during the first semester of 2015 (according to EWEA statistics)

Trianel Wind Farm Borkum (Source: http://www.adwenoffshore.com)





Overview

"Long-term experience at alpha ventus" – Model and measurement based life time estimation –

Topics:

- Lumping approach an innovative consideration of wind and wave data for an integral load analysis
- Dynamic Wake Meandering (DWM) Model Validation of fatigue loads based on long-term measurements at alpha ventus
- State Observer automatized life time estimation of support structures based on measured loading

This work is related to Adwen's current RAVE projects

- OWEA Loads
- GIGAWIND*life*





An innovative consideration of wind and wave data for an integral load analysis



Classical Analysis and New Approach

- Scatter matrices of hub wind speed (V_{hub}), significant wave height (H_s) and peak period (T_p) data form the basis for the classical analysis.
- From data correlation functions H_s (V_{hub}) and T_p (H_s) are derived
- <u>Classical fatigue load cases</u> defined using 30° WWM steps are too conservative for monopile structures
 30° WWM
 30° 30°

➔ More realistic representation is preferable

- New <u>Lumping approach</u> was developed using 29 years (1979 2007) DHI model data representative of German Bight
- Specific consideration: Eigenfrequency of structure critical T_p
- Wind-wave misalignment (WWM) is explicitly considered

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Exemplarily wind-wave misalignment (WWM) rose



Analysis Method

- Classification of (V_{hub}, WWM) for all events according to (T_p, H_s)
 → Basis of a representative load case set
- Critical: Peak spectral periods close to 1st eigenfrequency of the structure → higher resolution in critical range of T_p
 - For T_p > critical period: 4 sub-groups per (V_{hub}, WWM) combination specified
 - For T_p < critical period: 9 sub-groups per (V_{hub}, WWM) combination specified
- Results in ~1500 fatigue load cases for DLC1.1 / 6.4 acc. to GL 2012 (compared to 1300 in classic approach), but much more realistic
- Probability of occurrence for each load case assigned

➔ Accepted by DNV GL for definition of load case groups DLC1.1 and 6.4



Lumped (H_s , T_p) data compared to classical scatter data







Tower bottom side-side DELs: Classic vs. Lumping







Dynamic Wake Meandering Model

Validation of fatigue loads based on long-term measurements at alpha ventus

In cooperation with B. Schmidt¹, S. Lott²

¹ 8.2 Consulting AG ² Stuttgart Wind Energy (SWE), University of Stuttgart, Germany

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The Dynamic Wake Meander (DWM) model consists of, primary:



1) Velocity deficit: Extraction of kinetic energy reduces wind speed

- 2) Meandering: Wind speed deficit moves in space
- 3) Added turbulence: Blades and hub create vortices that add turbulence





Alpha Ventus park layout with wake distances









<u>Note:</u> Presented loads are **not design loads**, only valid for this comparison!

Blade Load validation with Bladed 4.3 DWM model





DWM model has also been validated for tower bottom fore-aft moment.



State Observer

automatized life time estimation of support structures based on measured loading



State Observer

Concept for monitoring of fatigue loads

- Simulated fatigue loads of OWECs may contain several uncertainties
 > significant potential expected by knowledge of real fatigue loads
- Therefore measured <u>fatigue loads</u> and real <u>lifetime estimation</u> needs continuous evaluation of strains at critical points
- Not trivial to calculate deflections / strains from accelerometers due to summation of sensor errors.
 → Model based Kalman Filter yields optimal estimates
- Benefits of Kalman Filter:
 - Considering all available (mixed) sensor information to improve accuracy
 - Realtime state estimate (deflection, forces) for advanced control
 - Realtime Sensor Integrity Monitoring





State Observer (Kalman Filter)



State Observer



Result: Strain time series from estimated state $\hat{s}(t) = T\hat{x}(t)$

- Lifetime estimation requires only amplitudes as long as mean stress values are small with respect to the measured amplitude



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State Observer

Perspective



With a sufficient state estimation, the accumulated damage can be calculated:



- Fatigue assessment possible for all points of the tower depending on the used model
- Remaining life time estimation





Summary

Outlook



Summary / Outlook

Long-term experience at alpha ventus

Lumping approach

- More realistic approach of applying environmental input for fatigue load calculation – accepted by DNV GL
- Benefit shown for monopile substructures

Dynamic Wake Meandering (DWM) model

- Wake effects shown from OWECs with more than 10D distance
- Implementation of DWM in Bladed and validation with RAVE data
- → further improvement using more than 4 years of AV measurement
- State observer for fatigue monitoring
- Realtime model + real measurements = real fatigue loads
- Functionality shown for simple model
- → Advancement planned for real-scaled structures, Validation @ AV



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Load Prediction

Load Validation

Load Simulation

Thank you !

