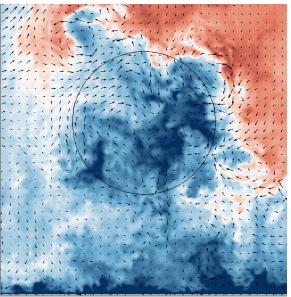


Sandia Wind Energy

SAND2017-12647PE

Uncertainty Quantification Applications



David Maniaci, Thomas Herges, Chris
Kelley, and Brandon Ennis

Wind Energy Technologies Department
Sandia National Laboratories



SAND2017- C

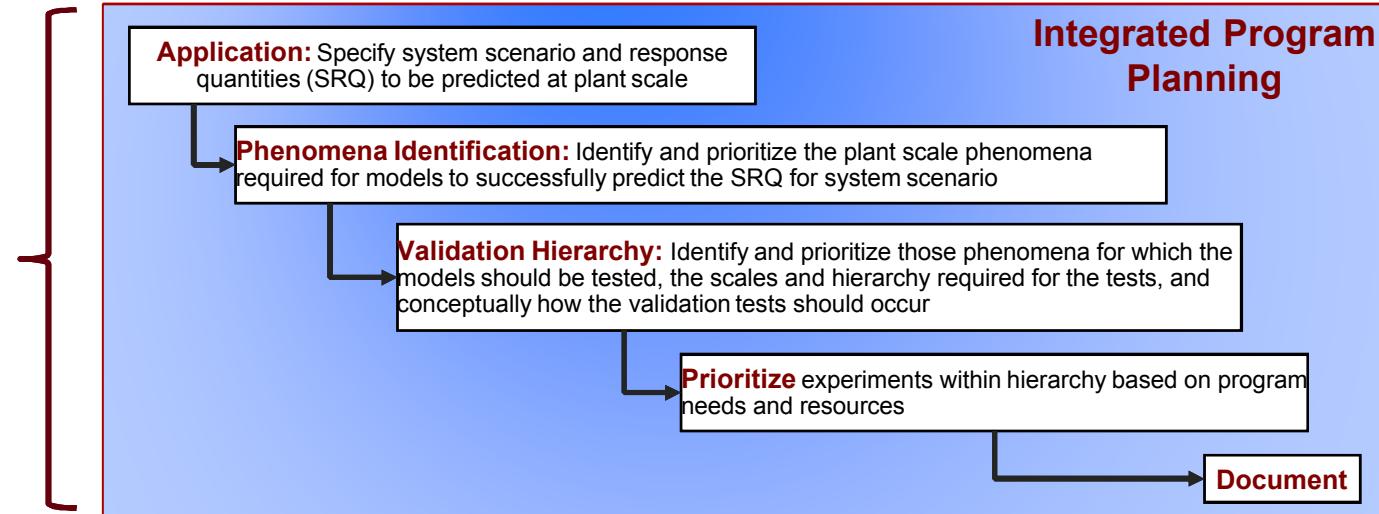
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V&V Framework

(2015 Hills, Maniaci, Naughton)

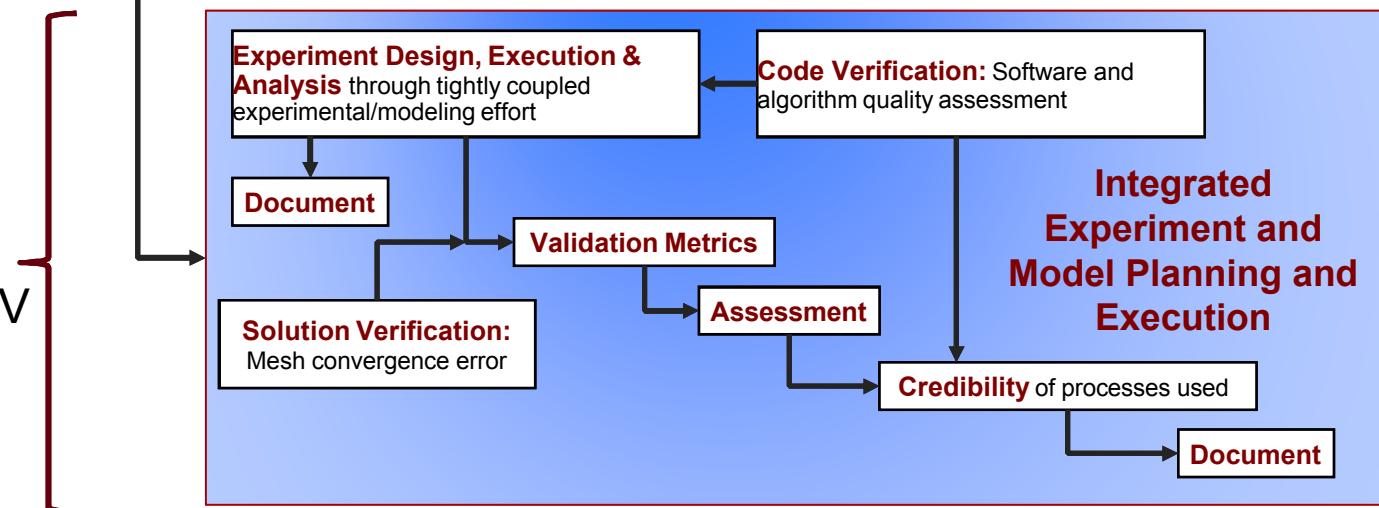
Integrated Planning

- Program leaders, modelers, software developers, experimentalists, V&V specialists

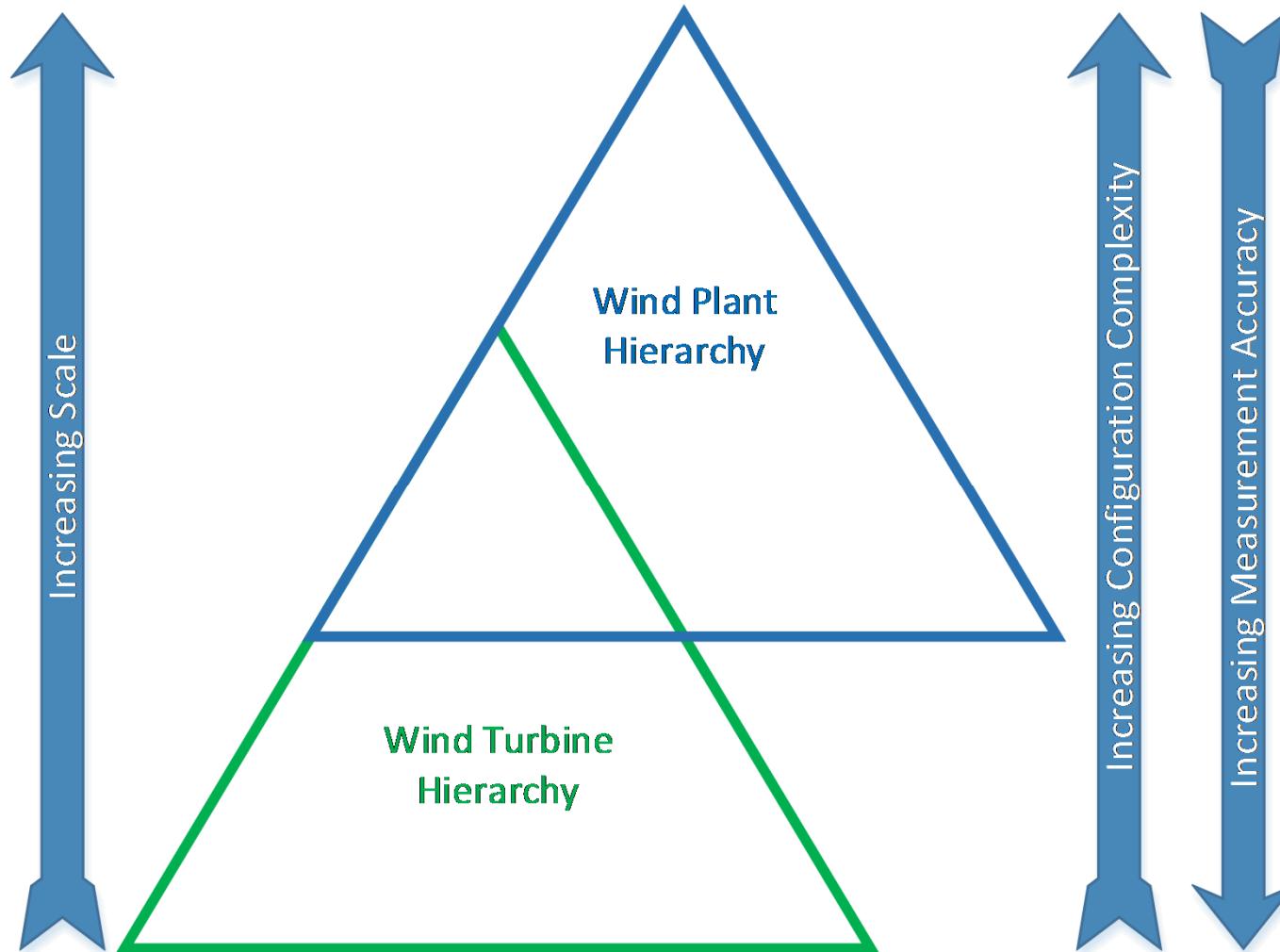


Validation Planning

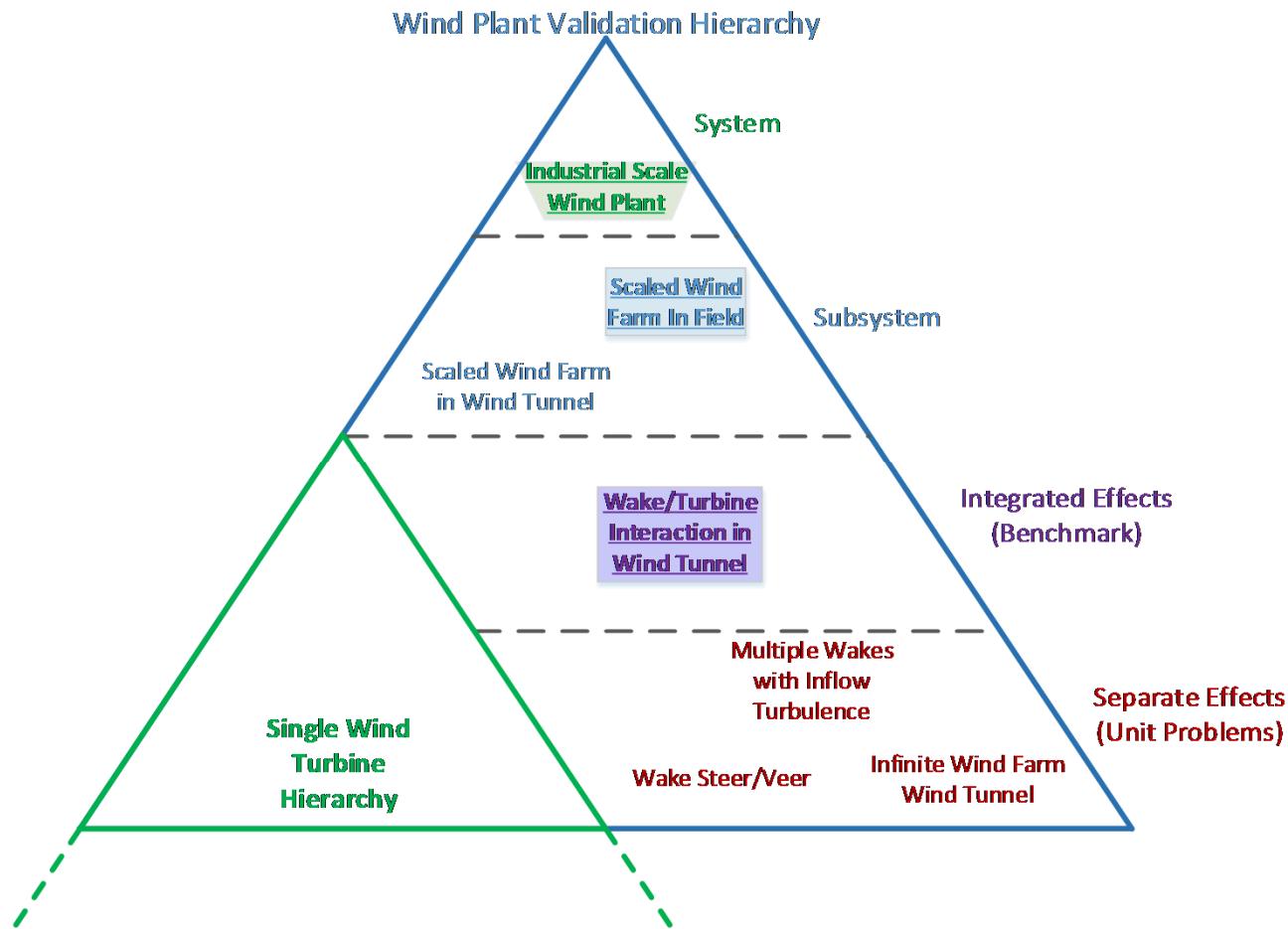
- Domain specific program leaders, modelers, experimentalists, V&V specialists, data acquisition specialists



Validation Hierarchy

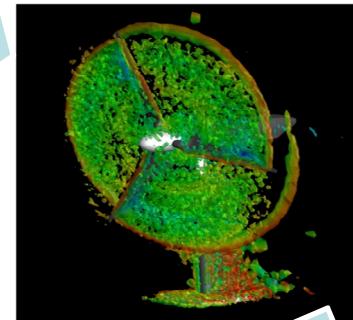


Wind Plant Validation Hierarchy

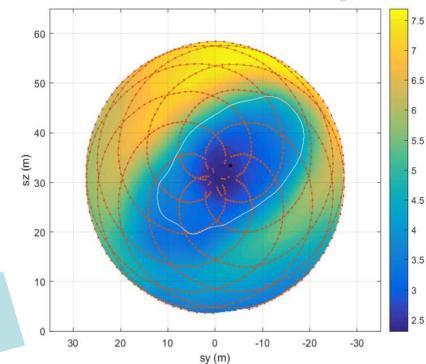
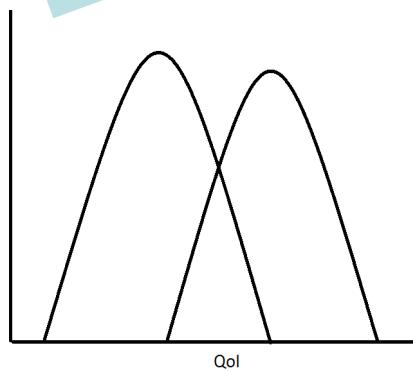


Slide 4

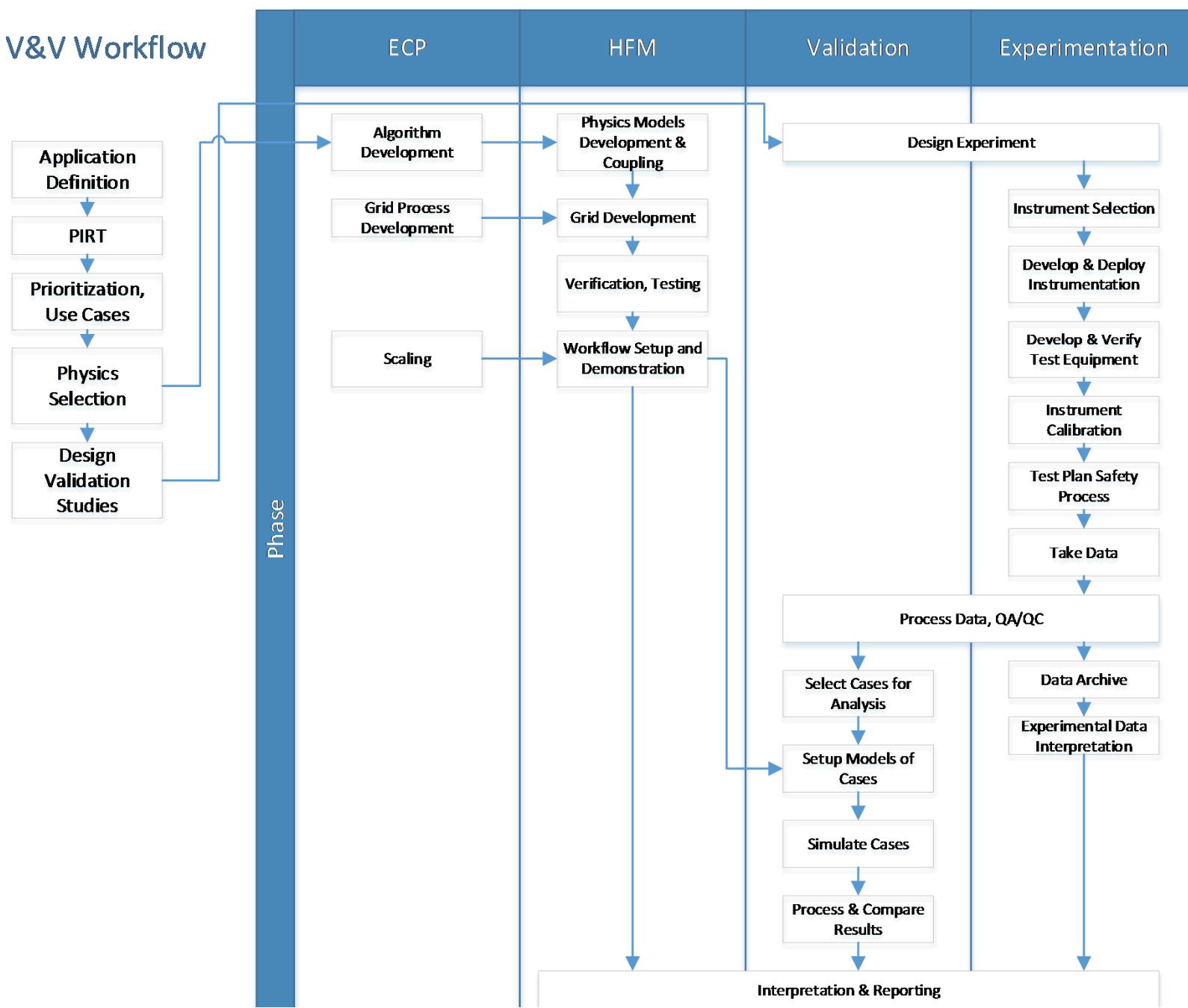
- MDC4** SWiFT is one assessment within many that will be utilized to build confidence in Nalu's predictive capability.
Maniaci, David Charles, 4/19/2017



Virtuous Cycle
Validation
Model Development
Experimentation
Uncertainty Quantification

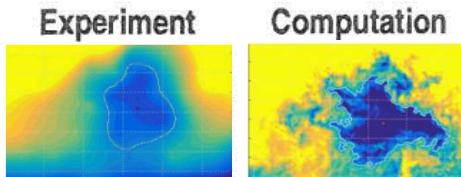


V&V Workflow

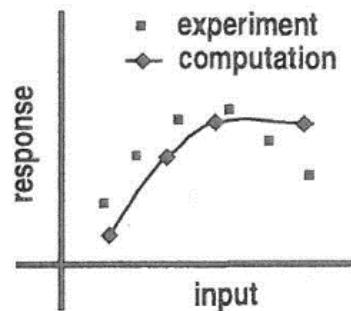


Uncertainty Quantification

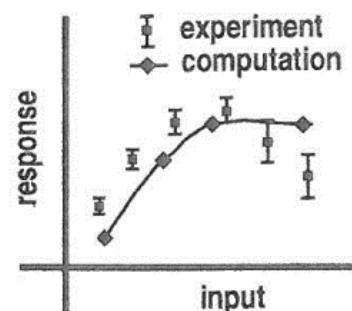
Levels of Precision



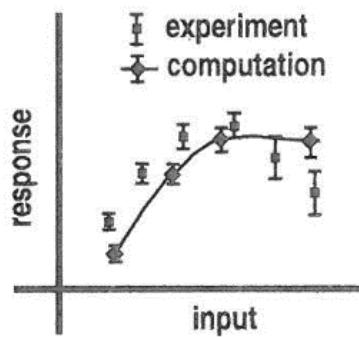
(a) Viewgraph Norm



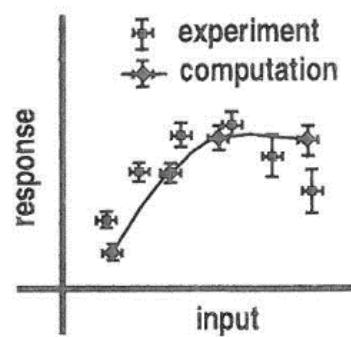
(b) Deterministic Simulation



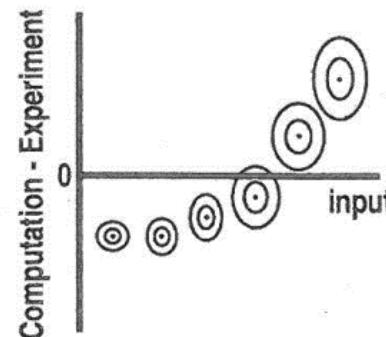
(c) Including Experimental Uncertainty



(d) Including Numerical Error



(e) Nondeterministic Simulation



(f) Statistical Mismatch

Slide 7

MDC2

Maniaci, David Charles, 4/17/2017

Projects

- SWiFT Site Experimental Uncertainty Quantification
 - Inflow, Turbine Loads and Acceleration, Wake Measurement
- DTU Spinner Lidar Wake Tracking Uncertainty
- Multi-level Uncertainty Quantification with LES
- V&V Framework and application to wind energy
- Wakebench (IEA Task 31): VV&UQ Framework and User Guidelines
 - 1. Validation framework for wind energy applications
 - 2. Uncertainty quantification procedures using subscale wake testing
 - 3. Uncertainty quantification procedures using lidar experiments
 - 4. Framework for wind plant uncertainty quantification

SWIFT Site Experimental Uncertainty Quantification

- Inflow Measurements

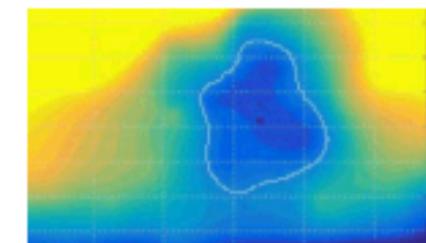
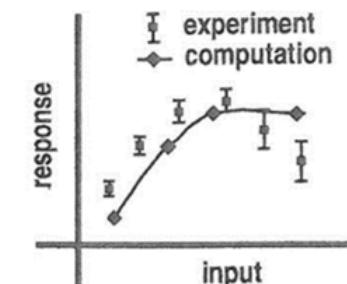
- α
- ρ
- p
- RH
- T
- TI (sonic)
- U_{∞} (sonic)
- U (sonic)
- V (sonic)
- U_{∞} (cup)
- U_{∞} (nacelle)
- V_r (sonic)
- WD (sonic)
- WD (vane)

- Turbine Measurements

- Aerodynamic power
- Rotor speed
- Aerodynamic torque
- Rotor thrust
- Individual blade root loads (instantaneous)
- Individual blade loading profile (instantaneous)
- Nacelle measured wind direction
- Nacelle measured wind speed
- Yaw heading
- Yaw misalignment
- Blade pitch
- Rotor azimuth
- Nacelle acceleration

- Wake Measurements

- DTU Spinner Lidar
- Wake identification and tracking
- Turbulence estimators



SWIFT Met Tower Uncertainty



Table 3.1. SWIFT METEOROLOGICAL TOWER UNCERTAINTY SUMMARY

Measurand	units	δ	$\delta/\text{mean} (\%)$	mean
α	—	1.88×10^{-3}	0.89	0.21
ρ	kg/m ³	7.9×10^{-4}	0.0731	1.08
p	Pa	1.53	1.71×10^{-3}	89500
RH	%	2.01	4.49	46.8
T	K	0.212	0.0733	289
TI (sonic)	—	1.23×10^{-3}	1.02	0.12
U_∞ (sonic)	m/s	0.0439	0.207	6.82
U (sonic)	m/s	0.01	0.207	4.82
V (sonic)	m/s	0.01	0.207	4.82
U_∞ (cup)	m/s	0.20	2.93	6.82
U_∞ (nacelle)	m/s	0.50	7.33	6.82
V_r (sonic)	°	1.73	86	2.0
WD (sonic)	°	1.22	0.69	176
WD (vane)	°	1.20	0.68	176

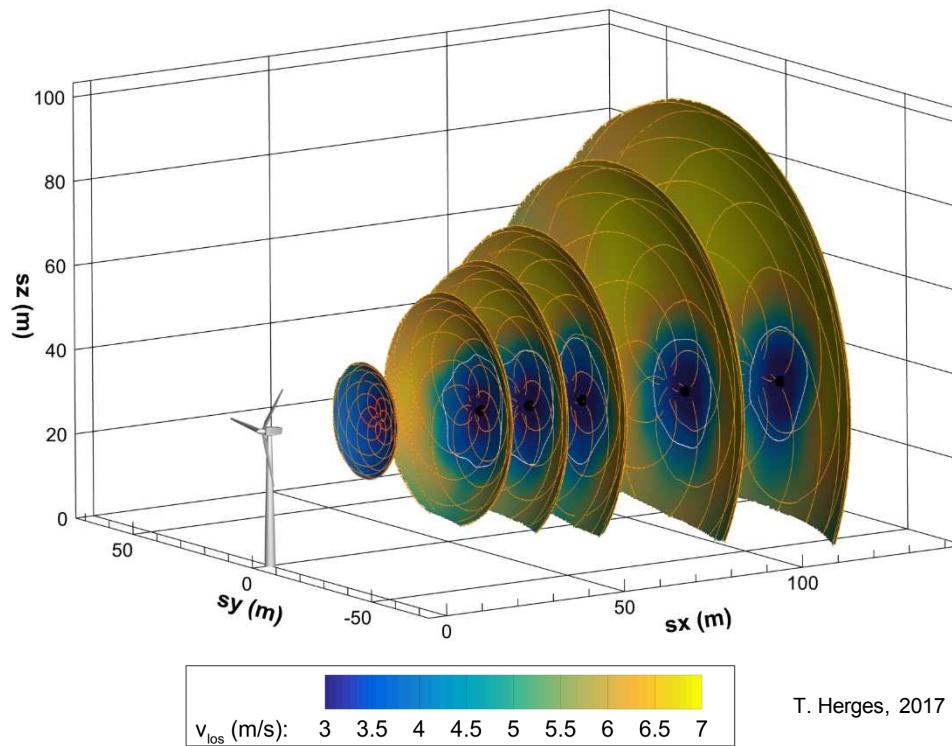
Five inflow quantities have uncertainties above 1% of their mean values: veer (86%), nacelle wind speed (7.3%), relative humidity (4.5%), and cup anemometer wind speed (2.9%).

The nacelle wind speed, humidity, and cup anemometer uncertainty are all dominated by low manufacturer specified accuracies. Therefore all of these uncertainties could be reduced by purchasing higher accuracy sensors.

SWiFT Wake Measurements

DOE/SNL Scaled Wind Farm Technology (SWiFT) facility
hosted by Texas Tech University (TTU)

Objective: Assess the ability of models to predict ***wake shape, strength, and deflection.***



T. Herges, 2017



MDC11 Build from validation using a few datasets to looking at ensembles of data.

Maniaci, David Charles, 8/30/2017

Thank you

"If a man will begin with certainties, he shall end in doubts; but if he will be content to begin with doubts, he shall end in certainties."
- F. Bacon - 1605.



dcmania@sandia.gov

Multi-level Uncertainty Quantification with LES

Demonstrated an order of magnitude reduction in computational cost for a cylinder flow problem by coupling Nalu (CFD) to DAKOTA (UQ) and using a multilevel UQ approach.

Problem description

- (Laminar) Flow over a cylinder ($Re=10-750$)
- Input parameters: Density and Viscosity
- QoI: Coefficient of Drag
- 4 levels of mesh resolution

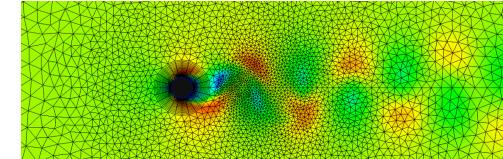
UQ approach

- Multilevel sampling-based estimator to accelerate convergence

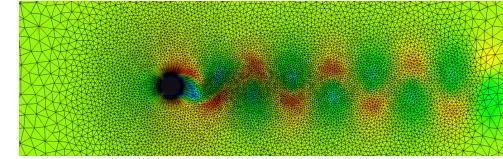
Impact

- Sampling methods are well suited for UQ problems with extremely high dimensionality (such as wind farm LES)
- Convergence is guaranteed for non smooth QoIs
- Demonstrated order of magnitude improvement in accuracy/cost of Multilevel estimators (MLMC) relative to conventional Monte Carlo (MC) for the cylinder problem

Coarse Mesh: 10 minute time to soln.



Medium Mesh: 4 hours time to soln.



Extrapolated Variance of the estimator

