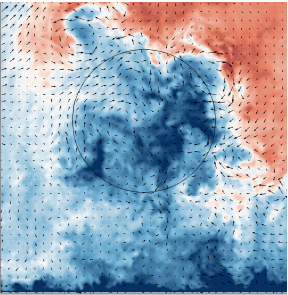


# Sandia Wind Energy SAND2017-12647PE Uncertainty Quantification Applications



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

Wind Energy Technologies Department  
Sandia National Laboratories



SAND2017- C

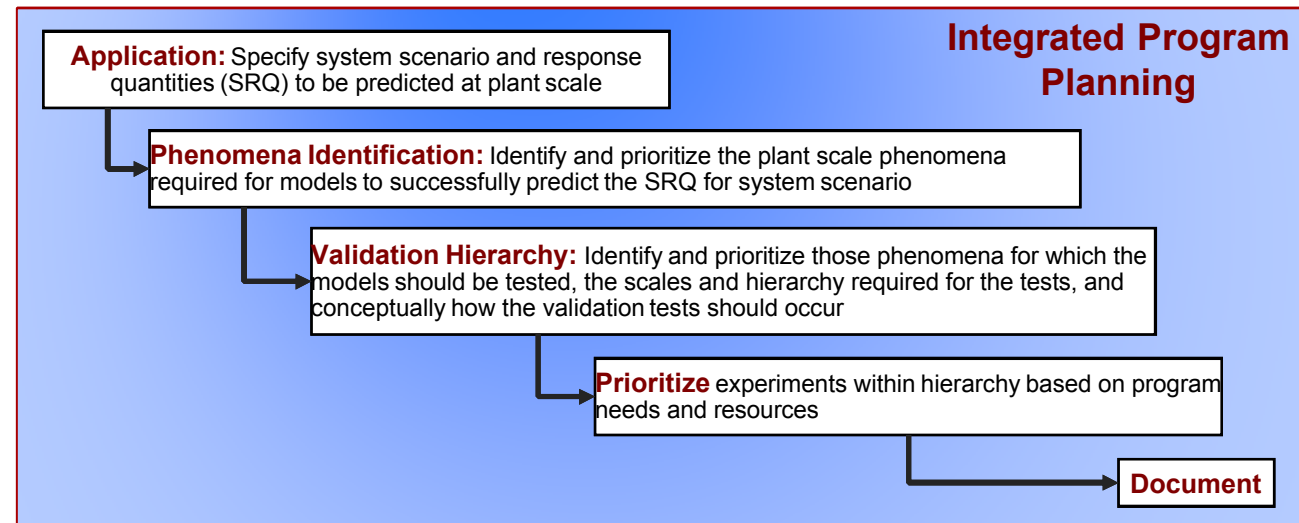
Sandia National Laboratories is a multimission laboratory managed and operated by National Technology and Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International, Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

# V&V Framework

(2015 Hills, Maniaci, Naughton)

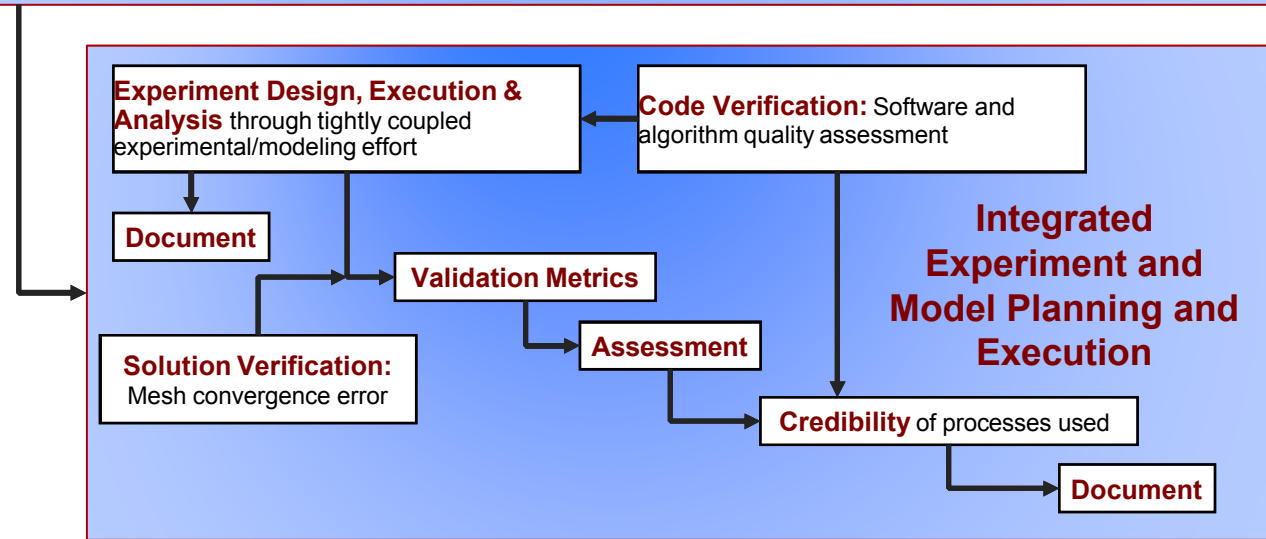
## Integrated Planning

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

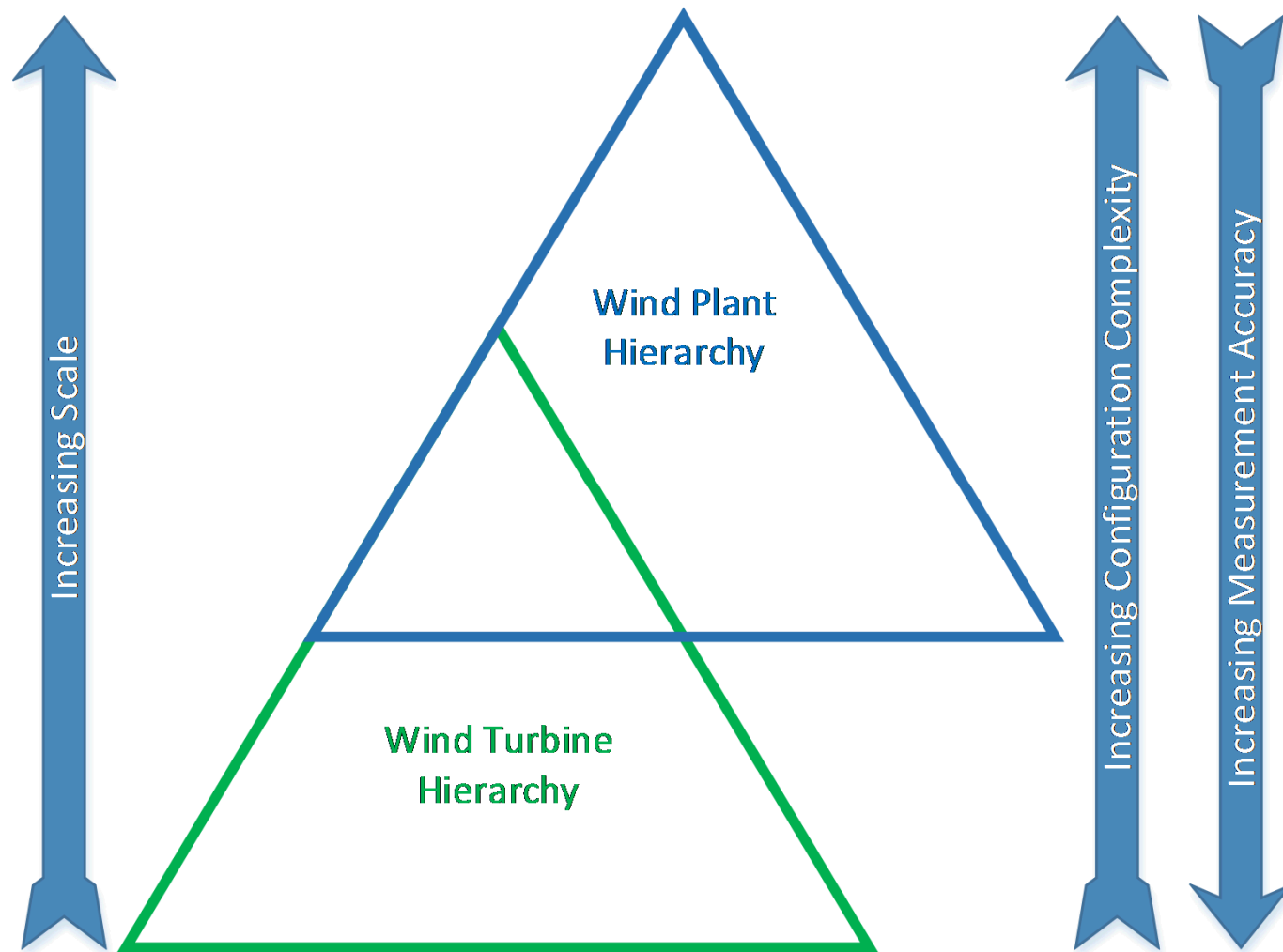


## Validation Planning

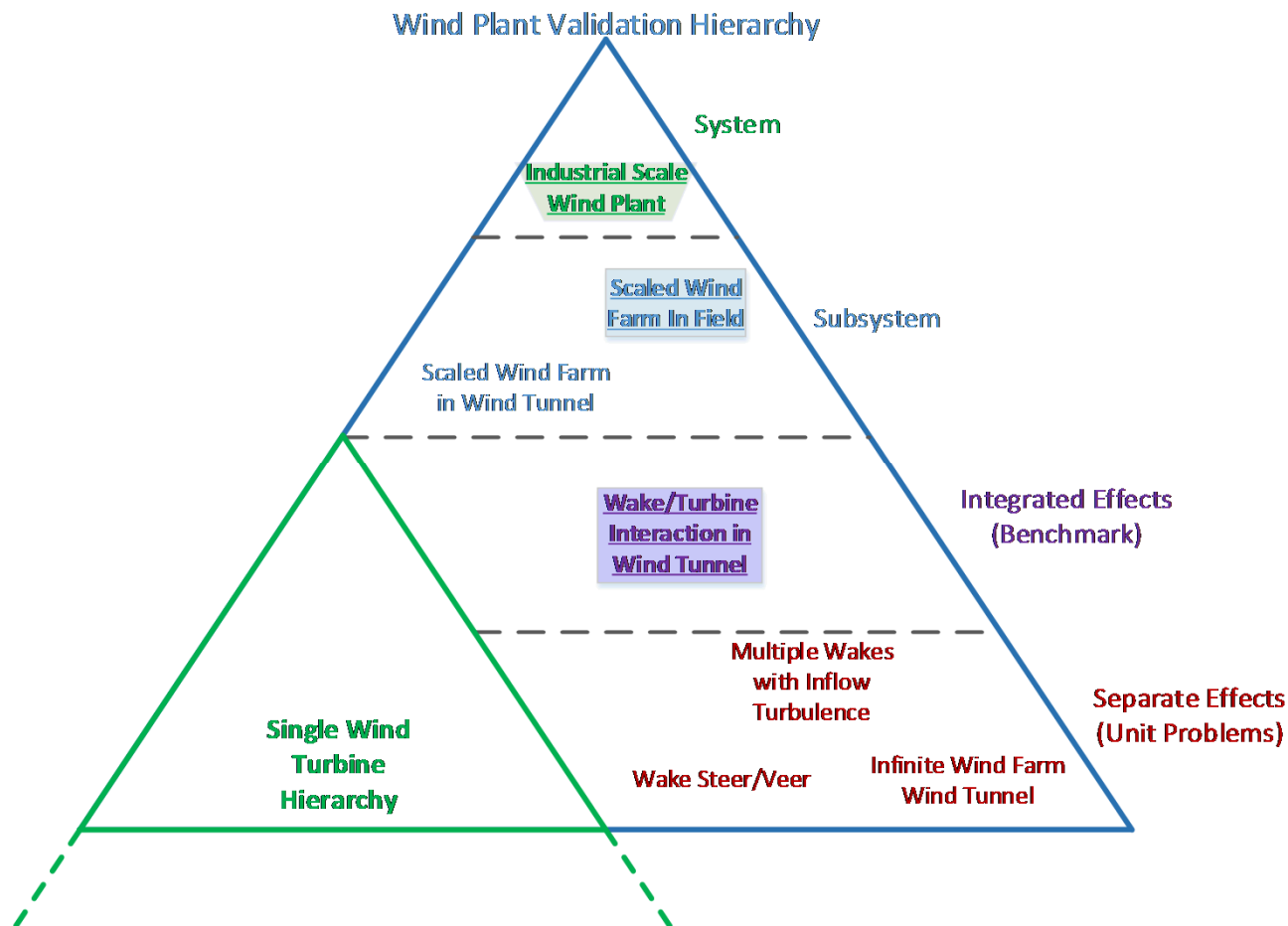
- Domain specific program leaders, modelers, experimentalists, V&V specialists, data acquisition specialists



# Validation Hierarchy



# Wind Plant Validation Hierarchy



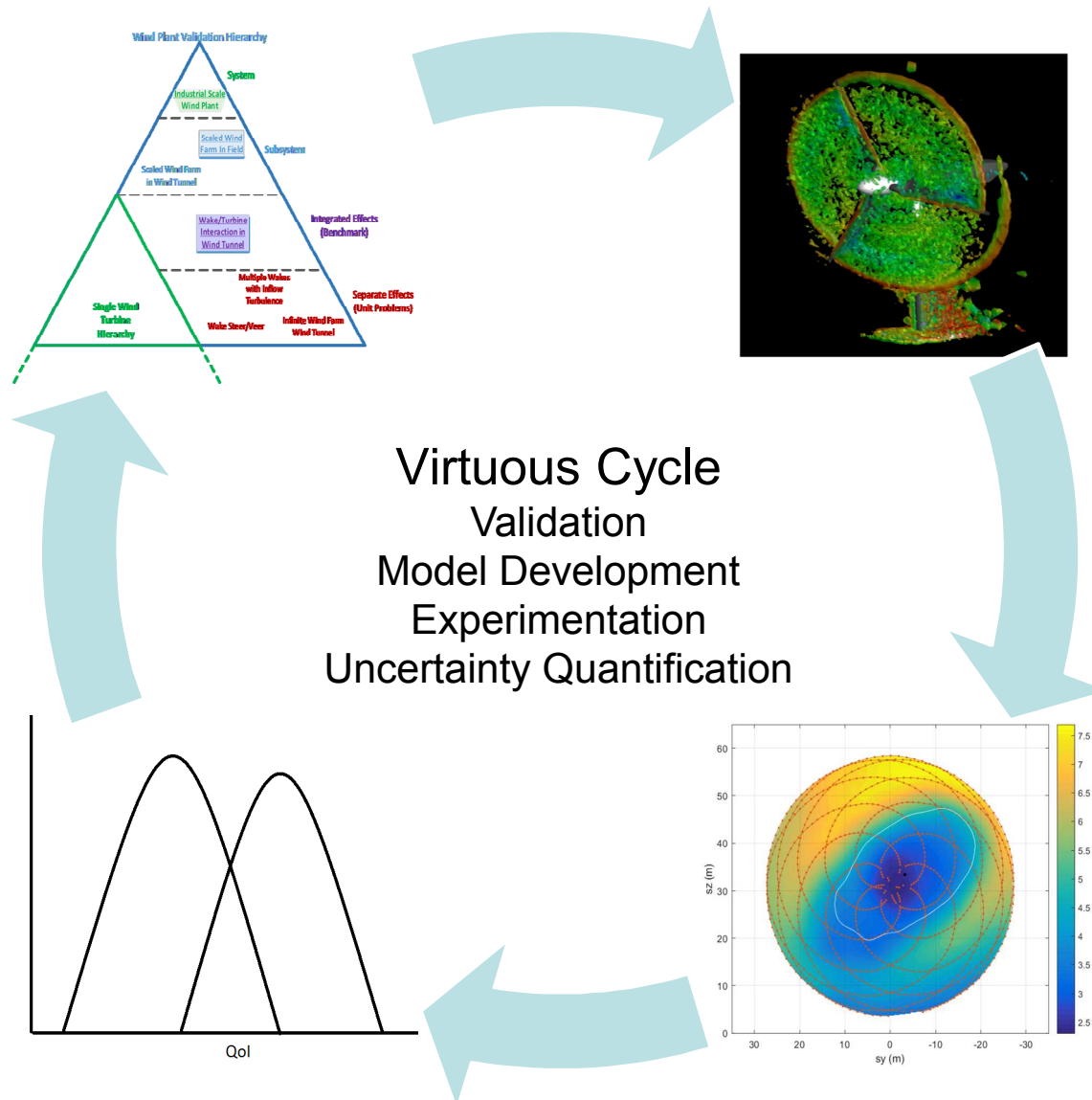
## Slide 4

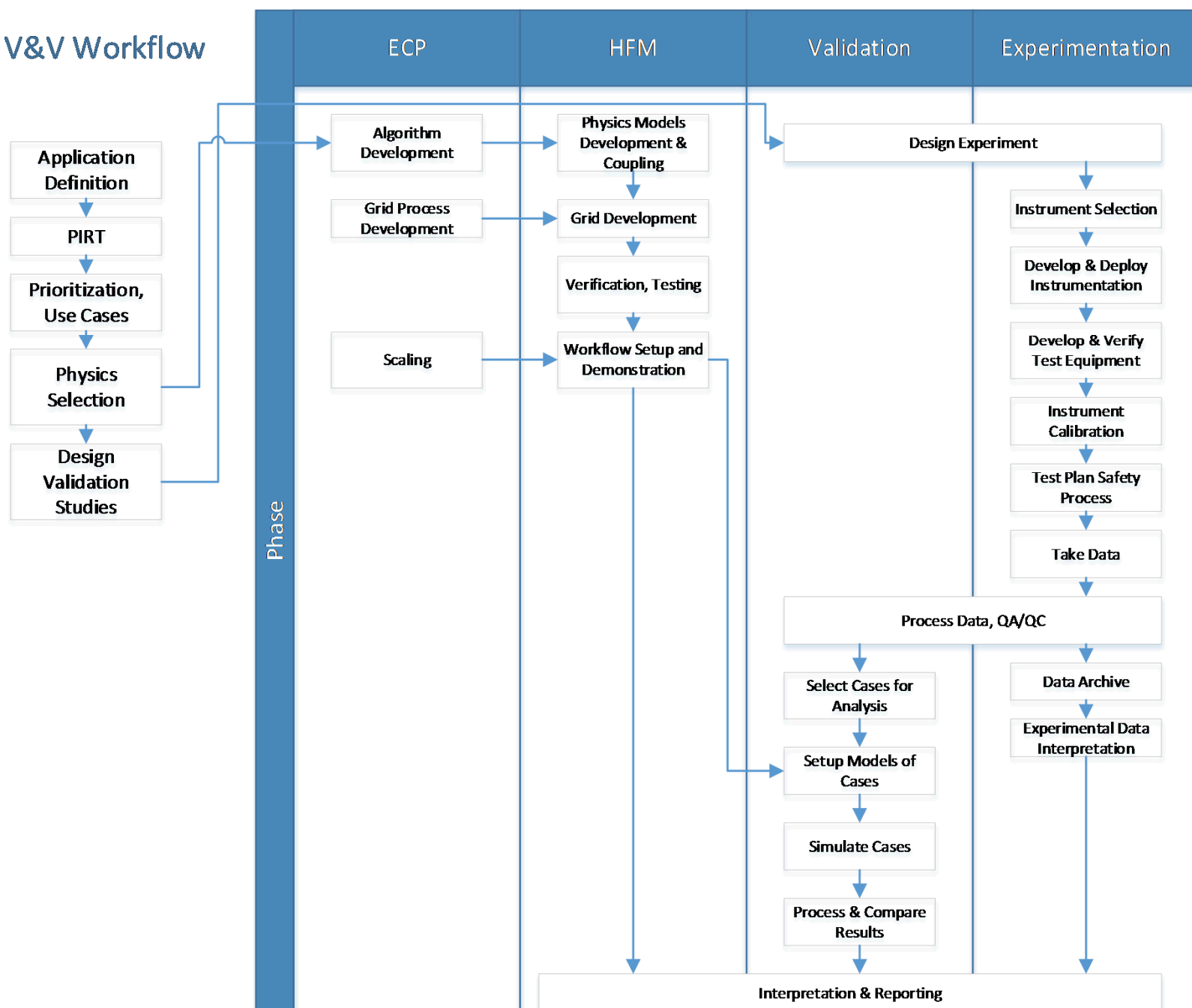
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### MDC4

SWiFT is one assesst within many that will be utilized to build confidence in Nalu's predictive capability.

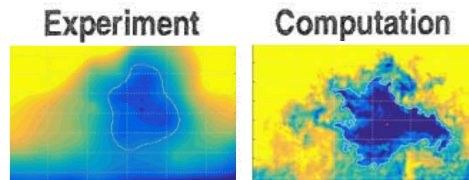
Maniaci, David Charles, 4/19/2017



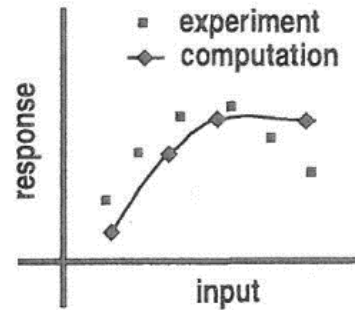


# Uncertainty Quantification

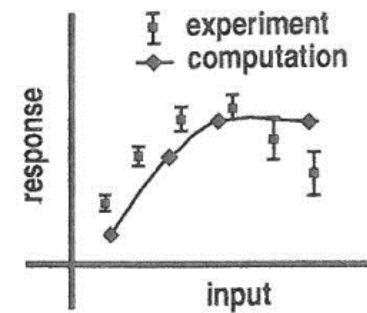
## Levels of Precision



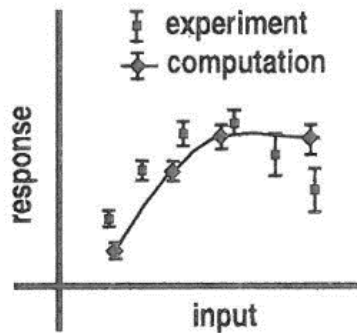
(a) Viewgraph Norm



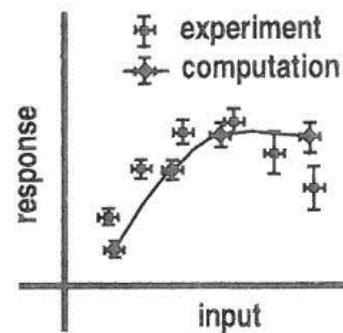
(b) Deterministic Simulation



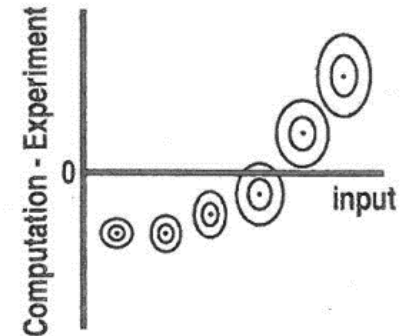
(c) Including  
Experimental  
Uncertainty



(d) Including  
Numerical Error



(e) Nondeterministic  
Simulation



(f) Statistical  
Mismatch

Modified from Oberkampf and Roy, 2012



## Slide 7

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**MDC2**

Maniaci, David Charles, 4/17/2017

- 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

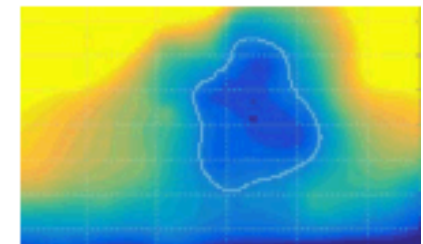
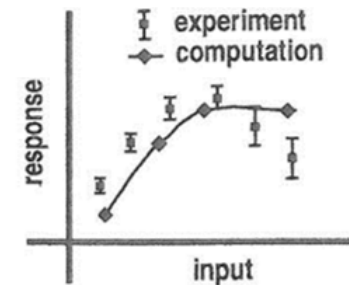
- $\alpha$
- $\rho$
- $p$
- RH
- T
- TI (sonic)
- $U_\infty$  (sonic)
- U (sonic)
- V (sonic)
- $U_\infty$  (cup)
- $U_\infty$  (nacelle)
- $V_r$  (sonic)
- WD (sonic)
- WD (vane)

## ■ 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	$\delta$	$\delta/\text{mean} (\%)$	mean
$\alpha$	—	$1.88 \times 10^{-3}$	0.89	0.21
$\rho$	kg/m <sup>3</sup>	$7.9 \times 10^{-4}$	0.0731	1.08
$p$	Pa	1.53	$1.71 \times 10^{-3}$	89500
$RH$	%	2.01	4.49	46.8
$T$	K	0.212	0.0733	289
$TI$ (sonic)	—	$1.23 \times 10^{-3}$	1.02	0.12
$U_{\infty}$ (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
$U_{\infty}$ (cup)	m/s	0.20	2.93	6.82
$U_{\infty}$ (nacelle)	m/s	0.50	7.33	6.82
$Vr$ (sonic)	°	1.73	86	2.0
$WD$ (sonic)	°	1.22	0.69	176
$WD$ (vane)	°	1.20	0.68	176

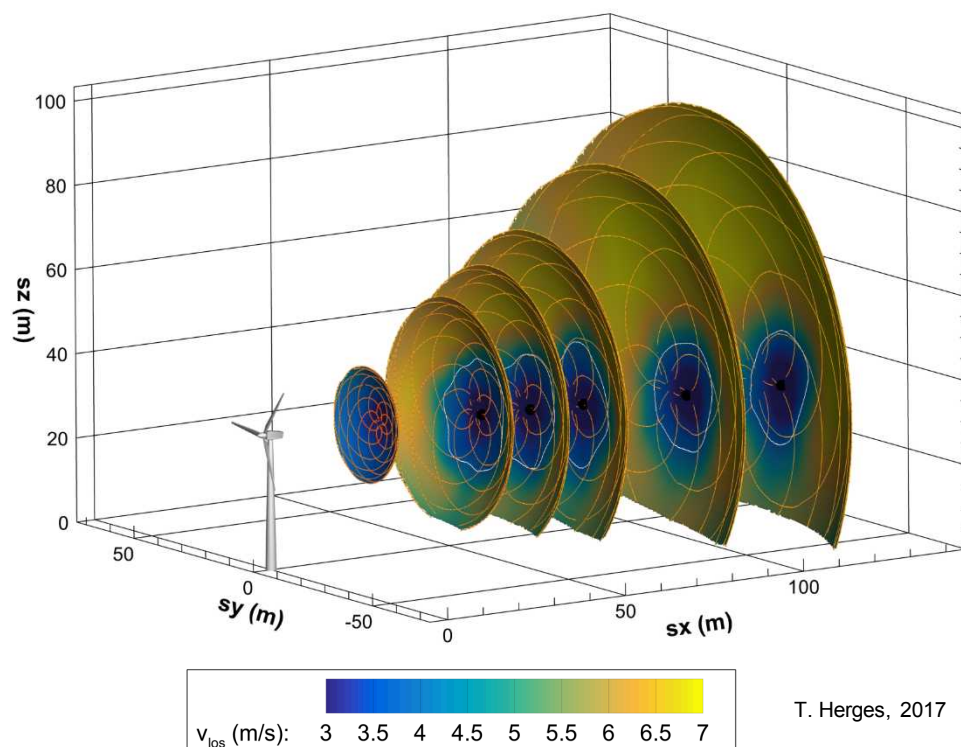
Five inflow quantities have uncertainties above 1% of their mean values: veer (86%), nacelle wind speed (7.3%), relative humidity (4.5%), and cup anemometer wind speed (2.9%).

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*.



## Slide 11

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**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.

dcmmania@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
- Qol: Coefficient of Drag
- 4 levels of mesh resolution

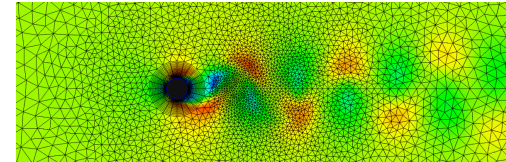
## UQ approach

- 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 Qols
- 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.

