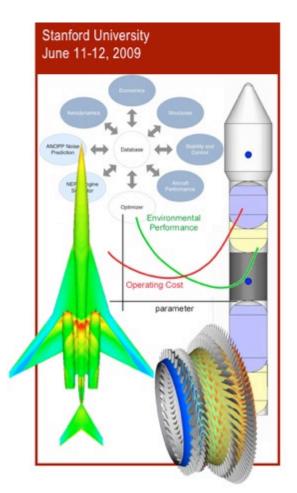
# Some Thoughts on Applicability of Aerospace Analysis and Design Techniques to Wind Energy

Wind Energy Systems Engineering Workshop NREL National Wind Technology Center Louisville, CO December 14, 201

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## **Brief Introduction to Our Work in Optimization**

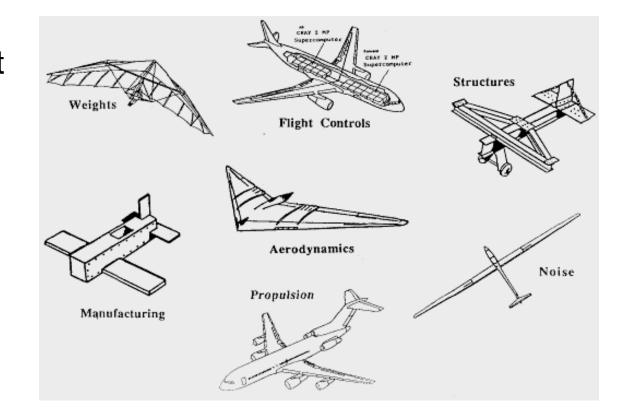
- Work in the department of Aero & Astro
- Main interests are in the application of high-fidelity simulation to the conceptual design of aerospace systems: aircraft, rockets, helicopters, engines (and by extension cars, trucks, UAVs, wind turbines, etc.)
- Significant difficulties/ challenges in many simulation and optimization topics



Annual Consortium meeting poster, 2009

## **Multidisciplinary Optimization (MDO)**

- What is really MDO?
  - A field of engineering/mathematics that allows the modification of an existing design, including multi-discipline interactions, to arrive at a better one
- Problem formulation is not obvious and requires engineering judgment.
- "One can only make one thing best at a time."



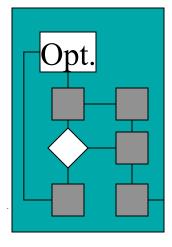
# **MDO in Aerospace Engineering**

- Arguably, aerospace engineering has been doing systems engineering and MDO for over 30 years. Are we done yet?
- We have gone through three generations of MDO tools in academia, but industry has only gotten (partially) to the second one.
- There have been many lessons learned and many useful techniques and tools developed. The key question is:

What can be learned from these experiences to help the development of Sys Eng techniques in wind power?

## **First Generation MDO**

- Integrated multidisciplinary
   analysis and optimization
  - Combine analysis/simulation with design method
  - Exploit advances in computational power to evaluate virtual prototypes, search design space, improve design



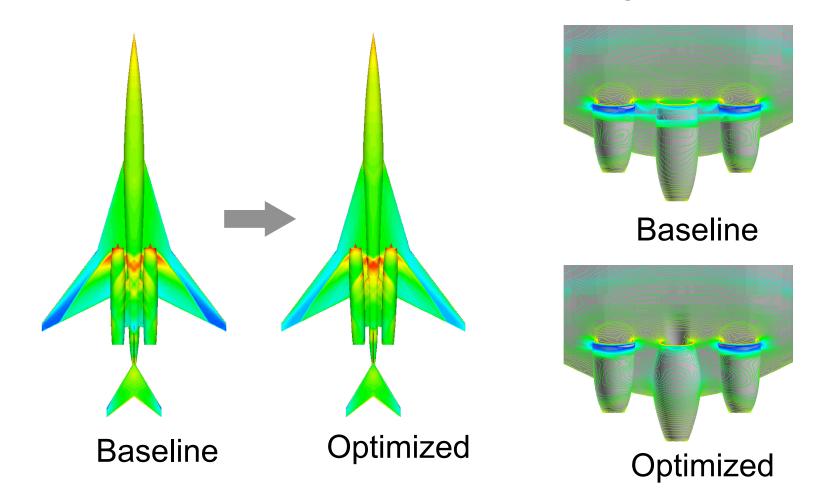
- Ideal for simple problems
- Focus on optimization efficiency



## **First Generation MDO**

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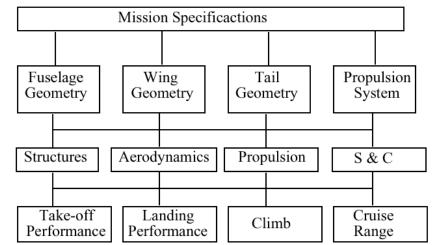
• SQP and related gradient-based methods have revolutionized aerospace design:



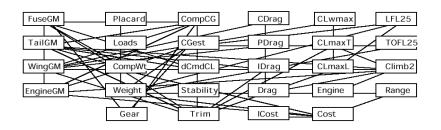
## **First Generation MDO**

 Integrated multidisciplinary analysis and optimization seems simple...

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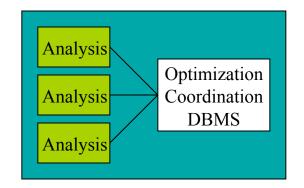


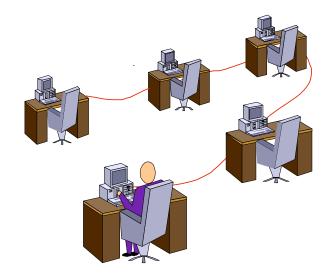
 ... but difficulties with complexity, extensibility, maintenance in real world, large-scale applications.



## **Second Generation MDO**

- Analysis management, distributed analysis, and optimization
- Modular analysis
- Focus on interdisciplinary communication



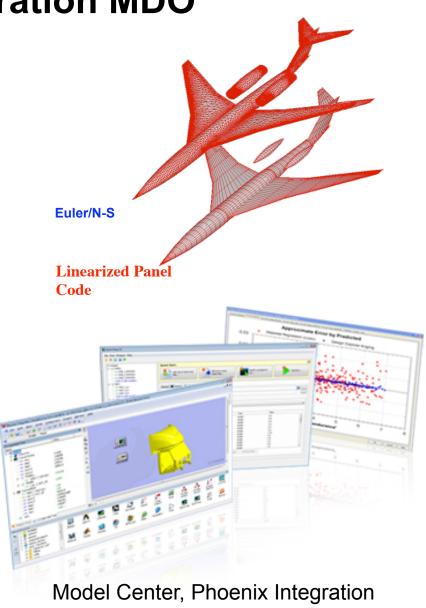


## **Second Generation MDO**

 Automation of analyses leads to process improvements

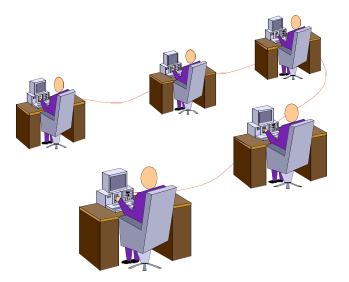
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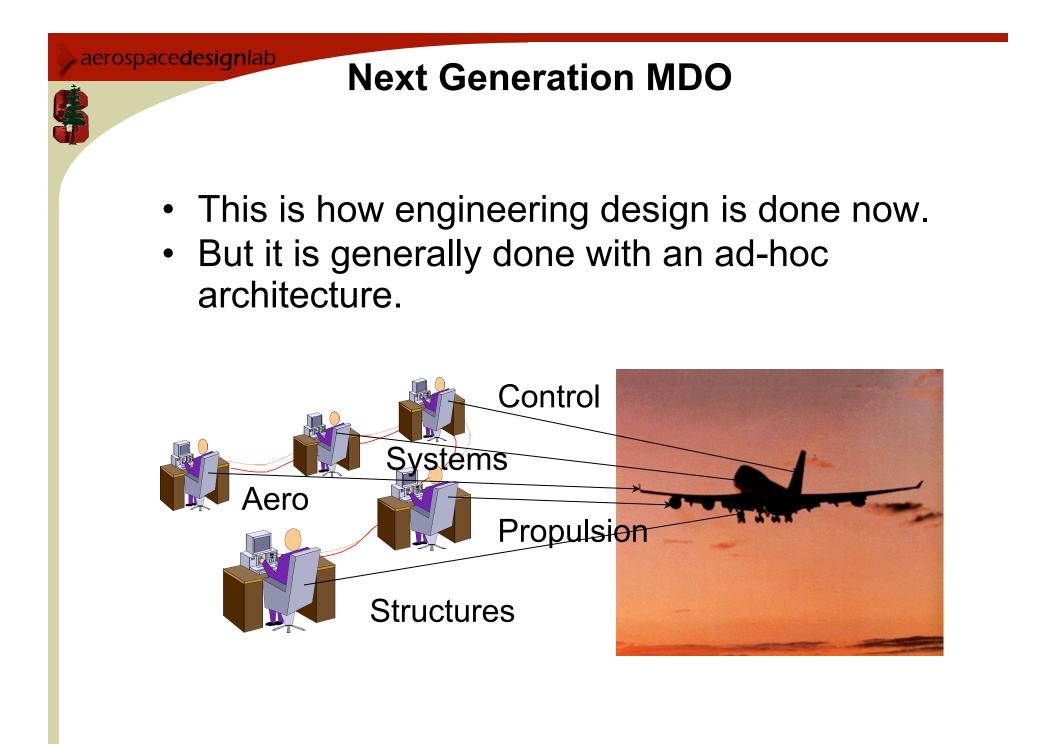
- Existing "recipes" prevent models of sufficient fidelity
- Infrastructure typically over-constraining



## **Next Generation MDO**

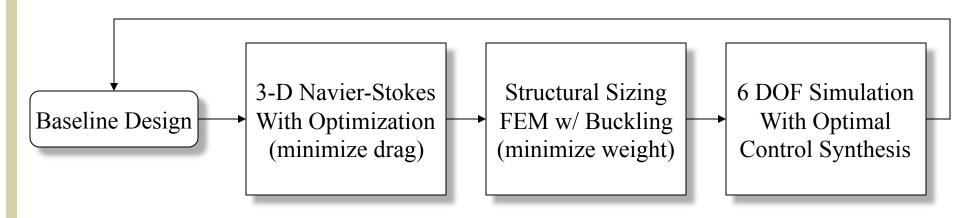
- The Goal: Distributed Design
- Why distributed design?
  - Improve efficiency
  - Manage complexity across scales
  - Permit use of disciplinebased design tools
  - Provide autonomy and exploit experience of design teams
  - Consistent with current organizational structures







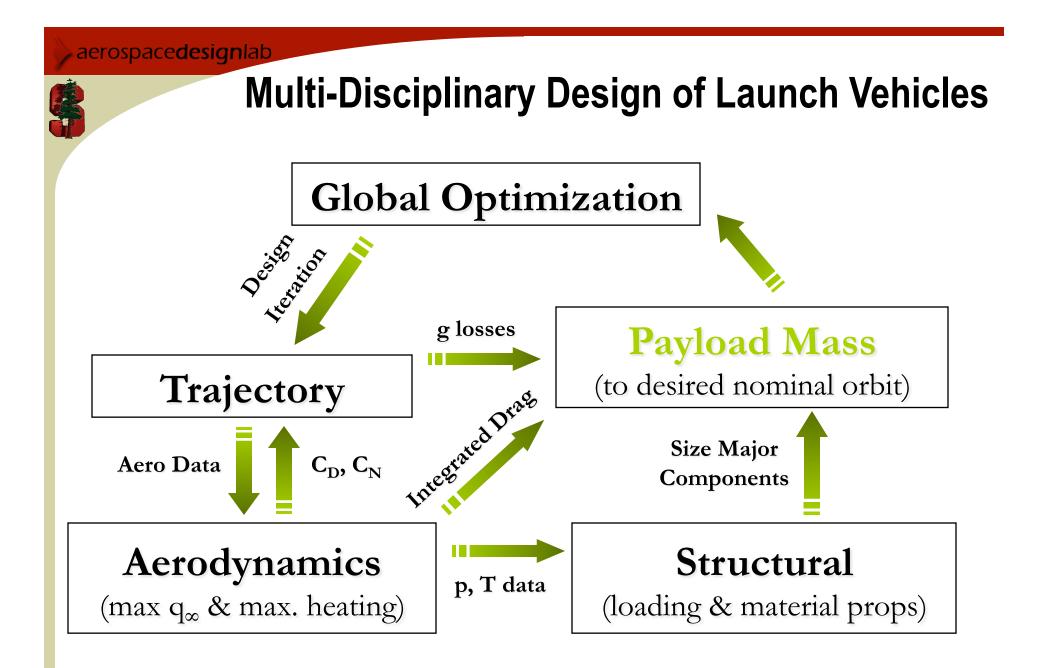
- Informal approaches often lead to incorrect results
- Iterations continue until one runs out of time or money



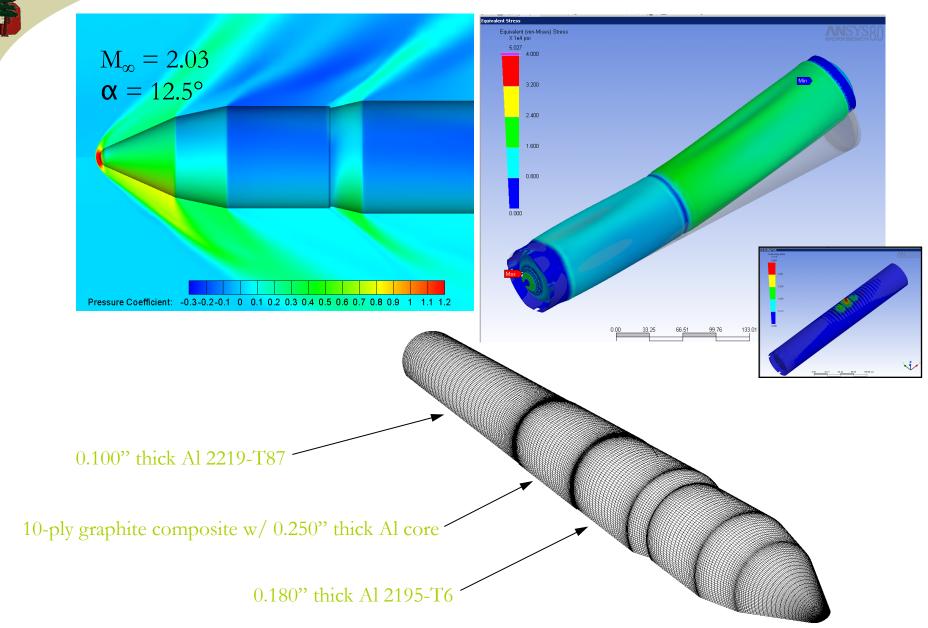
Actual Project Structure

## **Lessons Learned**

- Optimization techniques:
  - Which problems are best suited to which approaches?
  - What are the expected costs? Analysis? Sensitivities?
  - Single- vs multi-objective
- Approximation theory / surrogate modeling / trust region management
  - How can we reduce the cost of optimization?
  - What kinds of approximations are better suited to what problems?
  - How do we guide the analysis/optimization sequence to obtain accurate results robustly?
- MDO / Analysis / Design architectures
  - What are the best ways to partition a problem that is too hard to handle with 1<sup>st</sup> gen MDO?
  - How can we accelerate the analysis / optimization?
  - Have these methods been truly effective?



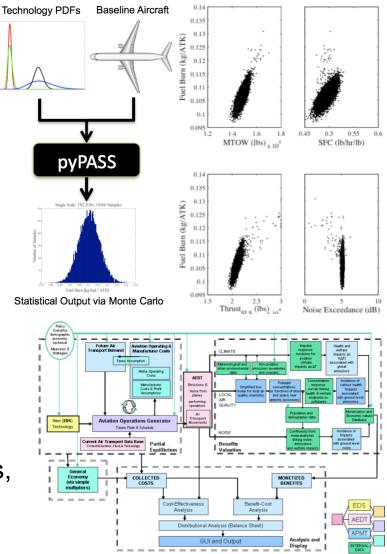
## **Multi-Disciplinary Design of Launch Vehicles**



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## **Design for Environmental Impact of the Entire NAS**

- Problem of similar complexity to wind power system analysis and design
- Probabilistic models for future aircraft fleet...
- flying the routes/schedules of multiple airlines...
- performance and economic models accounted for...
- interactions with current and future (NextGen) ATM procedures being investigated...
- FAA has pursued these tools/ environment for approximately 6-8 years, based on a previous legacy system (but with all components redeveloped). Sounds familiar?



FAA EDS/AEDT/APMT Schematic

# **Do These Apply to Wind Energy?**

- At the level of the isolated wind turbine, it is hard to argue against a fairly direct translation between aerospace (aircraft, launch vehicle, etc) and wind energy:
  - Similar complexity in the systems (including transient phenomena)
  - Similar disciplines involved (fluids, structures, reliability, loads, fatigue...)
  - Similarities in the design spaces to be explored.
- At a wind farm level, some techniques could be leveraged effectively:
  - Surrogate modeling
  - Multi-fidelity analysis and optimization techniques
- At the highest system level (equivalent to the complete National Air Transportation System), there are examples of how these problems could be tackled, but all answers are not available.

## What Are The Current Gaps?

- Because of the complexity of the analysis of even isolated wind turbines, wind-farm- and system-level simulations are likely to exceed the computational capabilities of the largest supercomputers.
- How can we manage the fidelity / accuracy / uncertainty vs. computational cost dilemma?
- What portions of the system-level models are fraught with epistemic uncertainties and how do we produce timely results to make decisions in a probabilistic environment?
- Our community is attempting to answer many of these questions at the moment.

# **Current Research Elements**

- Novel Optimization / Decomposition Methods
- Multi-fidelity System Modeling for Design
- Design Under Uncertainty

New Approaches for Distributed Design

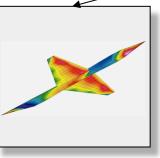
- Decomposition and distributed design
- Collectives and complex system design
- Efficient handling of highly-coupled MDO problems

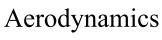
 Efficient handling of high-fidelity tools in MDO problems

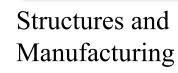
### Research Elements: Distributed Design

## Decomposition and Distributed Design

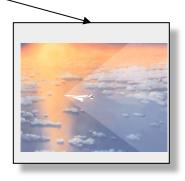
- Collaborative Optimization: A mathematical and software framework for large-scale design
- Collectives/Game Theory:
   Design of Systems of
   Interacting Agents







 $\frac{\sigma}{\sigma_{\max}} \le 1$ 



Environmental Compatibility

### Research Elements: Multi-Fidelity Modeling

Multi-Fidelity System Modeling for Design

- High-Fidelity design at the system level is prohibitively expensive
- Disconnect between high-fidelity analysis tools and lower-fidelity system tools
- Multi-Fidelity design frameworks provide the bridge –New methods using reduced order and surrogate modeling –Variable complexity design methods

### **Research Elements: Multi-Fidelity Modeling**

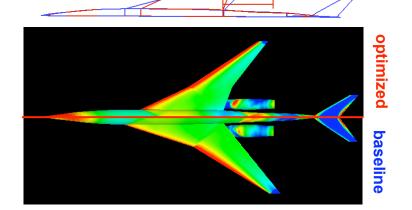
- Hierarchy of analysis methods
  - Constructed with analyst experience/intuition
  - Derived with formal reduced-order-modeling techniques
- Variable complexity model descriptions
  - Parameterization of design with different levels of fidelity
  - Parameterization of design environment (e.g. mission) evolving as design proceeds
- Formal management
  - Demonstrated techniques that dictate when it is appropriate to use one model vs. another
  - Guarantee of convergence to the true, highest accuracy optimum

### **Research Elements: Multi-Fidelity Modeling**

Level 1

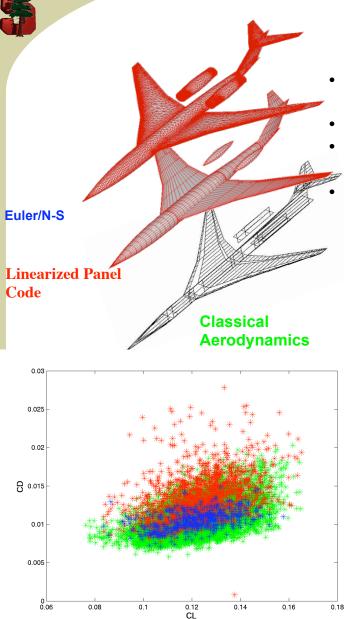
- Hierarchically enrich solution database and obtain high-fidelity approximations at low cost
- Surrogates used in realistic design (multi-objective, constraints)
  - Advanced, adaptive techniques (Kriging, RBF) to guide selection of all analyses in design space

Multi-modal optimization techniques to find approximate global optimum (GAs, Simplex, etc.)



### Level 2

Use adjoint-based, complete configuration, aerodynamic shape optimization techniques to recover performance that may have been lost by inaccuracy of the surrogate models in some regions of the design space Go to Level 1



## **Design Under Uncertainty**

- •Uncertainty in model, operation, parameters, requirements (failure/risk)
- •Essential element in design of complex systems
- •Multidisciplinary optimization traditionally with a narrow focus
  - -Continuous problems/models
  - –Deterministic

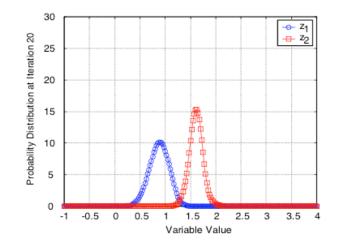
## Research Elements: Design Under Uncertainty

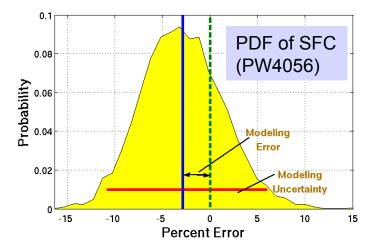
Design Under Uncertainty

Surrogate-based methods

•Monte Carlo simulation (importance based?)

- •Reduced-order modeling
- •Probability-based search methods





# **Some Overarching Comments**

Large-scale multi-physics simulations are coming of • age. They are beginning to be used in aerospace industrial practice and in wind power. However....



How do we take this?

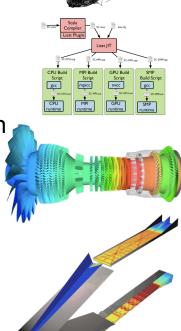
And turn it into this?

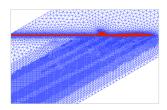
# Some Overarching Comments (II)

- Industrial design practice is *repetitive* by nature: *we must be able to repeatedly simulate with high fidelity*
- The real world is multi-physics and multi-disciplinary: we must be able to quickly and accurately integrate multiple physical scientific computing models
- The world is uncertain...input parameters are uncertain, models are uncertain: we must account for all sources of uncertainties in our simulations
- At the end of the day, engineering is about making a good system better: we must couple to optimization procedures

# Easier Said Than Done...

- Repeated simulations with high fidelity:
  - Fast and accurate solution procedures
  - Reduced order models
  - Leveraging high-performance computing architectures
- Integrating multi-physics models:
  - The science of integration: accuracy, stability, automation
  - The plumbing of integration: code coupling, parallelism, environments
- Uncertainty quantification:
  - Intrusive and non-intrusive techniques
  - Aleatory (non-reducible) and epistemic (reducible)
  - Curse of dimensionality
- Optimization:
  - Design with high fidelity: surrogates (multi-fidelity), adjoints
  - Design in "difficult" spaces: discontinuous and noisy
  - Design under uncertainty: robust design





## **Some Closing Comments**

- In the pursuit of a true, open, and flexible systemsengineering capability of wind energy systems, some key ingredients will be needed:
  - A flexible framework, based on open standards, to enable easy interfacing of multiple modules and testing of different ideas to properly architect the system
  - A hierarchy of component modules (analyses) with different fidelity/ cost ratios that are specifically designed for multi-disciplinary analysis and optimization
  - Single- and multi-fidelity surrogate modeling techniques (possibly reduced order models)
  - An uncertainty quantification framework to enable decision making (robustness, reliability, etc)
  - A variety of optimization techniques that are able to handle the variety of analyses and problem setups within the context of the framework.

# **Some Closing Comments (II)**

- Many frameworks exist (both commercial and open source): some experience with these and a representative sample of existing tools would be beneficial
- Clear articulation of intended goals / requirements for the framework will guide specifics of components and implementation
- Some logical progression of test cases (from single turbine to large systems) needed to validate approach and keep all developers honest
- "A journey of a thousand steps..."