



A Vision for High Fidelity Multi-Disciplinary Simulation Built Surrogates Influencing the Design and Assessment of Military Systems

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Definition: High-End Computing (HEC)



| Compute-Intensive | |
|---|---|
| Difficult | Easy |
| Tightly-coupled 100s to 1000s of nodes in lock-step | Embarrassingly parallel independent task / node |
| Fast / low- latency network | Slow / high- latency network |
| Fast / large memory and storage | Slow / small memory and storage |
| High Performance Computing | Basic Computing |

| Data-Intensive | |
|--------------------------------------|------------------------|
| Difficult | Easy |
| 10s to 1000s of accelerators | Few accelerators |
| 10s to 1000s of information feeds | Few information feeds |
| Large graphs | Small graphs |
| Large data sets | Small data sets |
| High Performance Data Analysis | Basic Data Analysis |

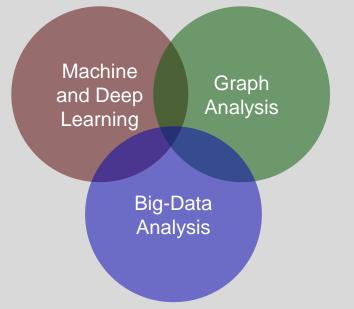
High-End Computing (HEC)

High Performance Computing (HPC)

+

High Performance
Data Analysis
(HPDA)

Data-Intensive Methods: a computer science approach to understanding phenomena and exploiting uncovered information



- Building artificial intelligence (AI) models (machine and deep learning)
- Uncovering relationships (graph analysis)
- Understanding large volumes of data ("big data" analysis)

Definition: Digital Engineering vs. MBSE



- Model-based systems engineering (MBSE) a formalized methodology used to support the requirements, design, analysis, verification, and validation associated with the development of complex systems (Nataliya Shevchenko, "An Introduction to Model-Based Systems Engineering (MBSE)," Software Engineering Institute (SEI), 21 Dec 2020)
- <u>Digital engineering (DE)</u> an **integrated digital approach** that uses authoritative sources of systems' **data** and **models** as a **continuum across disciplines** to support life cycle activities **from concept through disposal.** (Defense Acquisition University)
 - Integrated digital approach → a monolithic solution that suggests there can be no R&E / A&S divide
 - Data → must be regularly managed/updated/curated at each stage of the process
 - Models → both compute and data-intensive; range from basic analysis to high-end computing (HEC)
 - Availability

 accessible to anyone in the design, test, evaluation, sustainment, mission planning pipeline
 - Duration → from concept through disposal (and possibly beyond); decades to centuries
 - Extensity → continuum across disciplines; simultaneous accounting of all assets at all lifecycle stages to
 facilitate projection of current and future forces in a full range of scenarios, to determine gaps, to advise
 the characteristics of future platforms, and to build preventive maintenance plans for individual assets

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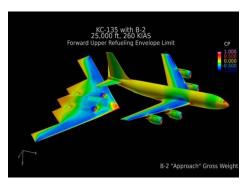
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HEC and High Fidelity
Physics (Physics
Based Analytics-PBA)
used for Deficiency
Diagnosis

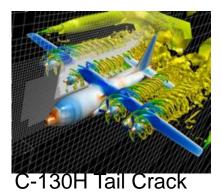
- Single Point Calc's (days/100k CPUHrs)
- Typically Occurs
 After Wind/Flight

 Test discoveries
- Fixes can be very costly since the vehicle is far into the design timeline
- Payoff for Programs (\$M's)
- Postdictive/Early Predictive



B-2/KC-135 Refuel





Increasing Program Relevance and Impact on Program Cost, Schedule, and Performance

Current Simulation Capability/Increasing Resources



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HEC and PBA used earlier in Design to Downselect From a Few Configurations

- Many More Single
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 Maneuvers
 (days/1Ms CPUHrs)
- Targets Specific
 Flight Cond's Mnvrs
 for Comparison and
 Risk Reduction
- Chosen Vehicle Better Meets Req's
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Accurate Surrogates built using HEC and PBA across the Life-Cycle

- Faster than real time Models
- Explore the Design Space
- Enable Mission Length Multiphysics Analysis
- Optimized Designs with Models of Many Disciplines Interacting
- Higher payoff for Programs (\$100M's)Predictive

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Revolutionary Approach to Building Accurate Models



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System Level Models

Built with Surrogates

- System Models with Associated Confidence Bounds
- Use AI/ML to Search Design Space for Deficiencies
- Virtual Flight Test
 Accomplished and
 Validated with
 Targeted FT
- Continual Model
 Updates Through
 Life Cycle
- Higher payoff for Programs (\$B's)Fully Predictive

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Revolutionary Approach to Building Accurate Models

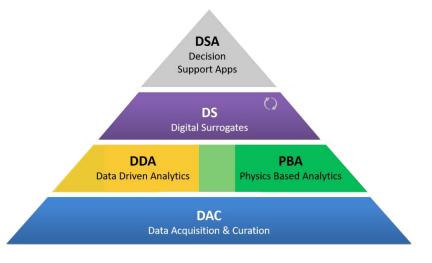
Physics Based Digital Engineering



Combat Air Vehicle Example



 Look at example Apps, Physics Informed Surrogates, Physics Based Analytics, Data Driven Analytics, and Data Acquisition and Curation



Physics Based Analytics - Aerodynamics



 Models are built to impact decision making in air vehicle design, development, and sustainment

PBA
Physics Based Analytics

- Examples Aerodynamics, Stability and Control (S&C), Propulsion, Structures
- High Fidelity CFD Models typically...
 - Take supercomputer resources to run depending on the fidelity of the mesh
 - Take flight conditions, surface and external volume meshes with boundary condition attributions as inputs
 - Compute the surface and volume aerodynamics of the vehicle using a finite volume or finite element solution algorithm
 - May be coupled with a structural solver, propulsion module, 6DOF solver, chemistry solvers for non-ideal gases, and conjugate gradient heat transfer solver to compute multiphysics simulations
- Reduced Order Loads Models typically...



- Are derived from simulations using the high fidelity CFD solver, sometimes coupled with other physics models (e.g. S&C 6DOF, Structures, Propulsion)
- Run faster than real time and can span the air vehicle envelope

Physics Based Analytics – S&C





- The Stability and Control Model is typically a 6 Degree of Freedom (6DOF) solver that...
 - Runs faster than real time and spans the vehicle envelope
 - Takes mass properties, vehicle configuration, and desired flight trajectory (e.g. stick inputs) or conditions as inputs
 - Uses information about vehicle aerodynamics, structures, and propulsion and the equations of motion to determine a vehicle's trajectory
 - Can incorporate "feedback control laws"
 - Can easily be improved for accuracy by replacing aerodynamics, structures, and propulsion data as higher fidelity information is known
- Stability and Control Model can be used for...
 - Stability and Control analysis and handling qualities
 - Sensitivity studies
 - Mission performance fuel burn, available payload weight, etc.
 - Driver for "Pilot in the Loop" Simulators





Physics Based Analytics - Propulsion



- The Propulsion Model has two variants
 - High fidelity turbomachinery integrated into a CFD solver
- PBA
 Physics Based Analytics

- Engine cycle deck
- Turbomachinery are geometry/mesh components (e.g. compressor or turbine rotor/stators, combustor) in the flowfield path of a high fidelity simulation tool
- The Engine Cycle Deck typically...



- Runs faster than real time and spans the vehicle envelope
- Takes flight conditions, compressor face conditions, exit nozzle conditions, and vehicle configuration as inputs
- Uses information about the engine and with one-dimensional engine pressure and temperature ratios through the engine stations, along with bleed losses and installation losses, determines compressor face and exit nozzle mass flow, velocity, and temperature data, as well as fuel burn and installed thrust
- Can be run offline, integrated into a high fidelity simulation tool (FORTRAN or NPSS Model) with pseudo-time accuracy with or without transient behavior, or integrated into an S&C 6DOF Model to provide thrust

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Physics Based Analytics - Structures



Two variants of Structures Models

- Finite Element Structural Model with a High Fidelity Simulation Tool
- Modal Structural Model

The Finite Element Structures Model typically...



- Takes from work station to supercomputer resources to run depending on the fidelity of the mesh
- Takes flight conditions, internal structure meshes with material property attributions, and external loads as inputs
- Computes the structural deformation, as well as, internal stress and strain of the structure using a finite element solution algorithm
- It may be coupled with an aerodynamic solver and/or a conjugate gradient heat transfer solver to compute fluid-structure, thermal-structure, or fluid-thermal-structural interactions

The Modal Structural Model typically...



- Is a linearized Reduced Order Model derived from the Finite Element Model through an eigenvalue analysis
- Runs orders of magnitudes faster than the Finite Element Model



Detail/Accuracy

Increasing

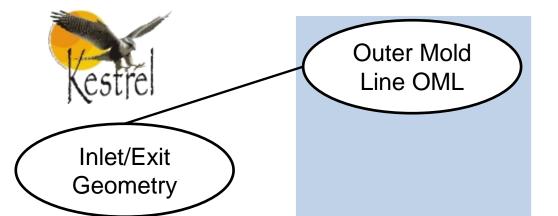
 During the design process we can eliminate poor design choices by increasing the fidelity of the PBAs as more information is known (objects and connection notional)



Outer Mold Line OML



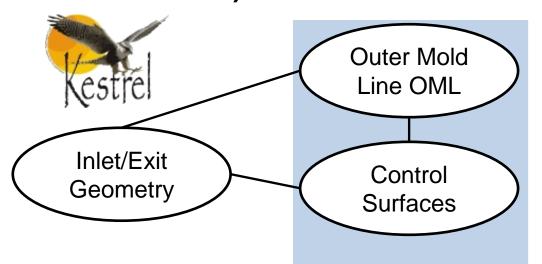








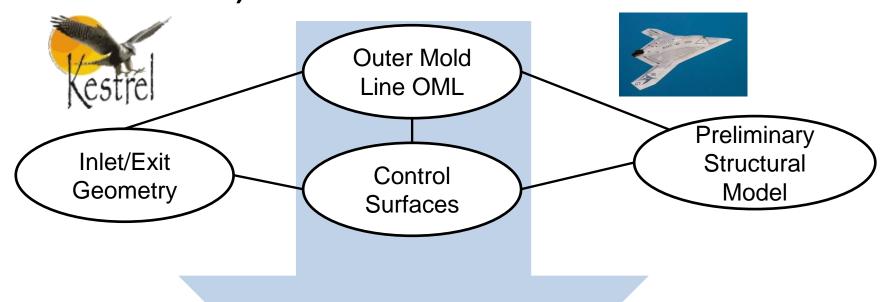










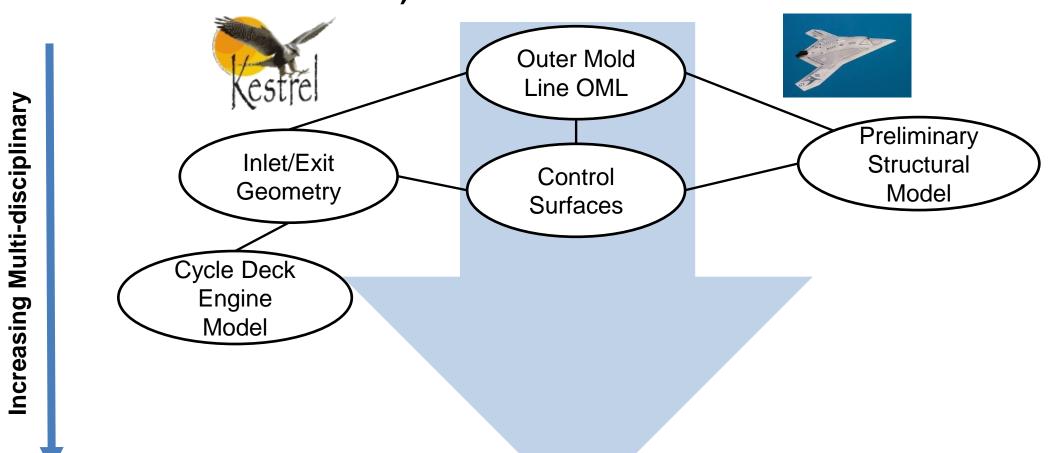






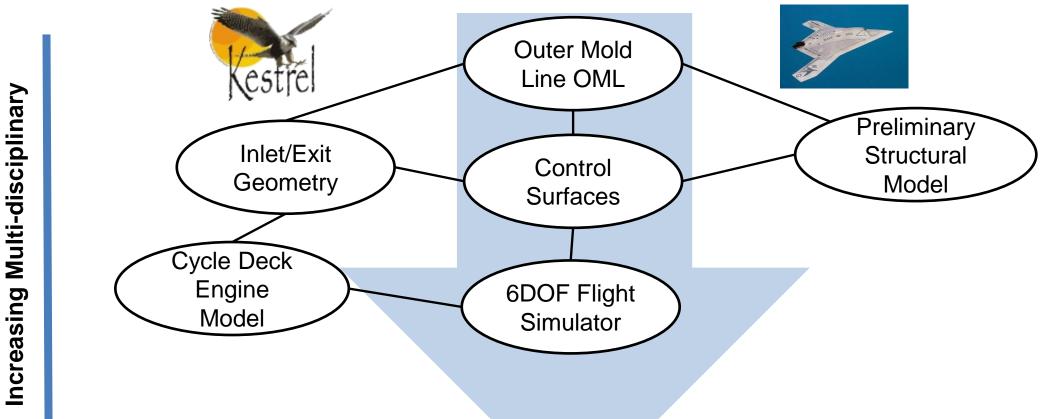
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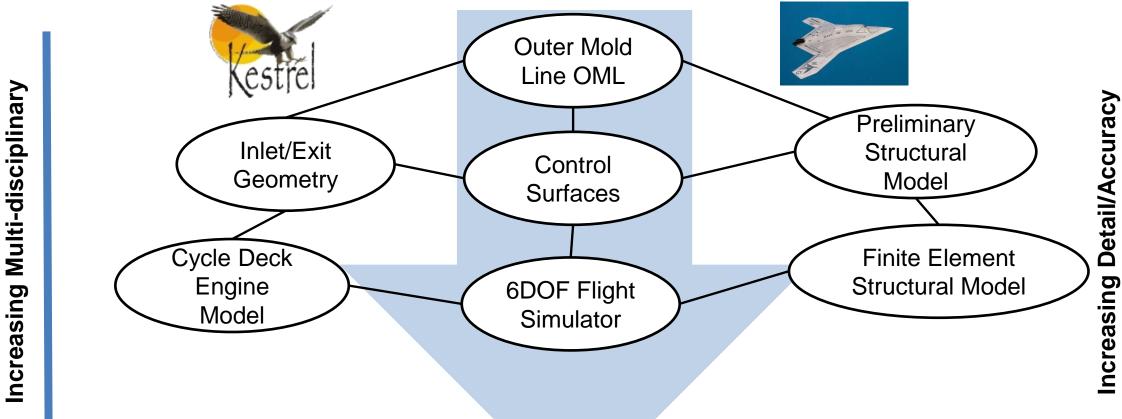






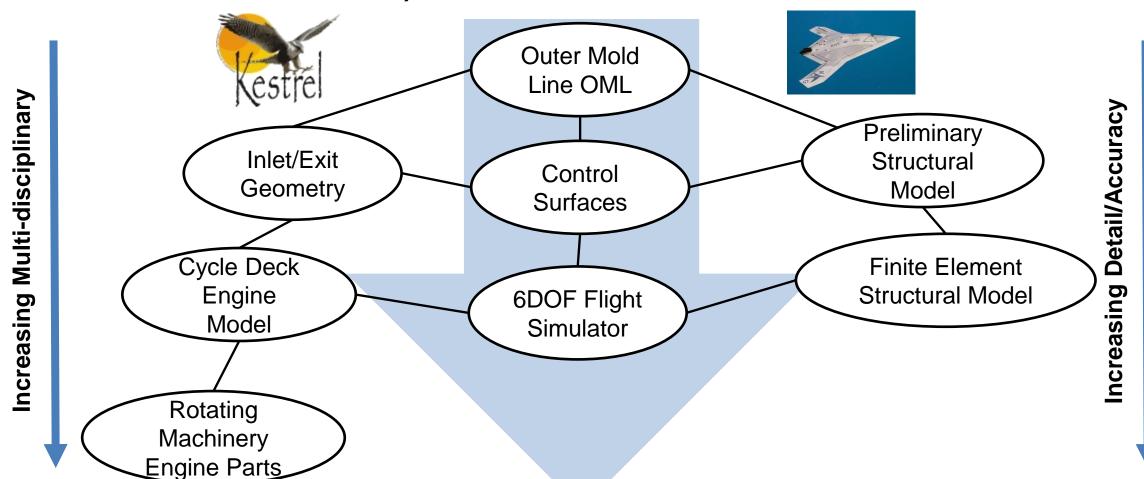






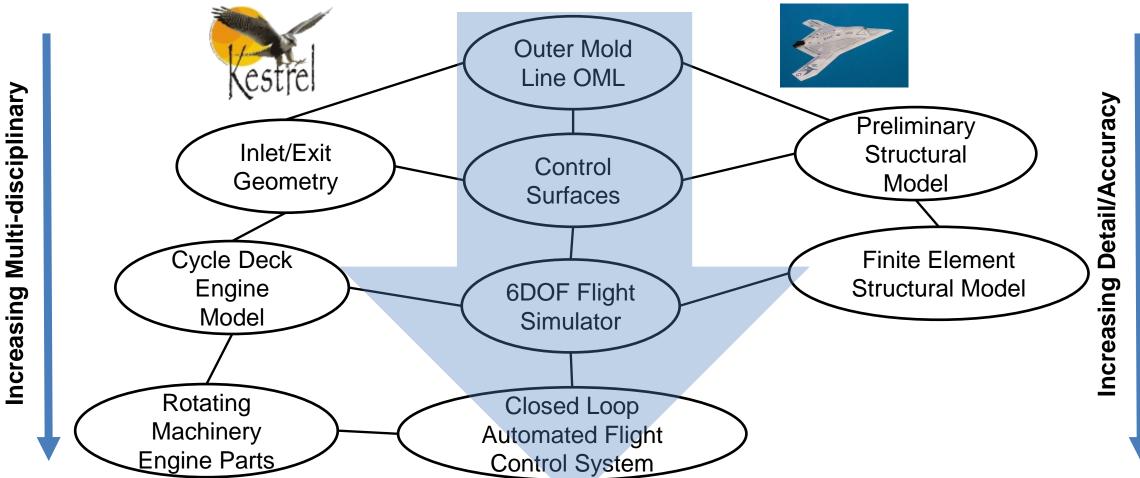








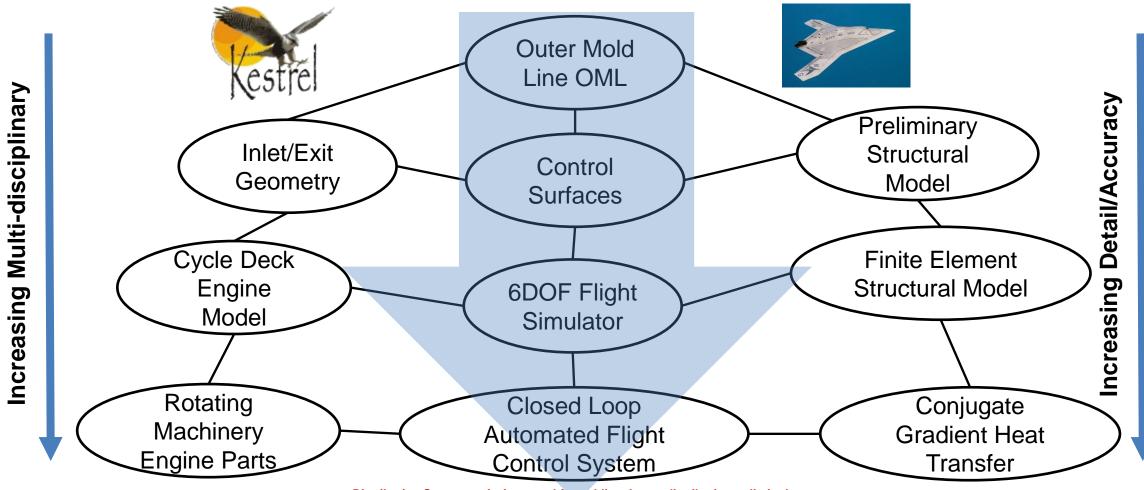








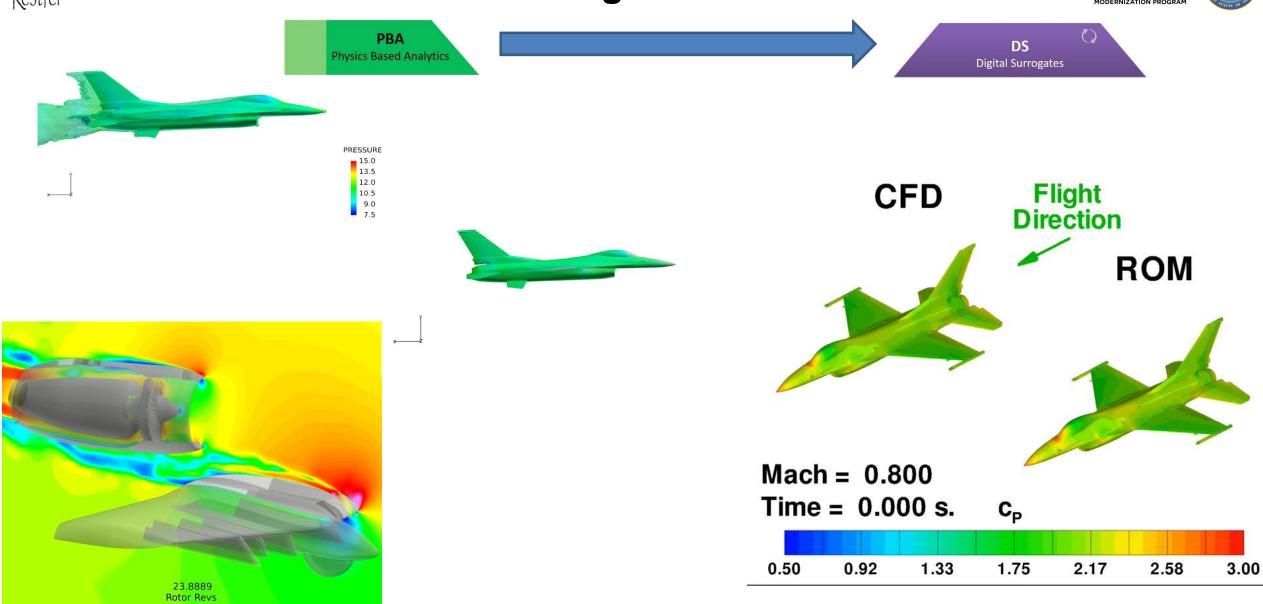
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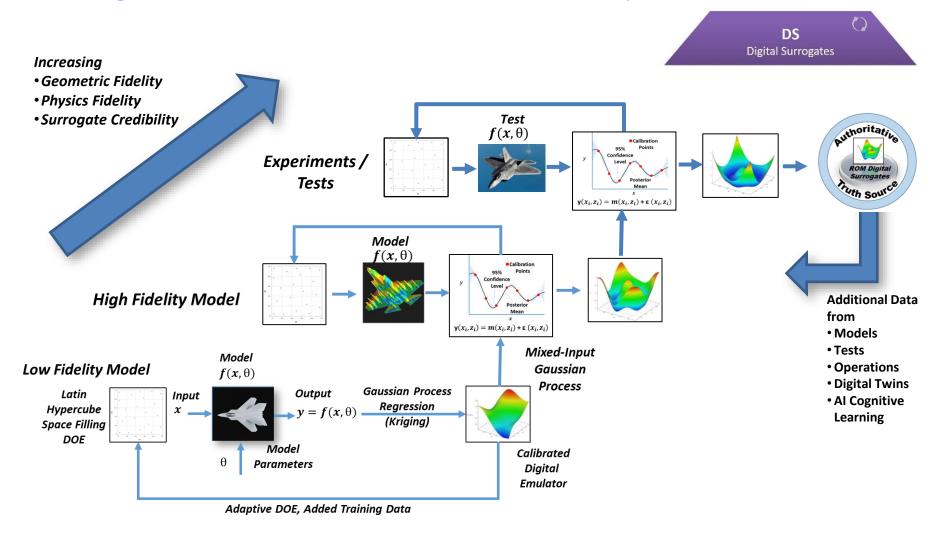
Reduced-Order Modeling





Example of Developing an Authoritative Digital Surrogate Reduced Order Model for Aerodynamics

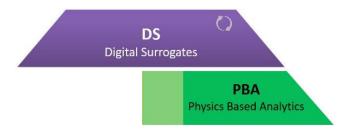




Edward M. Kraft, "Development and Application of a Digital Thread / Digital Twin Aerodynamic Performance Authoritative Truth Source," AIAA-2018-4003. Aviation Systems Conference, Atlanta, GA, June 25-29, 2018



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- There is also a wealth of information available (E.g. Ground test data, flight test data, fleet maintenance data)

DAC
Data Acquisition & Curation

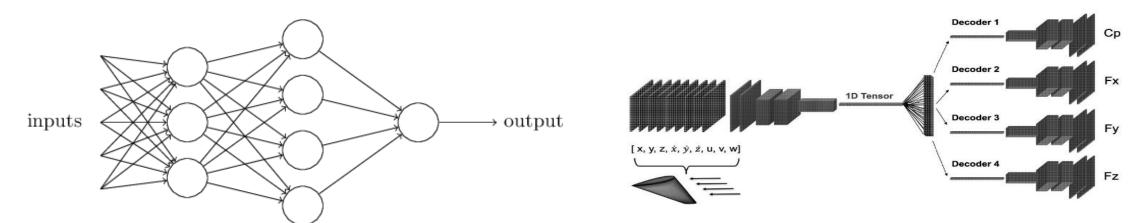


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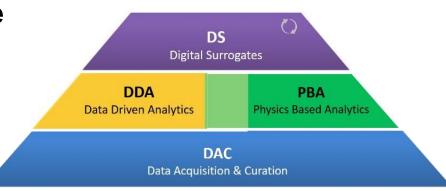


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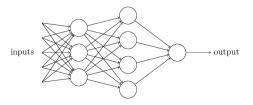


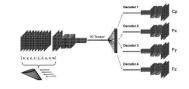


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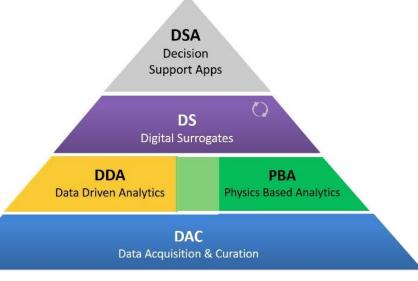








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- The most powerful surrogates are "constrained" by the physics to create a combined DDA/PBA surrogate that accounts for the un-modeled physics/tail number specifics
- Decision Support Apps can be built on this foundation of DDA/PBA Mission Analysis, Structural Failure Analysis, etc.



Concluding Remarks and Discussion



- The Physics Based Digital Engineering (PBDE) vision, such as described here, is the inevitable progression of technology (viz., high performance computing, physics, data analytics, and software engineering)
- The union of Machine Learning and Digital Surrogate Training via Physics-Based virtual test is what will deliver decision support data at the speed of relevance
- Machine Learning and Physics-Based virtual test both require HPC resources that can be reliably delivered by the HPC Modernization Program
- This PBDE vision can be applied to next generation combat air vehicle development through its Life-Cycle and specific examples have been shown
- Decision Support App software based on physics and system data can be pushed UP the leadership chain to aid in fast, accurate, decision making with high impact

Acknowledgements



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Questions?