



Design for real
Optimus[®]

MDO Methods and Tools

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Noesis Solutions

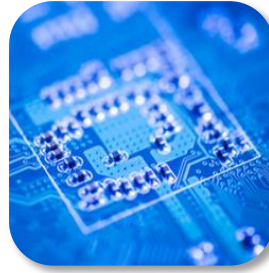
CEAS SCAD Symposium – November 25-27, 2014 – Onera, Toulouse

Outline

- Introduction about Noesis
- MDO Tools and methods: context & challenges
- High dimensional MDO
- Electrical Wire Harness case
- Conclusions

Noesis Solutions

who we are



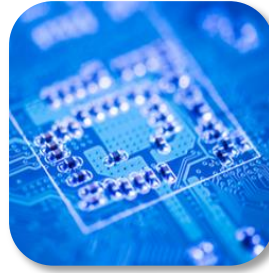
Noesis Solutions is an engineering innovation partner **to manufacturers in engineering-intensive industries.**

We are specialized in **simulation process integration & numerical optimization**

and help our customers resolve their toughest **multi-disciplinary engineering challenges.**

Noesis Solutions

what we do: Optimus



Our **Optimus software platform** enables customers to adopt an “Engineer by Objective” development strategy.

It **automates** ‘trial-and-error’ simulation based design processes

and directs simulation campaigns toward the **best product designs**.

Optimus by Noesis Solutions

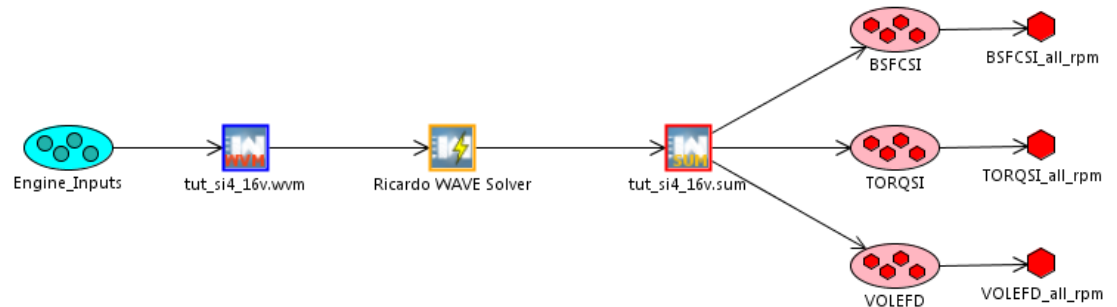
The leading vendor-neutral solution for engineering optimization

Process Optimization



- create automated, repeatable simulation processes
- integrate in-house tools & methods
- capture simulation knowledge
- custom applications

FASTER



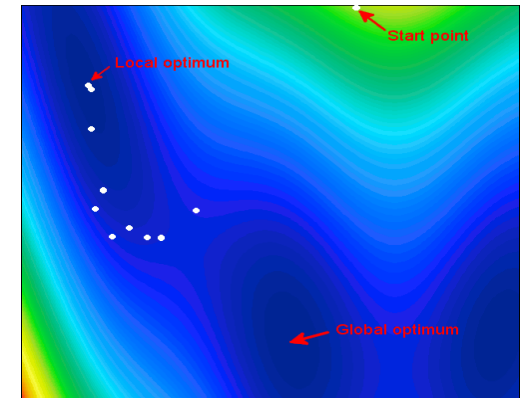
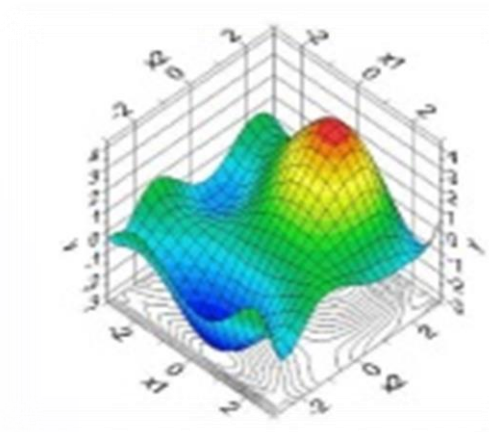
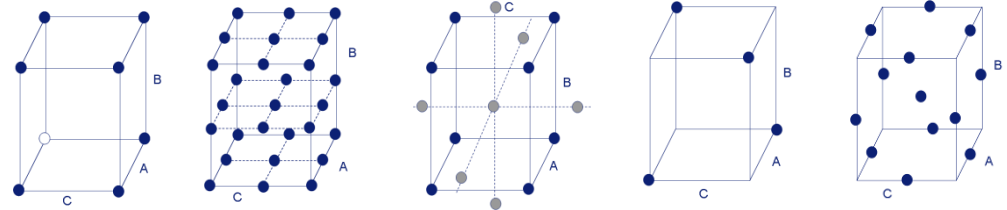
Optimus by Noesis Solutions

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Design Optimization



- identify influential design parameters
- gain critical engineering insights, enabling smarter decisions
- eliminate non-feasible designs, identify leading design candidates



BETTER

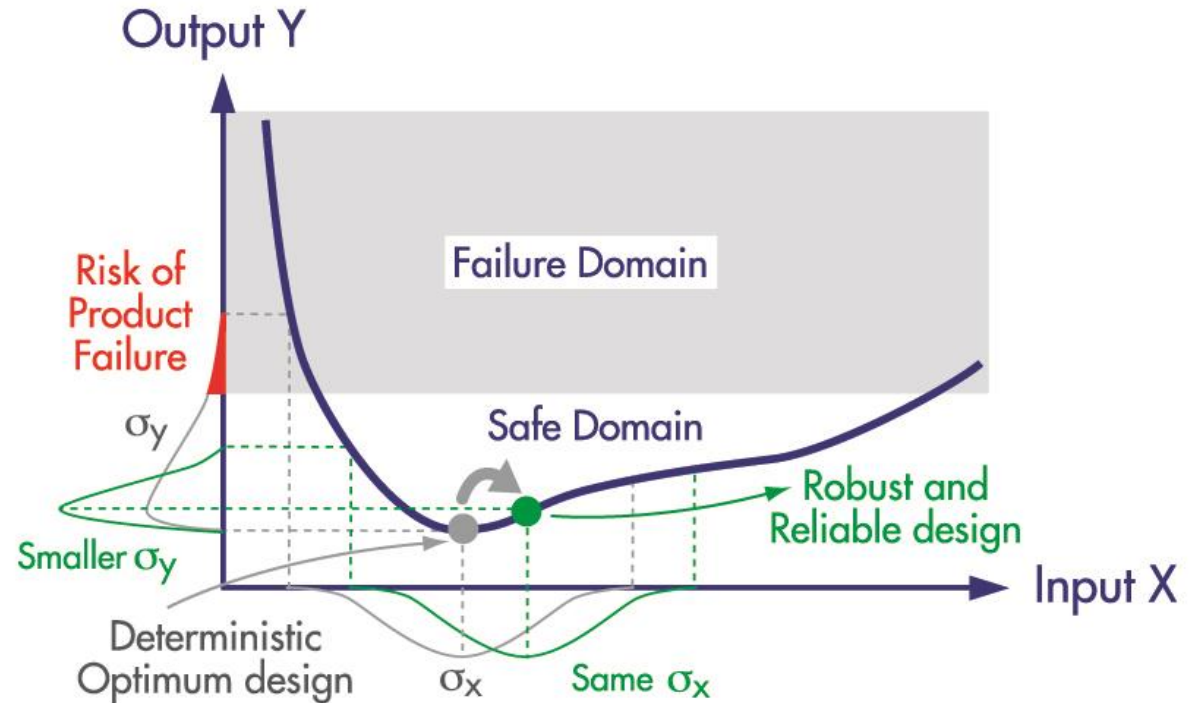
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Robust Optimization



- develop robust & reliable products
- ensure consistent product quality throughout product lifecycle



MORE RELIABLE

Open environment *with Python scripting*



Automating in-house processes



- Facilitate stress analysis
- Automate stress Report
- Formalize Simulation Process
- Reduce Human Involvement

Crafting dedicated applications



- Improve Active/Passive Crash
- Select Dominant LC
- Simultaneous Euro / USCAP
- Speed-up Crash Simulation Decision

Capturing corporate knowledge

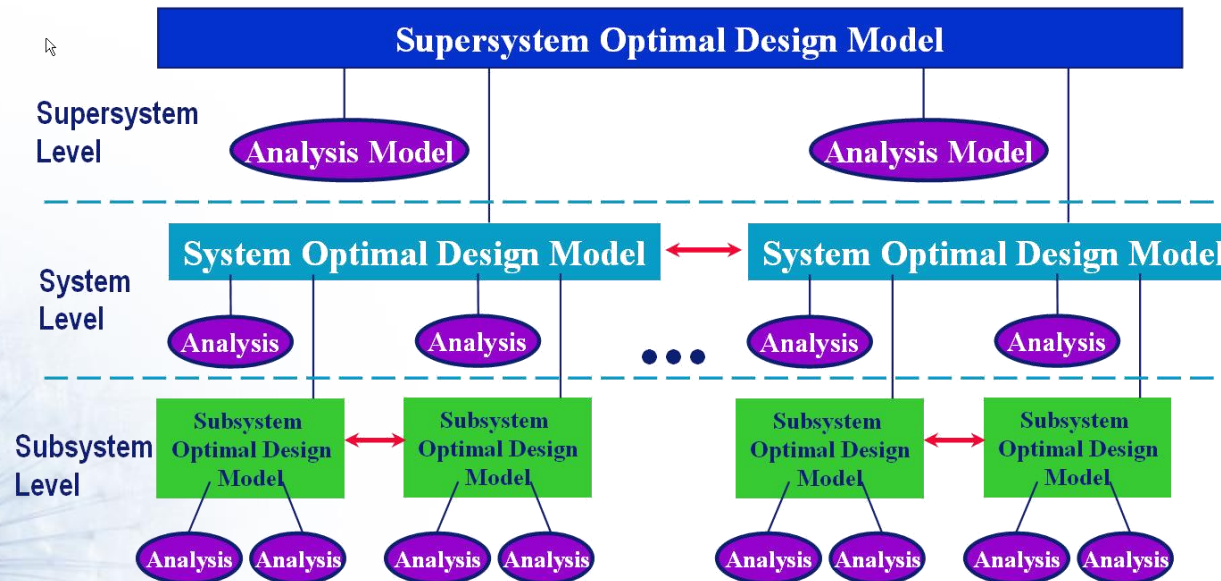
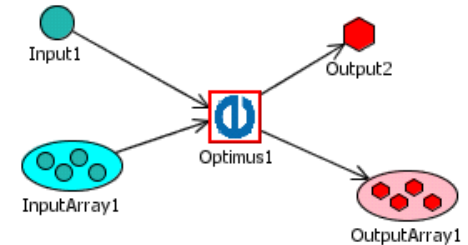


- Support selection of best optimization algorithm
- Build method cartography
- Validate deployment of methods

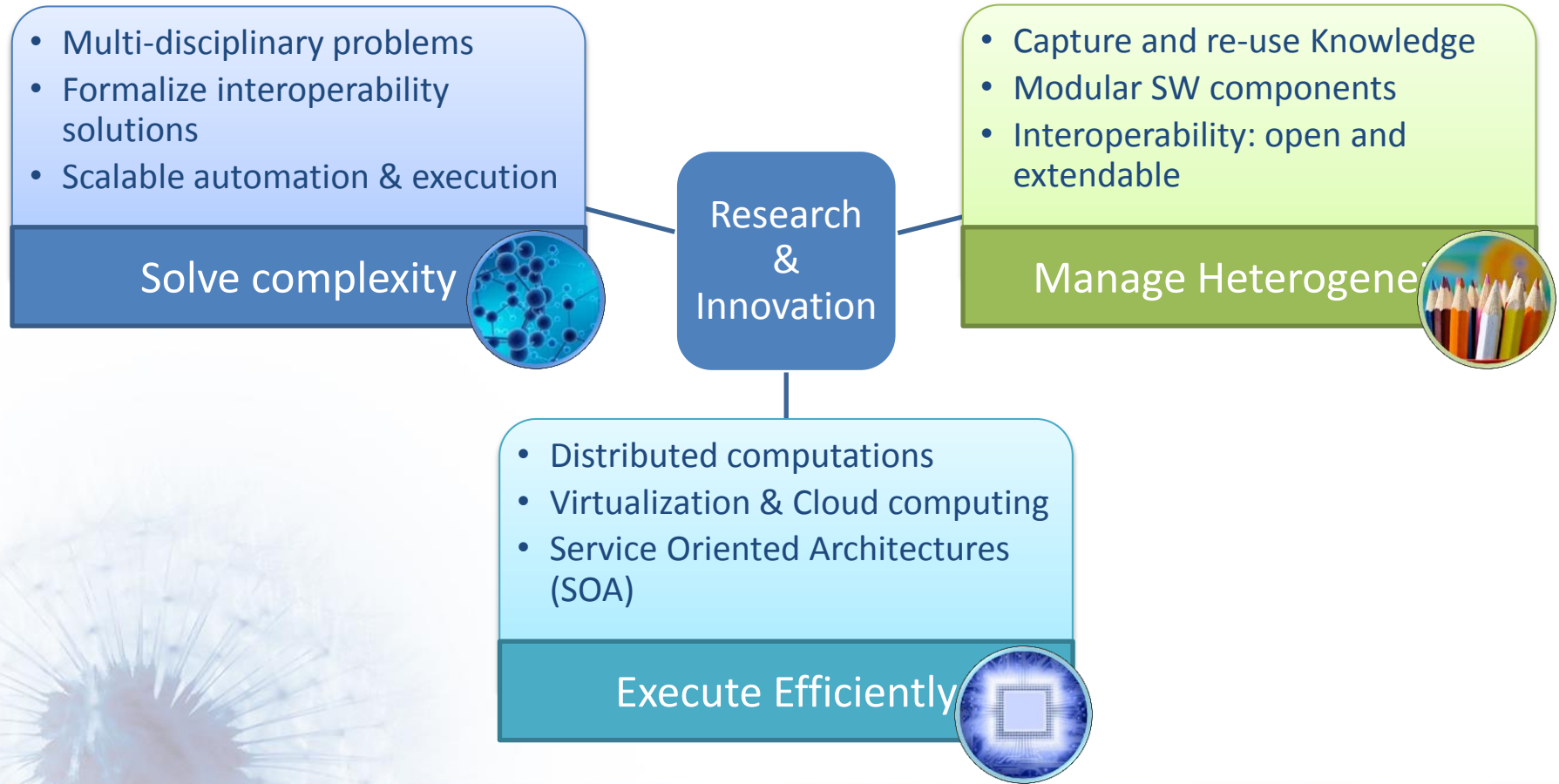
Open & extendable environment to respond to any custom need

... with a Multi-Level approach

- Multi-Level environment:
 - Drive simulation cascaded process with OPTIMUS
 - Handle Bottom-level workflow with a Top level OPTIMUS
 - Carry Up and Down Inputs/Outputs/Constraints



Research and Innovation



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Context

- Complex engineering problems use high fidelity simulation models and expensive experiments
- Ever increasing competition imposes time and budget constraints on the development of new products
- Need to infer maximum amount of information
 - from limited number of simulation runs
 - obtain competitive solutions

Problem

- We never achieve 100 % accurate simulation or experiments
 - if they are high fidelity, they are expensive.
- Problem scale keeps increasing
 - number of design parameters and data size
- Design is progressive
 - The user initially does not know proper parameter bounds, constraints, what to model and what to ignore

What we can trade in for finding an “optimal solution”

- Quantitative accuracy of performance is not always of primary importance but **relative merit** is
- Very often a “**competitive**” solution is all we need, **not necessarily the best possible** solution
 - Especially true if the problem is multi objective or highly constrained (best **difficult or expensive to achieve**)
 - Overall sound solution important in safety critical applications (best **too good to be true**)
- The point is to have **different shades of optimality**

Why accept such relaxation?

- “Best” is capricious
 - Predicting World Cup top 8 teams is easier than predicting the winner.
 - Engineers can usually agree on good designs but may disagree on the “best one”.
- “Best” depends
 - We are usually uncertain of what we mean by best.
 - Optimization problem set up is an art, not science.
- “Competitive” is intuitive
 - Humans are good at discerning it
 - We could allow human expertise and past statistical data to take part in the search for solution

Large Scale MDO Challenges

- Large design space to explore involving heavy sampling
- Risk to linearize design space by reducing sampling
- High probability to violate constraints
- Challenge to bound a confidence region that fulfills all constraints AND improves the design
- Access to an adequate environment
 - Visualize the data and rely on metrics to validate the optimum
- Hardware path
 - Increase CPU Services with HPC Cluster
 - Increase RAM to store large matrix
- Simulation Solver Path
 - Enable analytic sensitivities directly in the solver



How to do it?

- Differentiate Dimensional and Domain reductions
- Identify inter-relations and use them
- Methodologies:
 - Graph decomposition
 - Feasible region identification
 - Deep Learning and SOM based adaptive sampling (SOMBAS)
 - Domain shrinkage using Sobol Indices (DSSI)
 - Interaction Indices

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Research and Innovation tracks



Simulation Framework

- Web 2.0 based corporate federation toolkit
- Dynamic multi disciplinary design workflow



Machine Learning

- Knowledge capture and reuse,
- Optimize large, highly dimensional full scale models



Cloud and HPC

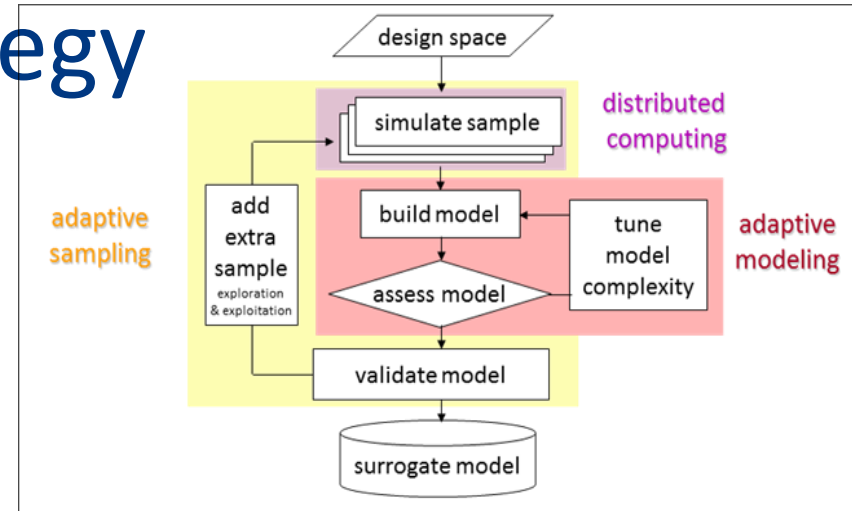
- Virtualize execution via cloud computing
- Web Services and access from everywhere

HAROS-HD

- Hybrid Adaptive Robust Optimization Strategy for High Dimensional systems
 - Optimization strategy (opposed to optimization algorithm)
- Main Objectives
 - Provide new optimization techniques suitable for large scale systems with mixed variables
 - Discover problem 'features' and use them for efficient optimization of subsystems

Overview of the strategy

- Multiple steps involved



Problem
decomposition

Dimensionality
reduction

Feasible region
identification

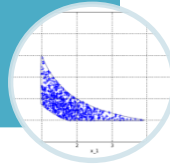
Sub-system
optimization

Tuning

Main stream technologies

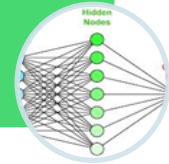
- Non-convex space density learning
- Unsupervised feasible region identification
- Initial space indifferently in feasible/unfeasible solution space

Find feasible solution space



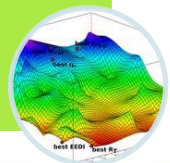
- Adaptive learning sampling
- Identify design space features (ex: search of non-linearities)
- Improve accuracy in optima surrounding
- Mixed variable algorithms (HAROS-HD)

Machine learning:
Sampling/optimization



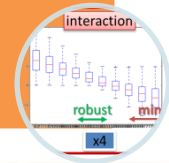
- Aggregate meta models based on confidence maps
- Estimate confidence maps with Neighborhood identification
- Learning profiles to model non-linearities, minima/maxima, etc

Machine learning:
Ensemble modeling



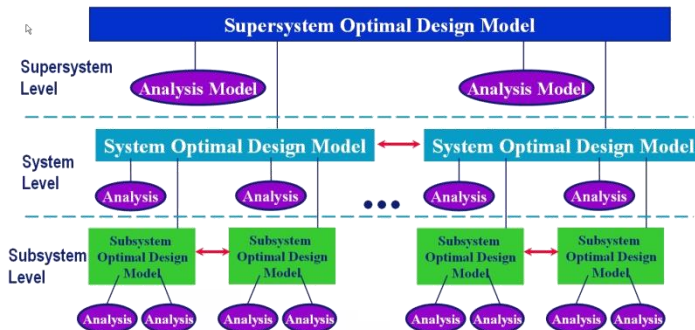
- Decompose large problem in smaller and spot robust solution
- Use interaction & Sobol indices
- Measure degree of interaction between parameters
- Detect parameters independence

Dimension reduction

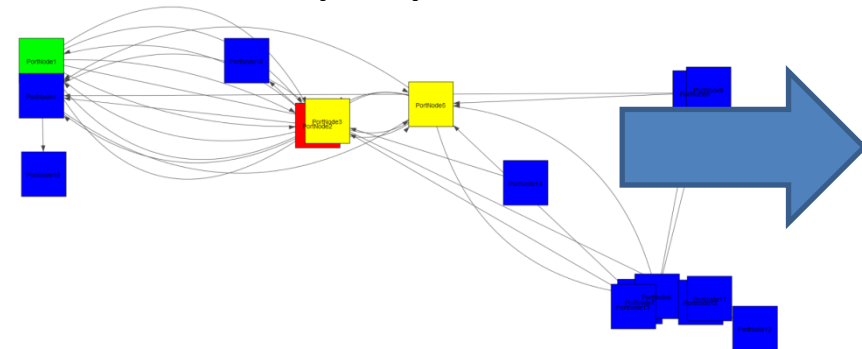


MD Dimension reduction approach

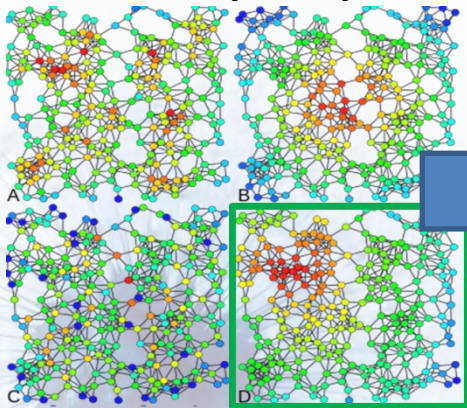
MDO problem formulation



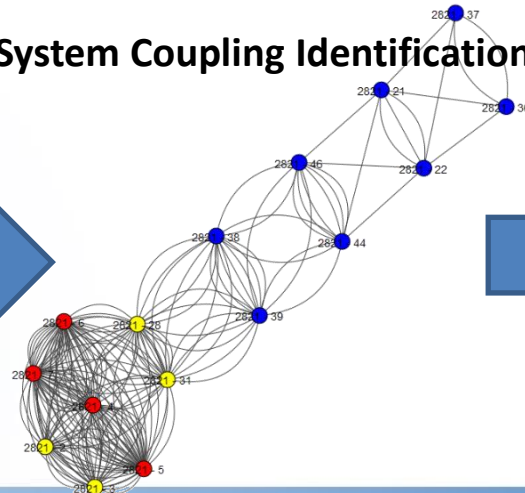
MDO Graph representation



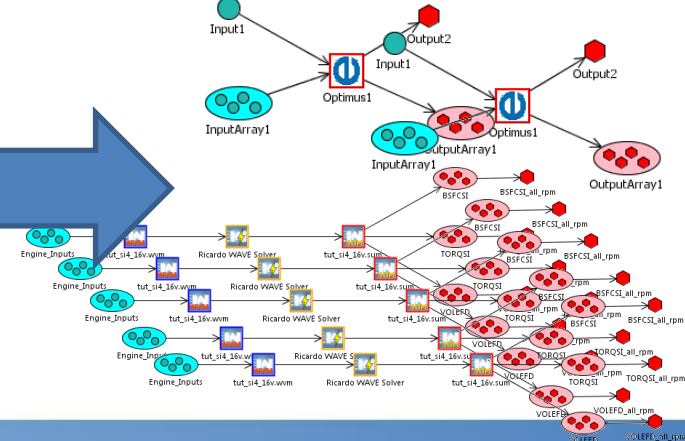
Centrality Analysis



System Coupling Identification



Executable simulation WF



Advantages

- **General:** can be applied to all optimization problems
 - HAROS-HD will learn about the design space and employ the appropriate algorithms as it proceeds toward finding an optimized solution
- **Flexible:**
 - HAROS-HD will intelligently adapt the optimization strategy by selecting the most appropriate method to use
- **Scalable:**
 - It can be used for high dimensional problems

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- **Electrical Wire Harness case**
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Context

$>10^3$
Wire
harnesses

$>10^5$
Electrical
links

$>2*10^3$
Kg of
electrical
installation

$>4*10^4$
Pathway
segments

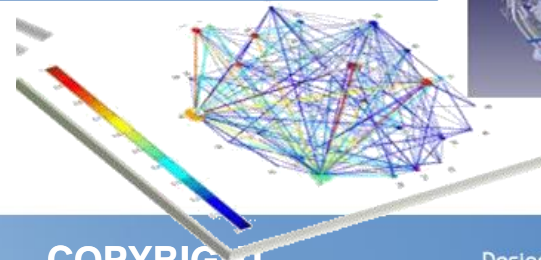
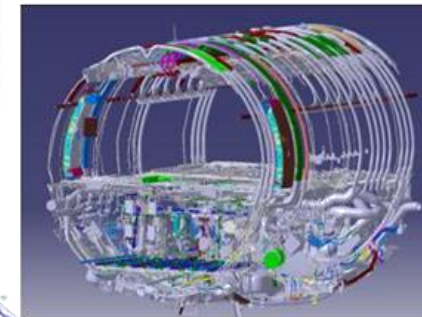
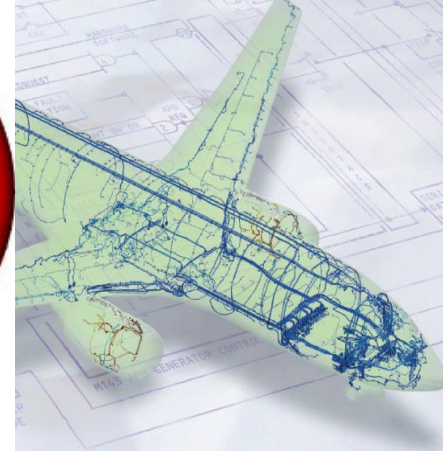
$>10^4$
Connectors

$>3*10^4$
Installation
interfaces

$>1.5*10^5$
Meters of
cable

$>1.5*10^5$
Signals

Typical
single aisle

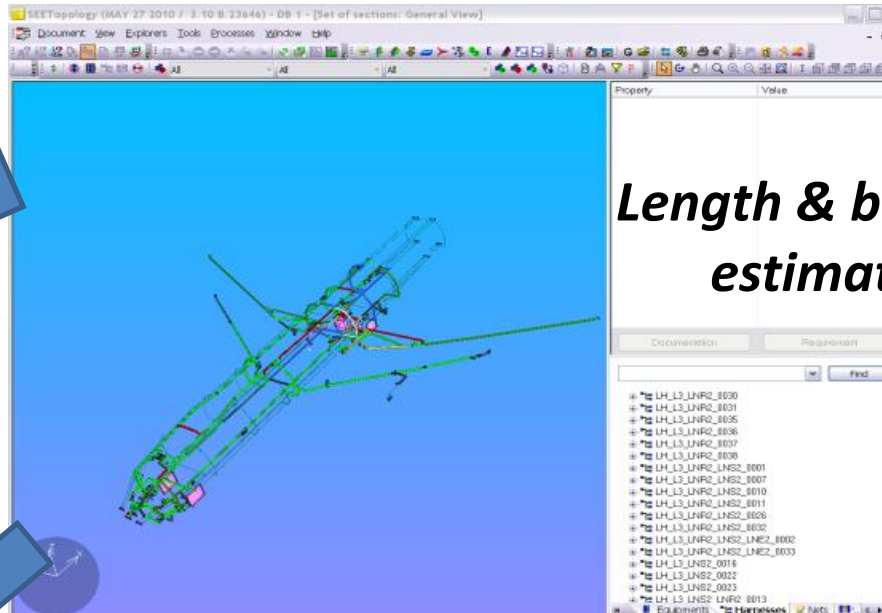
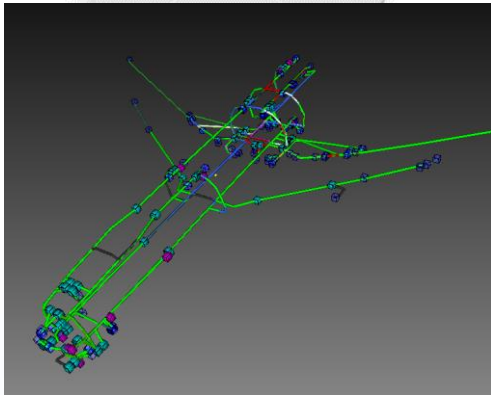


Wire design

ROUTING SYSTEMS

In the space reservation

Respecting safety & segregation
rules

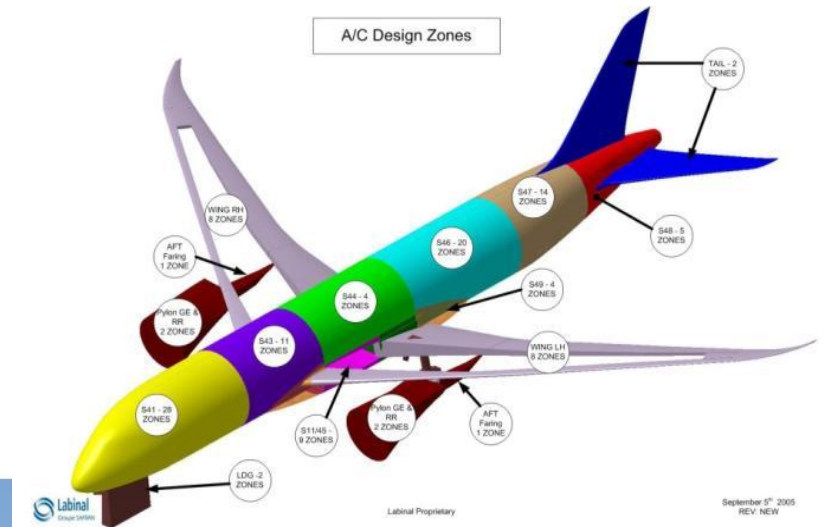
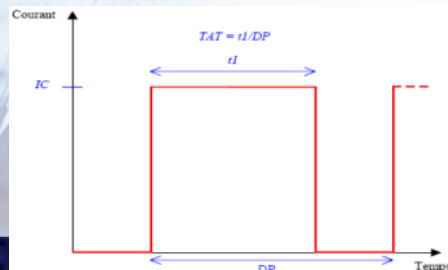
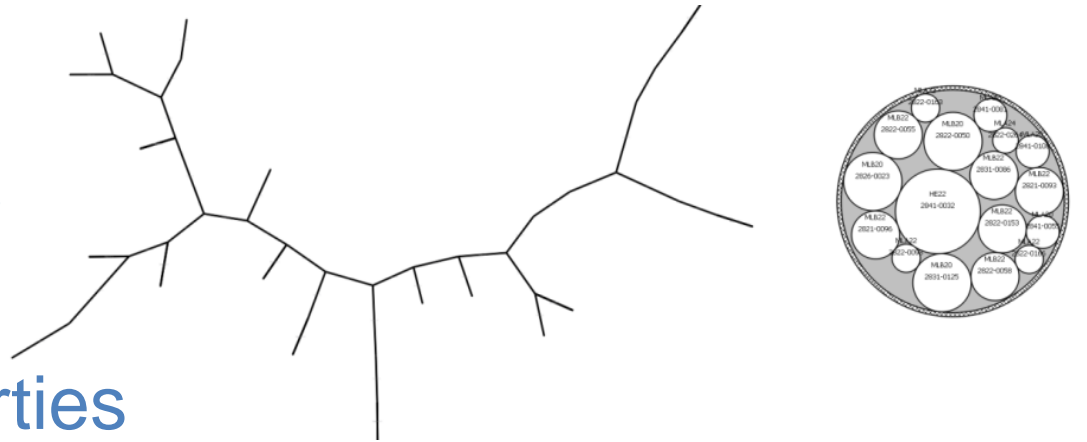


*Length & bundling
estimation*



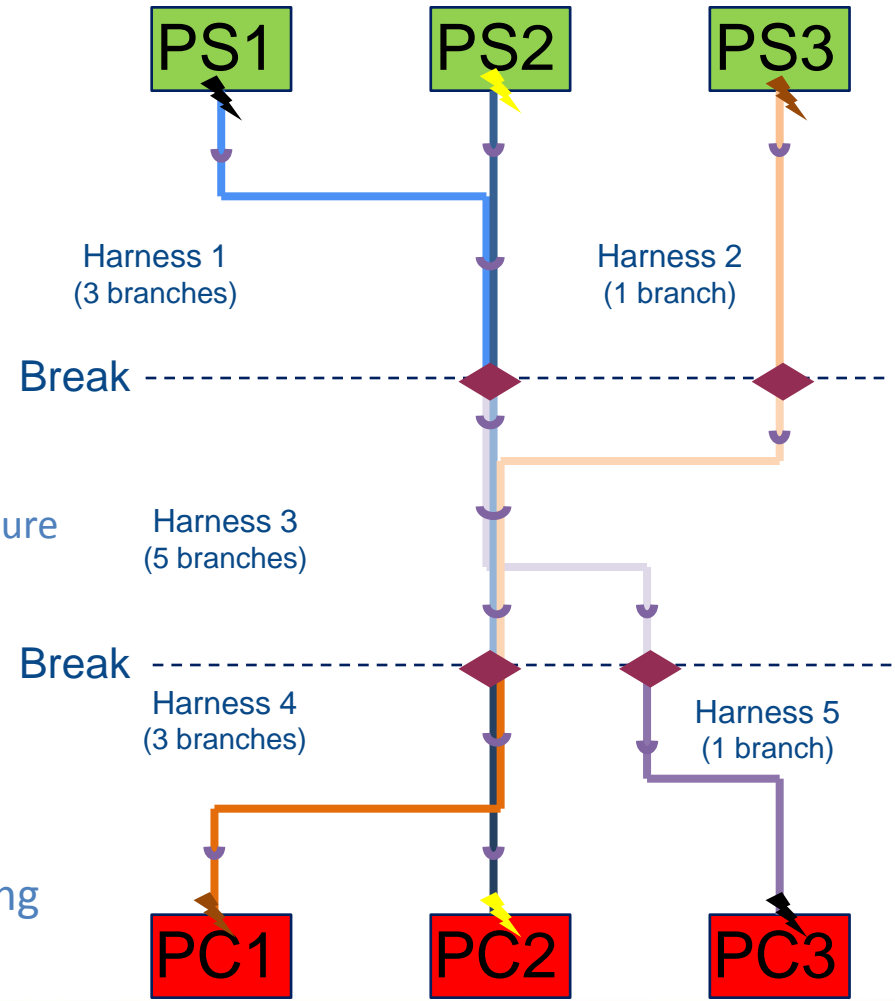
Cable sizing

- Harness topology
- Harness protections
- Cable properties
- Electrical link properties
- Aircraft flight phases
- Aircraft areas
- Environmental parameters



A typical industrial case

- **Wire gauge optimization**
- **Cable design**
 - Mechanical design
 - Tensile strength
 - Electrical design
 - Maximum voltage drop
 - Thermal design
 - Maximum overheating & temperature
- **Problem definition**
 - Harnesses interactions
 - Voltage drop computation
 - Cables interactions
 - Overheating computation
 - Voltage drop depends of overheating
 - Discrete problem



Optimization challenges

- Cable gauges optimization problem
- Large scale system
 - Up to 1000 wire harnesses
 - ~10000 cables - optimization parameters
 - ~40000 harnesses branches - thermal constraints
 - ~10000 electrical links – voltage drop constraints
 - Discrete problem (categorical)

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Conclusions

- Complexity and heterogeneity challenges need to be addressed in MDO
- High dimensional systems soon become common
- New methods and tools exist to develop new optimization strategies for large systems
 - Paradigm shift in classical design optimization approaches



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Thank you!

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