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Cybermatrix: A novel approach to computationally and collaboration intensive MDO for transport aircraft design

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Knowledge for Tomorrow



Two backgrounds of MDO for aircraft design

- Background 1: classic aircraft design
 - Focus on process automation, many disciplines, data modeling
 - No specific focus on high-performance computing (HPC)
 - No formal optimality criteria, suboptimal designs by construction
- Background 2: mathematical optimization
 - Focus on analysis fidelity, modeling constraints and adding disciplines
 - Explicit consideration of optimality criteria and often high HPC use
 - Simplified tools, poorly scalable in number of disciplines/experts
- **The present approach aims at balancing the two backgrounds**
 - Developed within the **DLR project VicToria**
 - Optimality criteria explicit but applied in a heuristic manner
 - Parallelism built in ground-up, in participation of experts and use of HPC
 - Assembly (human) and execution (computer) phases with analogous communication and control in a matrix-like structure → **cybermatrix**



Approximate optimal design

➤ Any design process can be viewed as an **approximate** optimization process:

$$\frac{df(\widehat{w}, p)}{dp} - \lambda \frac{dc(\widehat{w}, p)}{dp} = 0, \quad c(w, p) = 0, \quad r(w, p) = 0$$

where f goal, c constraints, r consistencies (residuals), w states,
 p design parameters, λ constraint scales (Lagrange multipliers)
 → approximate KKT optimality condition

➤ Expanded for three disciplines A, B, C and global goal function F:

$$\frac{\widehat{\partial F}}{\widehat{\partial f_A}} \frac{\widehat{df_A}}{dp_A} + \frac{\widehat{\partial F}}{\widehat{\partial f_B}} \frac{\widehat{df_B}}{dp_A} + \frac{\widehat{\partial F}}{\widehat{\partial f_C}} \frac{\widehat{df_C}}{dp_A} - \lambda_A \frac{\widehat{dc_A}}{dp_A} - \lambda_B \frac{\widehat{dc_B}}{dp_A} - \lambda_C \frac{\widehat{dc_C}}{dp_A} = 0, \quad \underline{c_A} = 0, \quad \underline{r_A} = 0$$

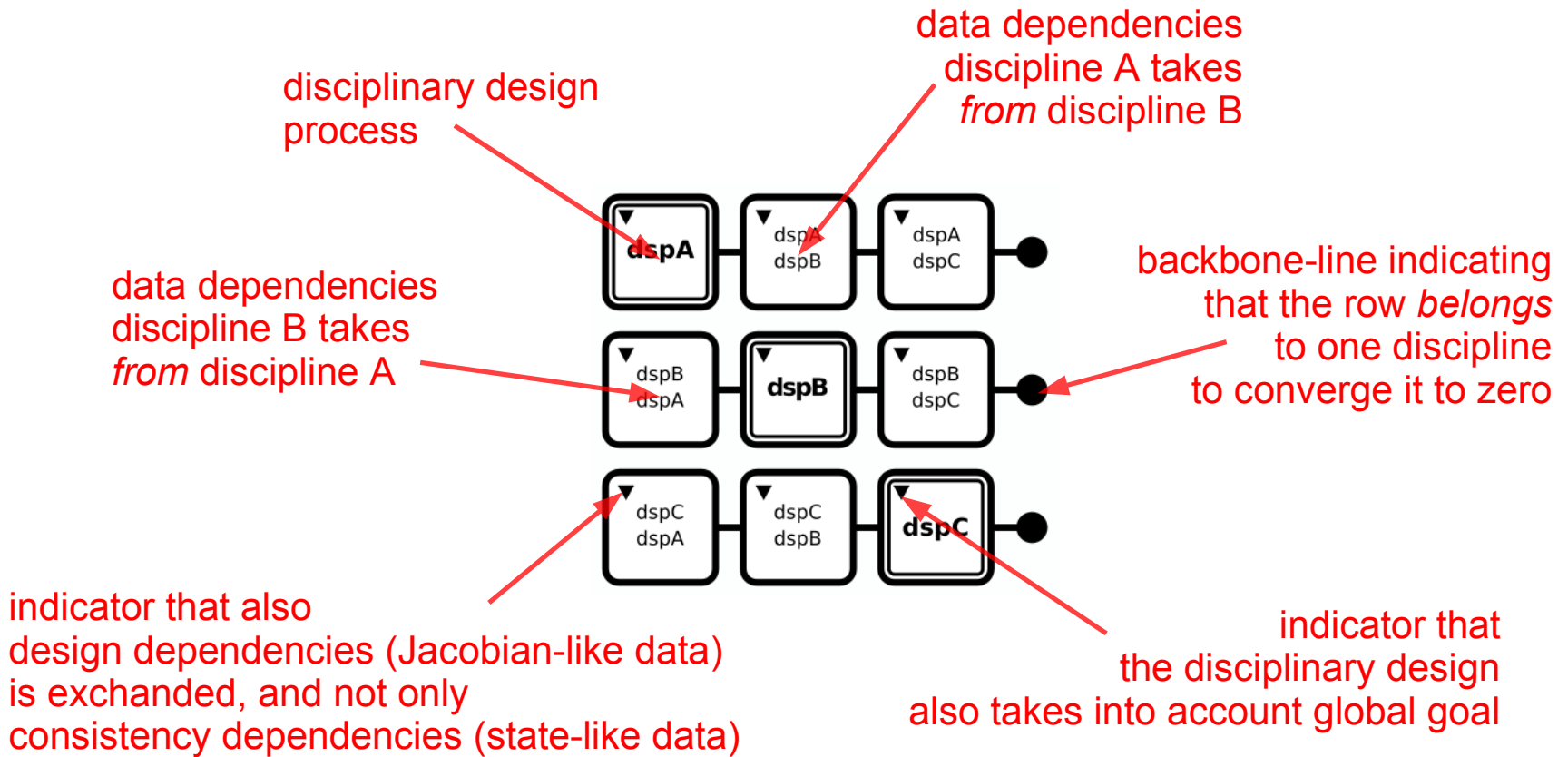
$$\frac{\widehat{\partial F}}{\widehat{\partial f_A}} \frac{\widehat{df_A}}{dp_B} + \frac{\widehat{\partial F}}{\widehat{\partial f_B}} \frac{\widehat{df_B}}{dp_B} + \frac{\widehat{\partial F}}{\widehat{\partial f_C}} \frac{\widehat{df_C}}{dp_B} - \lambda_A \frac{\widehat{dc_A}}{dp_B} - \lambda_B \frac{\widehat{dc_B}}{dp_B} - \lambda_C \frac{\widehat{dc_C}}{dp_B} = 0, \quad \underline{c_B} = 0, \quad \underline{r_B} = 0$$

$$\frac{\widehat{\partial F}}{\widehat{\partial f_A}} \frac{\widehat{df_A}}{dp_C} + \frac{\widehat{\partial F}}{\widehat{\partial f_B}} \frac{\widehat{df_B}}{dp_C} + \frac{\widehat{\partial F}}{\widehat{\partial f_C}} \frac{\widehat{df_C}}{dp_C} - \lambda_A \frac{\widehat{dc_A}}{dp_C} - \lambda_B \frac{\widehat{dc_B}}{dp_C} - \lambda_C \frac{\widehat{dc_C}}{dp_C} = 0, \quad \underline{c_C} = 0, \quad \underline{r_C} = 0$$



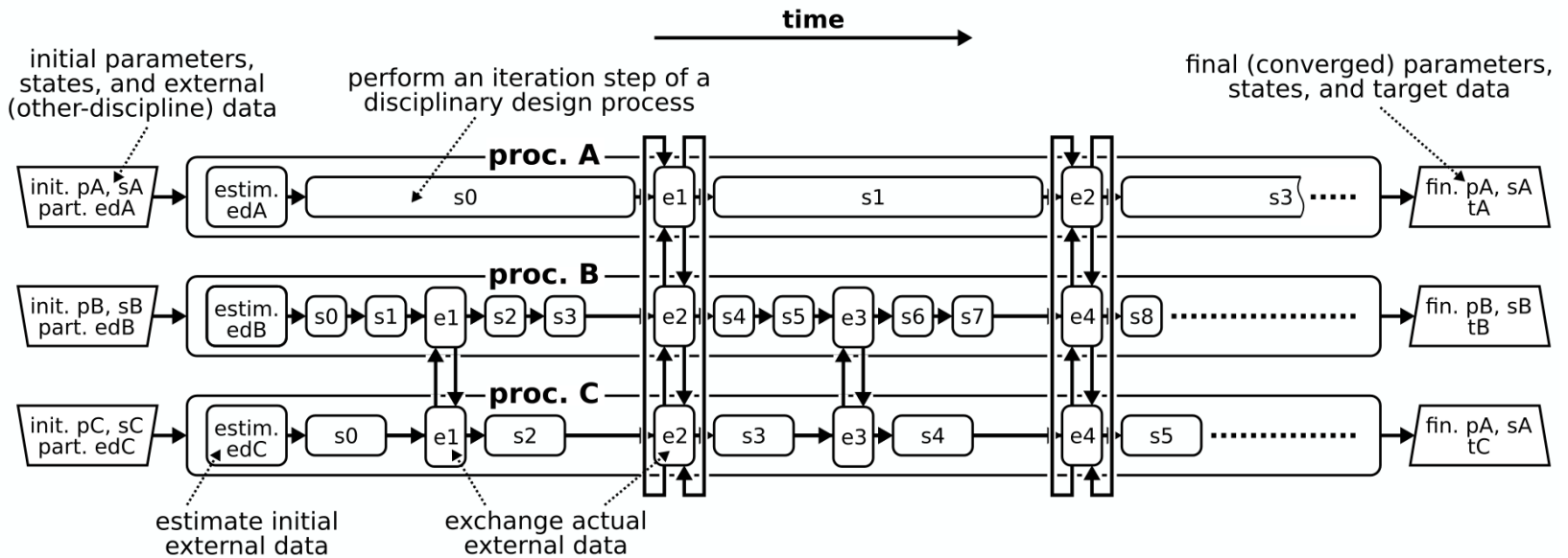
Cybermatrix representation

- Since the design equation is usually implied, use a schematic representation
- Let each row belong to one discipline (all related to its design parameters)



Communication protocol

- Each disciplinary design process can have any form, so long as **iterative**
- Equip it additionally with **data exchange points** and **initial data estimator**



- Different disciplines may have different exchange periods
- Selection of rows and exchange periods recover **all known MDO architectures**
- **In practical cases always a hybrid architecture**



A realization on HPC clusters

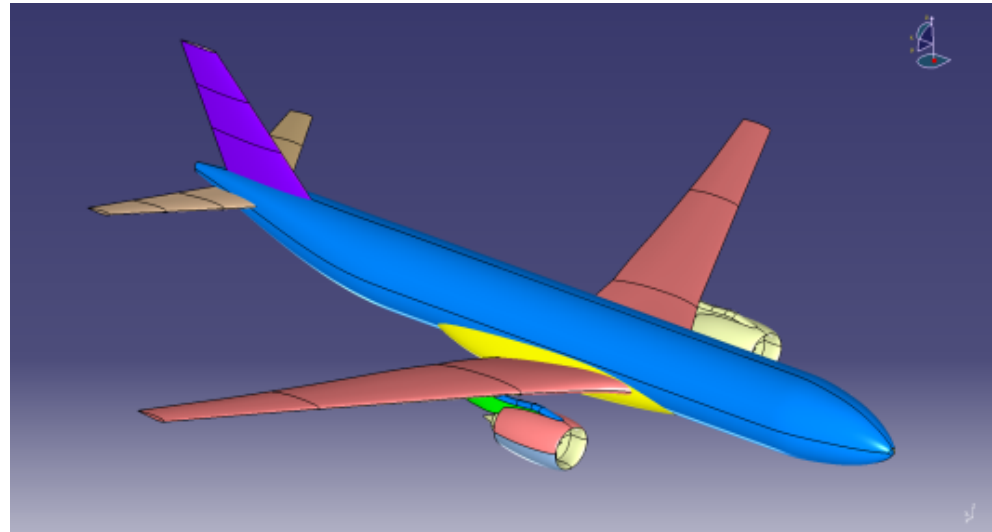
- A cybermatrix process integration framework for HPC clusters in development
 - Starts disciplinary processes, assigns resources, monitors progress
 - Triggers data exchanges and determines global convergence
- Disciplinary experts do not work with the framework directly
 - No need to learn yet another integration framework
 - Only provide **input collector** scripts to copy data from other disciplines
 - The whole MDO process implementation: a directory of input collectors
- Maintainable by standard software engineering tools and practices
 - Set of input collectors under source version control
 - Integration framework is an **interpreter** of the set of collectors and some meta-data (data exchange periods, etc)
- Currently data exchange performed over parallel on-disk file system
 - Parallel in-memory or area-network file system possible in principle
 - No changes to disciplinary processes in any case



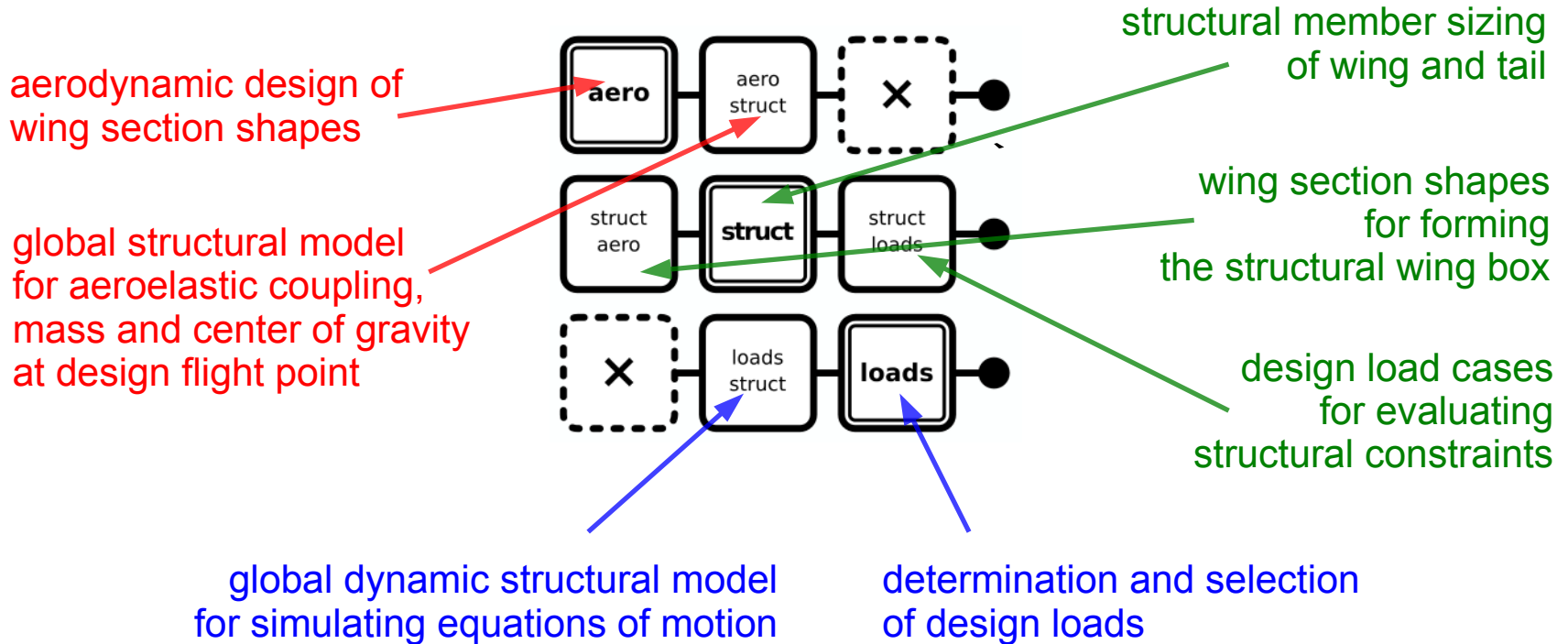
Example: MDO of long-range transport aircraft

- Configuration:
 - twin-engine wide-body
 - long-range transport aircraft
 - Wing-body-tail-pylon-flow through nacelle
 - 250 t max. take-off mass class
- Global goal function: minimize fuel consumption
- Constraints: all local (assigned to disciplines)

- Involved disciplinary processes:
 - Aerodynamic design of wing section shape (**aero**)
 - Structural member sizing of wing and tail (**struct**)
 - Determination and evaluation of design loads (**loads**)



Example: Matrix setup



Only consistency dependencies, no design dependencies



Example: Disciplinary subprocesses

➤ **aero:**

Adjoint gradient-based static-aeroelastic optimization

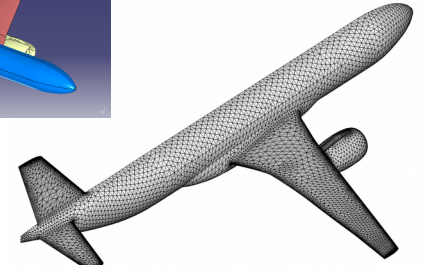
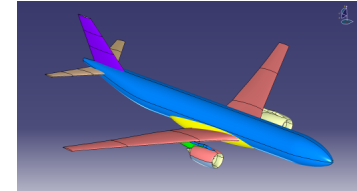
Design point: $M = 0.83$, $h = 11000$ m

CAD+ROM B-spline airfoil definition, 126 design parameters

Hybrid-unstructured RANS CFD mesh, 544,000 pts, 1,130,000 els

Goal: minimize drag; constraints: trimmed flight

Between data exchanges: one gradient evaluation and one line search



➤ **struct:**

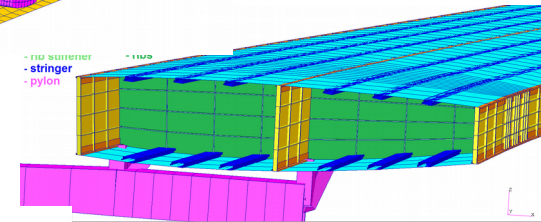
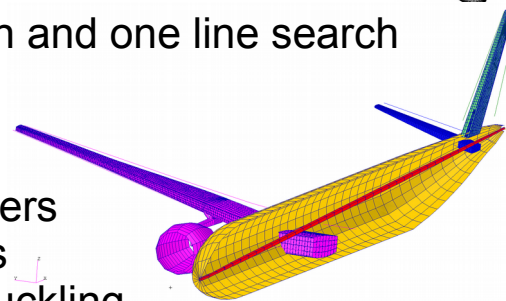
Gradient-based sizing of structural regions

Referent region thicknesses, 364 design parameters

Global FE model, 18,000 nodes, 42,000 elements

Goal: minimize mass; constraints: strength and buckling

Between data exchanges: one full sizing



➤ **loads:**

Transient dynamic simulations of gust and turbulence excitations

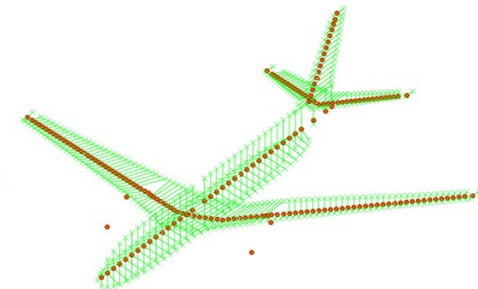
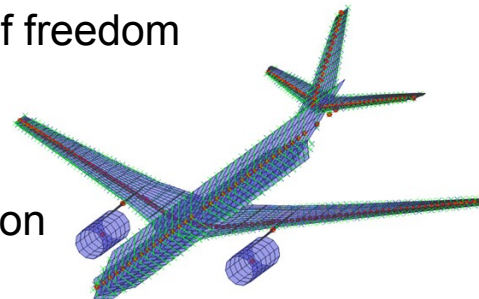
Dynamic structural model, 1068 degrees of freedom

Panel aerodynamic model, 1163 boxes

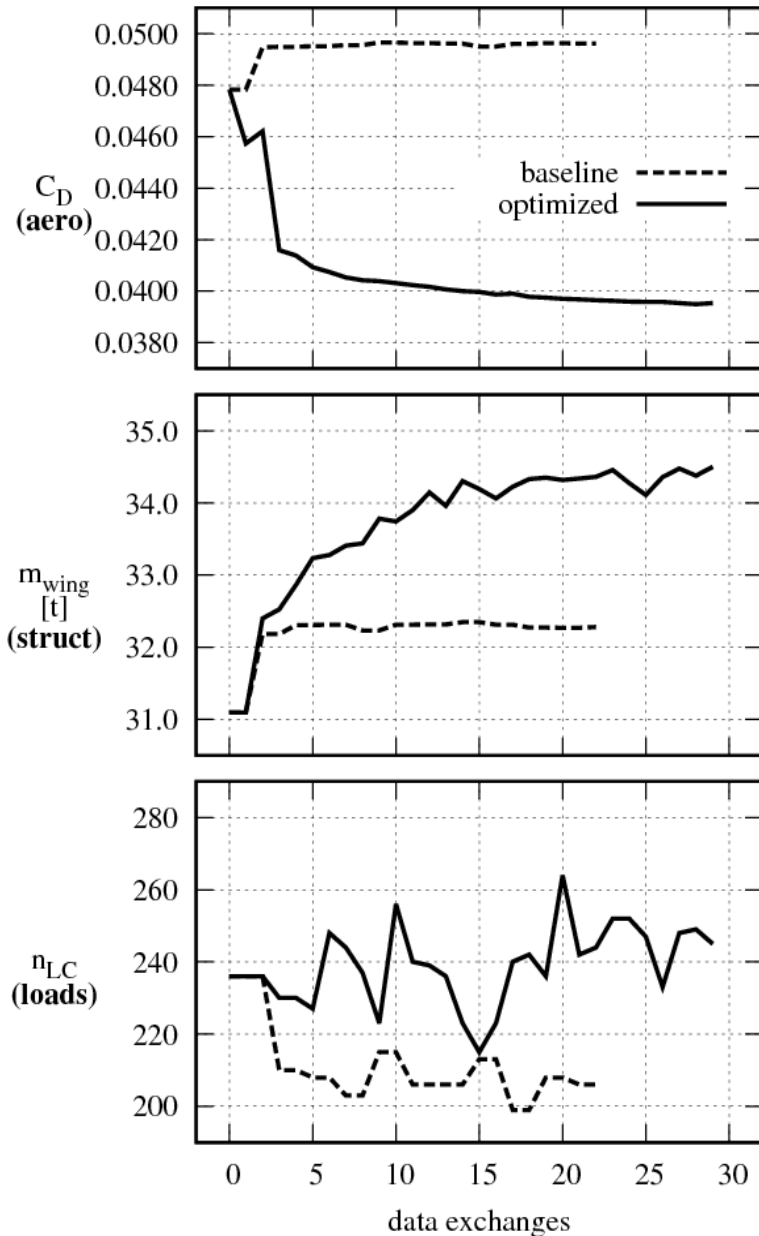
1284 load cases and 2 mass cases

No goal/constraints, no design parameters

Between data exchanges: one full evaluation



Example: Optimization result



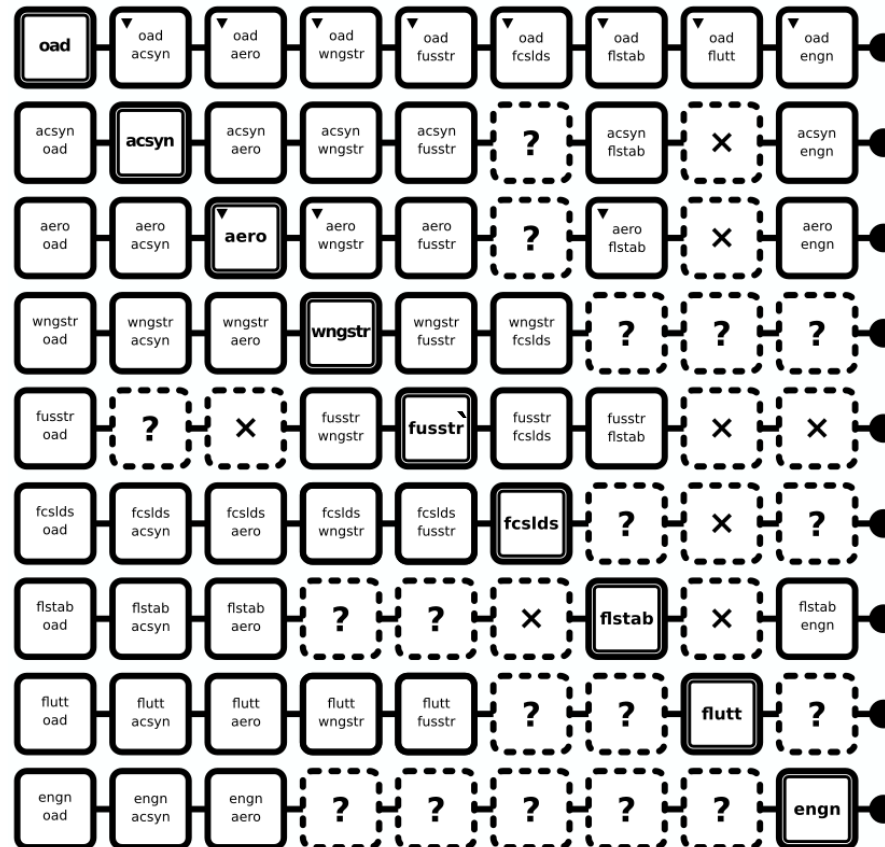
- Total run time: **110 hours on 64 cores**
 - Time between data exchanges: 4.1 h avg
 - Set by **aero** on 48 c; **struct** 2.2 h on 4 c; **loads** 1.2 h on 12 c
- Main effect: high drag reduction (**-16%**) for modest increase in mass (**4.6%** wing, **0.8%** total)
 - Wing sections slightly retwisted and reshaped to reduce shocks (drag level high due to coarse mesh)
 - Somewhat less favorable spanwise load distribution results in higher design loads
 - Variation in number of design load cases not large, but not negligible
- What is the baseline for comparison?
 - "0" on data exchange axis has no meaning
 - Intention-dependent: here result of an optimization with fixed aerodynamic des. par.
- Global goal function (fuel consumption) not explicitly considered in any discipline
 - A missing design dependency
 - Local goal functions may increase after data exchanges

Conclusions and Outlook

- A core of a cybermatrix-based MDO process demonstrated
 - Aero-structural approximate optimization with variable number of design load cases
 - CAD-based shape parametrization through reduced order modeling
 - Realistic loads process following certification regulations
- Improvement to the core process
 - More flight points and powerd engine for aerodynamic design
 - Control laws and high-fidelity corrections for loads
 - Design dependencies (Jacobian-like information)
- Beyond the core process
 - Higher fidelity structural modeling (separate wing/fuselage disciplines)
 - Tighter geometry and mass synthesis (aircraft synthesis discipline)
 - Modification of wing planform shape (overall aircraft design discipline)
 - Flutter analysis (eliminate planforms exhibiting inherent flutter)



Thank you for your attention!



this time
next year
(more-or-less)

