





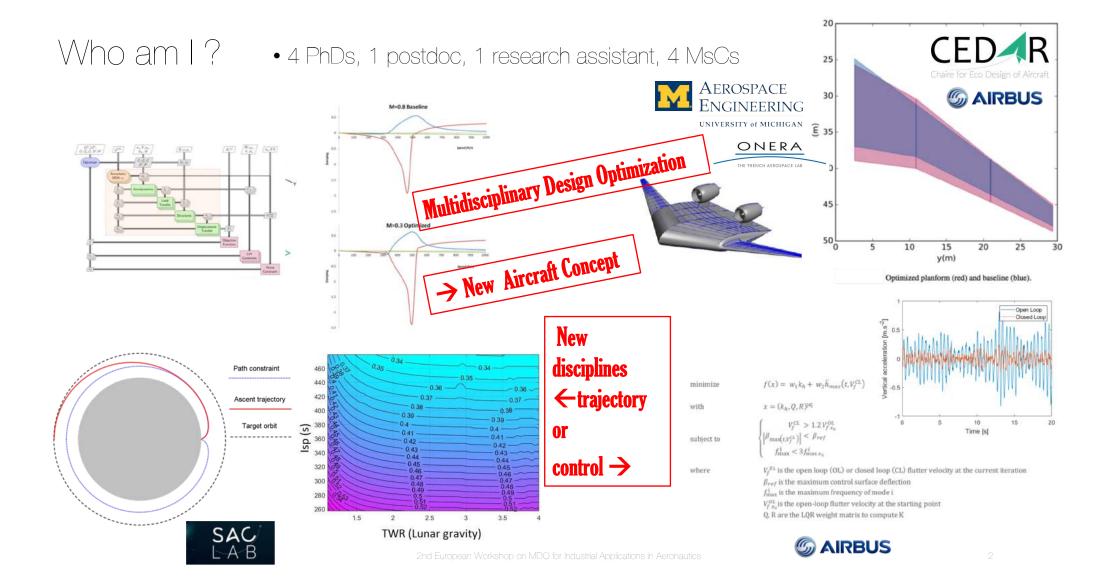
MDO in Academia: From classrooms to research

And vice versa

Nathalie Bartoli, Thierry Lefebvre, JOSEPH MORIER

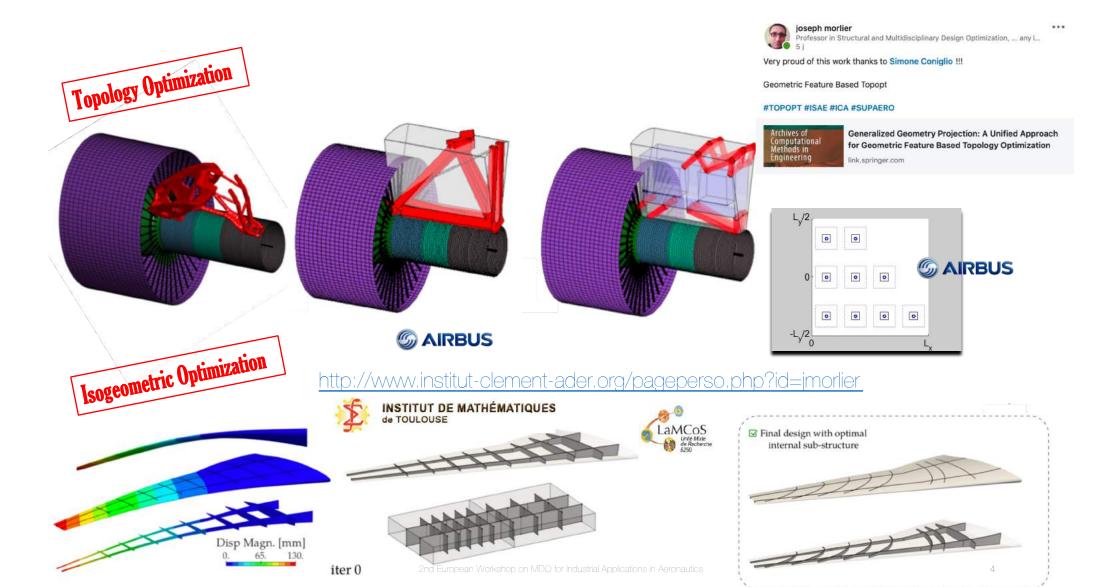




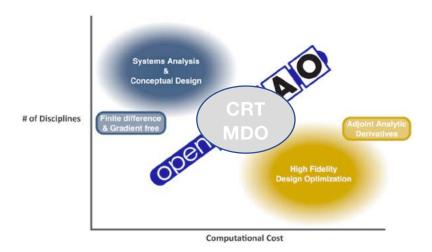


Structural Optimization can really change the Design, Isn't it?

Multidisciplinary Design Optimization



Common Research Team (ONERA+SUPAERO+...ENAC):



CRT 10+ researchers

Opensource tools

FBHALE provides a conceptual multidisciplinary design framework for high altitude long endurance (HALE)

Structural and Multidisciplinary Optimization https://doi.org/10.1007/s00158-019-02298-4

EDITORIAL



tribute

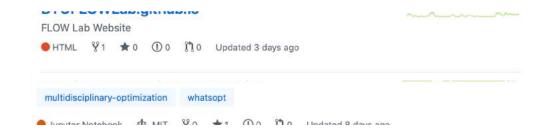
Replication of results

Raphael T. Haftka 1 0 · Ming Zhou 2 · Nestor V. Queipo 3

Received: 28 January 2019 / Revised: 18 April 2019 / Accepted: 30 April 2019

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Outlines for today

#EfficientGlobalOptimization

#Multifidelity

#Typeofdesignvariables

#Studentskills

- 1. Starting point X₀
 - 2. Examples
- 3. Synergy with SMT
- 4. Classrooms and add-ons
- 5. Conclusion and future works

Outlines for today

1. Starting point X₀

- 2. Examples
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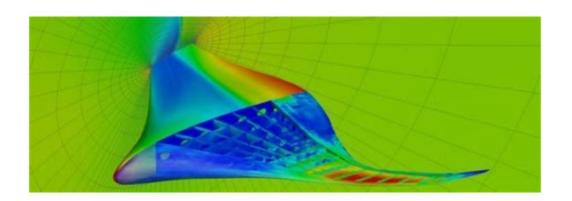
A long story short

• A French ANR project over the period 2005-2009 (OMD-RNTL titled as, Multi-Disciplinary Optimization coordinated by Rodolphe Leriche of the Engineering school "Mines de St Etienne") has allowed to federate the French researchers in computational mechanics and design optimization. (Most of our works done in SCILAB... the french free clone of MATLAB)

→...OpenMDAO appeared in 2010

- In 2015 ISAE-SUPAERO hosted Prof. Martins* as a visiting professor via an European grant (H2020-MSCA-IF-2014_ST)
- January 2015, MDA-MDO project launched at IRT Saint-Exupéry (GEMS framework)
- June 2015, The European H2020 project AGILE has contributed to the formalization of the MDO processes and the development of numerous opensource tools for aircraft design.
- In 2017, the MDOlab released the SMT python toolbox conjointly developed by NASA, University of Michigan UoM, ONERA and ISAE-SUPAERO.

Popularization for our common research ONERA-SUPAERO



http://mdolab.engin.umich.edu

https://www.linkedin.com/pulse/optimiz ation-mdo-connecting-people-joseph-

Optimization [MDO] for connecting people?

Publié le 14 février 2019



✓ Voir les stats



ioseph morlier Professor in Structural and Multidisciplinary Design Optimization, ... any idea? 2 articles







Outlines for today

1. Starting point X_0 ?

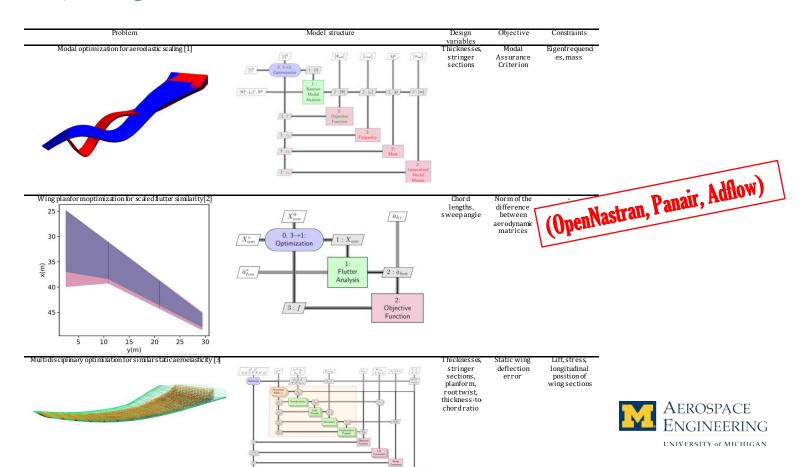
2.Examples

- 3. Synergy with SMT
- 4. Classrooms and add-ons
- 5. Conclusion and future works

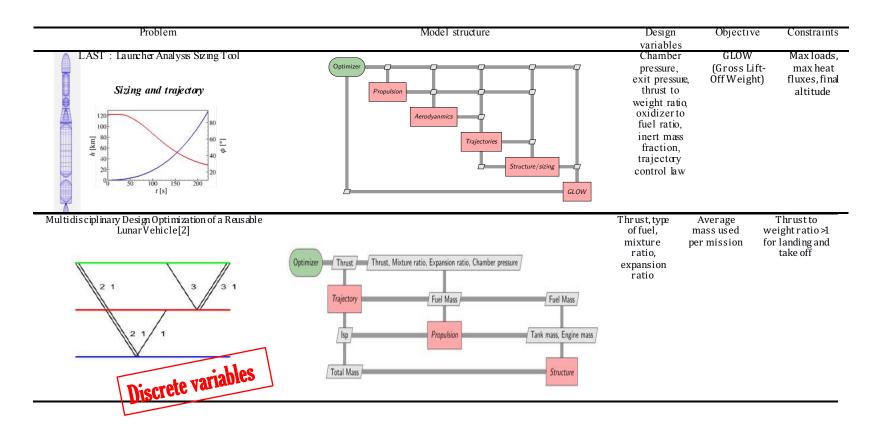
PhD's projects

https://github.com/mid2SUPAERO/aerostructures

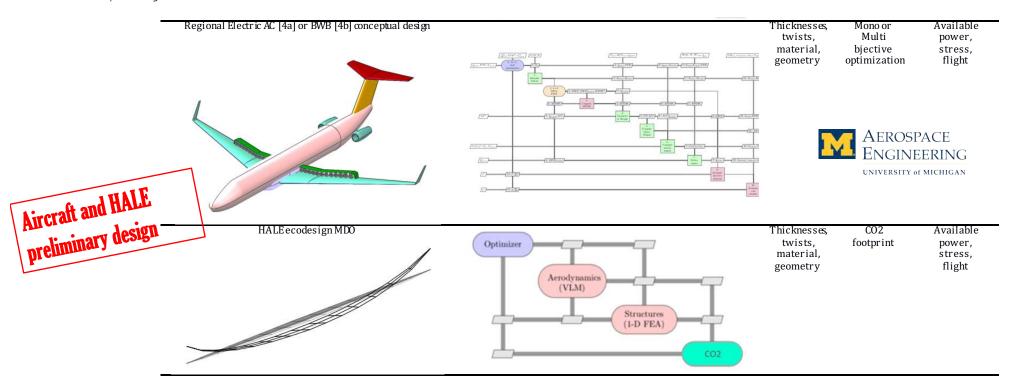
Aeroelasticity
Flying Demonstrator



And more recently... for Space Application at SUPAERO



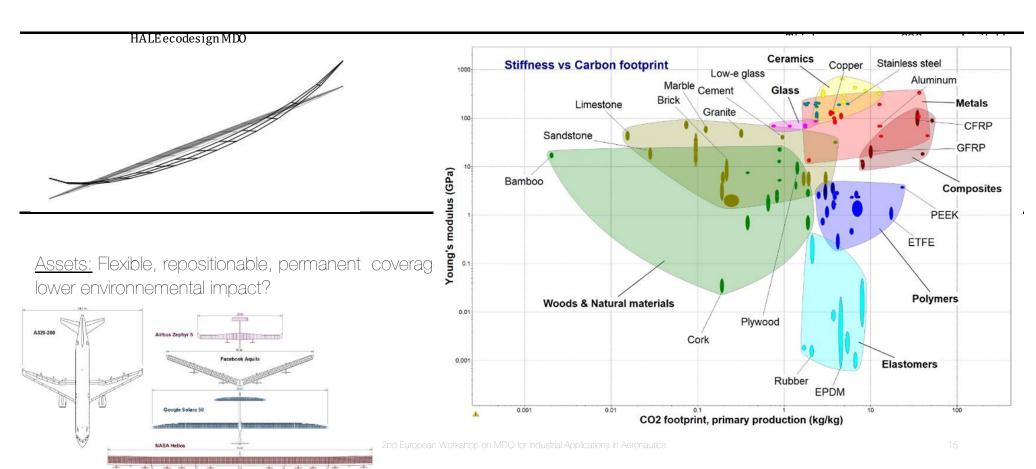
PhD's projects



ECO Hale

https://www.openscience.fr/Maitriser-la-conception-des-drones-solaires-a-voilure-souple-vers-l-avenement

Derive OAS 2.0 to treat a HALE pseudo sattelite Design problem



ECO MATERIAL SELECTION

Turn discrete problem of material selection into continuous inspired from SIMP

Zuo, W., & Saitou, K. (2016). Multi-material topology optimization using ordered SIMP interpolation. Structural and Multidisciplinary Optimization, 55(2), 477-491. doi:10.1007/s00158-016-1513-3

- Acces material properties through density (only one material design variable):
 E(p), G(p), oyield(p), CO2(p) (production CO2 footprint)
- Example for young modulus: Between two materials i and i+1,

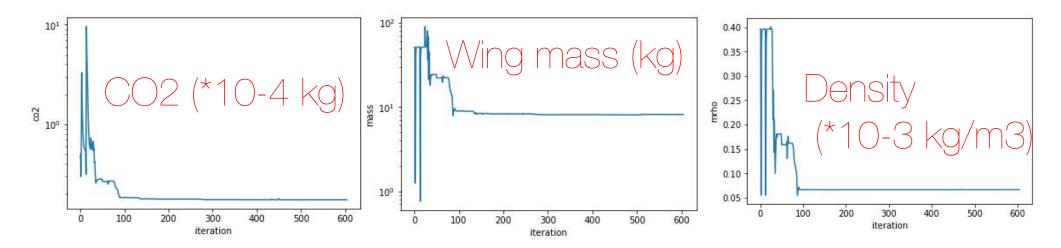
$$E_e(arrho_e)=A_E*arrho_e^p+B_E$$
 , with $A_E=rac{E_i-E_{i+1}}{arrho_i^p-arrho_{i+1}^p}$ and $B_E=E_i-A_E*arrho_i^p$

- Power p added during optimisation so that only real materials are optimal
- Inversed curvature for CO2 as smaller CO2 is more advantageous

Our design variables

Variable	Lb	Ub	XO (multistart)	unit
Skin thickness	0,0001	0,1	0,01-0,001	m
Spar thickness	0,0001	0,1	0,01-0,001	m
Wing span	1	1000	50	m
Wing chord	1	500	2	m
Wing taper	0,01	0,99	0,1	-
Wing thickness over chord ratio	0,01	0,20	0,1	-
Twist	-15	15	5	0
Density	50	8200	50-8200	Kg/m3

RESULTS

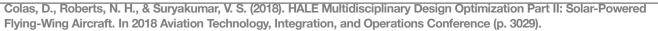


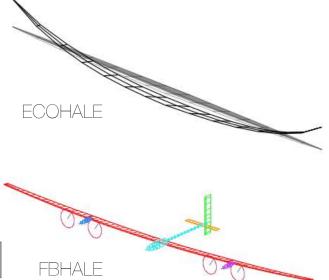
- The final material is a sandwich panel (UD CFRP expanded PS foam UD CFRP)
- The optimal material in terms of CO2 is very close to the optimal material in terms of weight, because battery is the most impacting on CO2

ECODESIGN IN THE MDO LOOP

- HALE could be a smart alternative to satellites (launching) => important to make them as clean as possible.
- Our method of material selection can be adapted to any aerostructure. Still working on multimaterials

Variable	ECO HALE OpenAeroStruct	FBHALE
Span (m)	48	45
Chord (m)	1,04	1,6
Total mass (kg)	107	320
Battery+PV mass (kg + %total)	54(50%)	170 (53%)
Payload+avionics mass (kg + %total)	20,5 (19%)	28 (9%)
Wing Structure mass (kg + %total)	30 (28%)	67 (21%)





Outlines for today

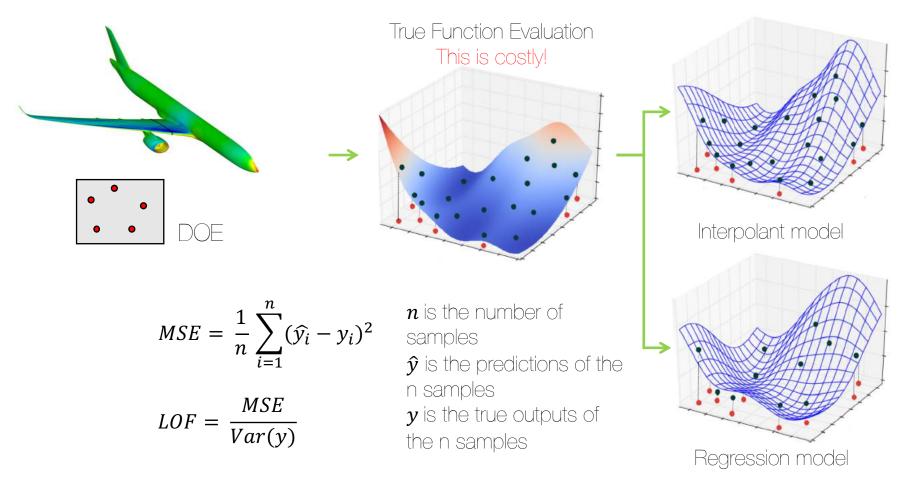
- 1. Starting point X₀
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3. Synergy with SMT

4. Courses and add-ons

5. Conclusion and future works

Surrogate modeling Recipes



3/ Choose a Kernel/Construct Kxx Matrix view of Gaussian Process and Hyperparameters tuning 1/ Get your inputs/outputs data 2/ You wan to predict at x* $k(x, x') = \frac{\theta_1^2 \exp(-\frac{(x - x')^2}{2\theta_2^2})}{\|x\|^2}$ = $[K \times \times]$ $[K \times X]$ and variance of estimate $m(x_*) = K_*[Kxx]^{-1}y$ $var(x_*, x_*') = K_{**} - K_*^T [Kxx]^{-1} K_*$ 4/ compute mean

«Costly code» reduction with GP





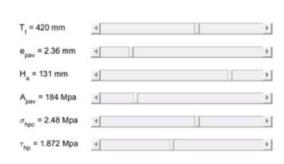
joseph morlier

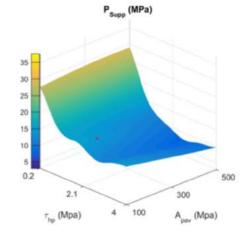
Professor in Structural and Multidisciplinary Design Optimization, ... any i... 2 i

#ML

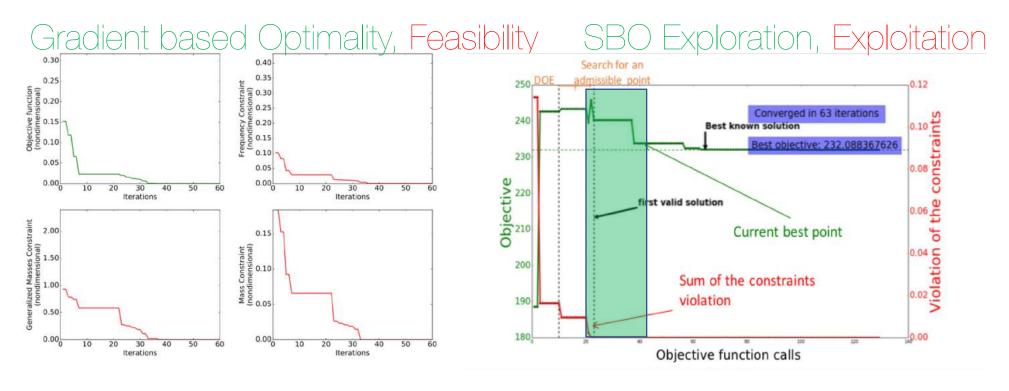
Have a look to one of our 2018 paper, where Machine Learning or Surrogate modelling technics help to understand Complex mechanical behaviour (impact on sandwich shield)

https://lnkd.in/dr_WSqA





New graphs

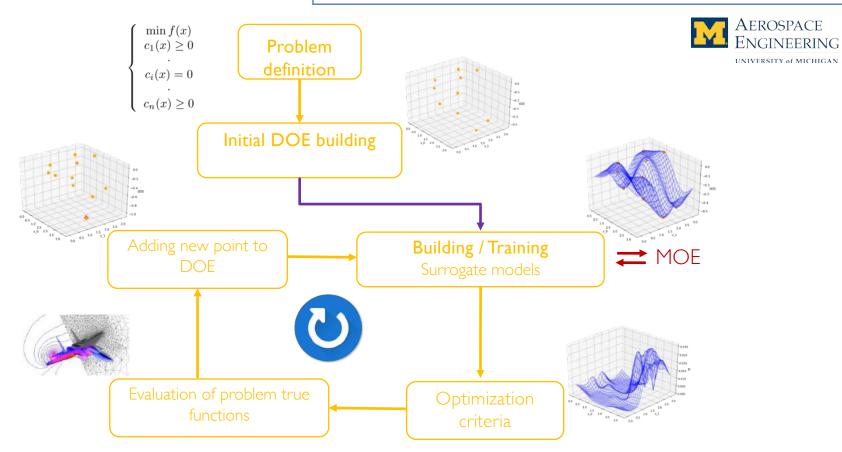


Stopping criteria: tolfun, tolx, maxiter

Stopping criteria: Max Budget (Function calls)

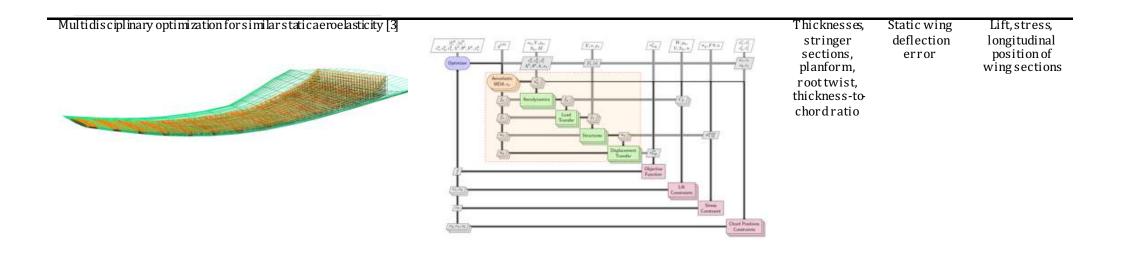
SEGOMOE algorithm

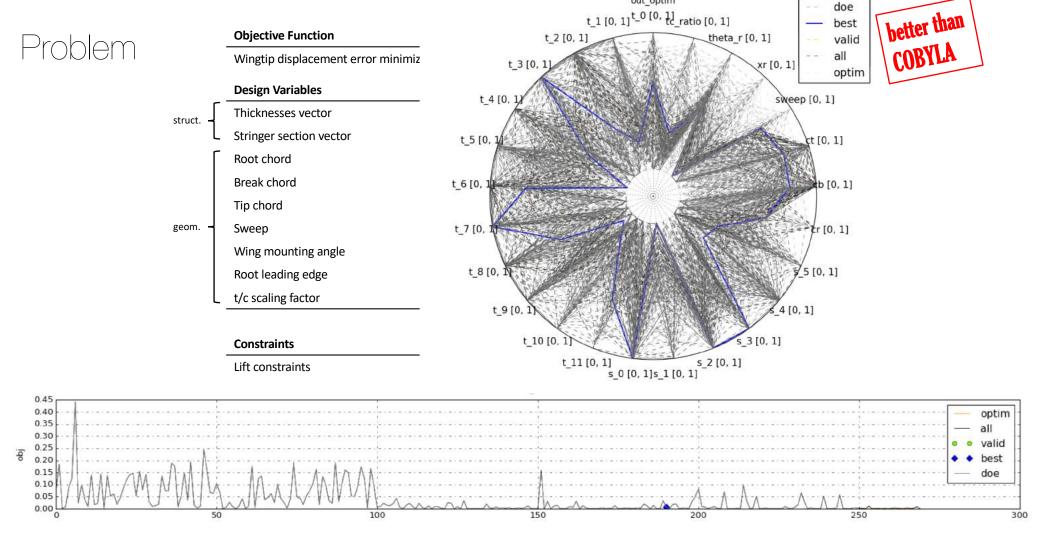
Bartoli, N., Lefebvre, T., Dubreuil, S., Olivanti, R., Priem, R., Bons, N, Martins, J. R. & Morlier, J. (2019). Adaptive modeling strategy for constrained global optimization with application to aerodynamic wing design. Aerospace Science and technology, 90, 85-1



UNIVERSITY of MICHIGAN

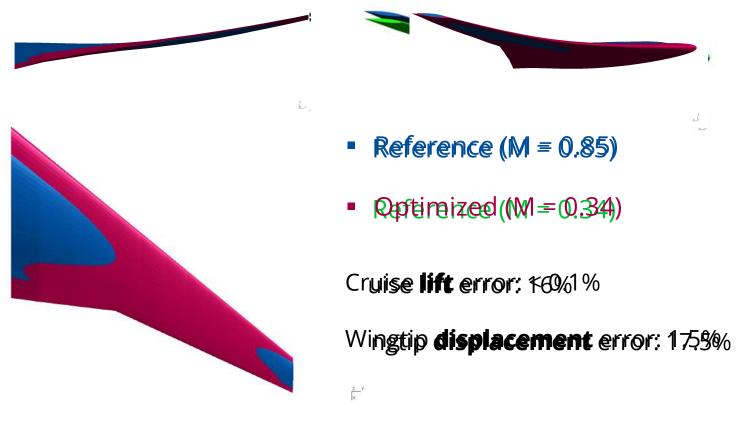
Static Optimization (SEGOMOE)





out_optim

RESULTS



Surrogate Model Toolbox: SMT https://github.com/SMTorg/S



Mohamed **Bouhlel**

Table of Contents

SMT: Surrogate Modeling Toolbox Cite us Focus on derivatives **Documentation contents** Indices and tables

Next topic

Getting started

This Page

Show Source

SMT: Surrogate Modeling Toolbox

The surrogate modeling toolbox (SMT) is an open-source Python package consisting of libraries of surrogate modeling methods (e.g., radial basis functions, kriging), sampling methods, and benchmarking problems. SMT is designed to make it easy for developers to implement new surrogate models in a well-tested and well-document platform, and for users to have a library of surrogate modeling methods with which to use and compare methods.

The code is available open-source on GitHub.

Cite us

To cite SMT: M. A. Bouhlel and J. T. Hwang and N. Bartoli and R. Lafage and J. Morlier and J. R. R. A. Martins. A Python surrogate modeling framework with derivatives. Advances in Engineering Software, 2019.

```
@article{SMT2019,
       Author = {Mohamed Amine Bouhlel and John T. Hwang and Nathalie Bartoli and Rémi Lafage
        Journal = {Advances in Engineering Software},
       Title = {A Python surrogate modeling framework with derivatives},
       pages = {102662},
        year = {2019},
        issn = \{0965 - 9978\},
       doi = {https://doi.org/10.1016/j.advengeoft.2019.03.005},
       Year = {2019}}
```

Focus on derivatives

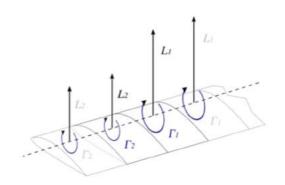
M.-A. Bouhlel, J. T. Hwang, N. Bartoli, R. Lafage, J. Morlier, J. R.R.A Martins (2019), A Python surrogate modeling framework with derivatives, Advances in Engineering **Software**

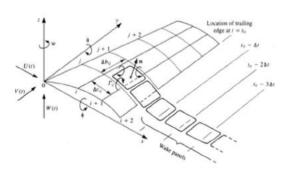
KPLS, RMTS, GEK

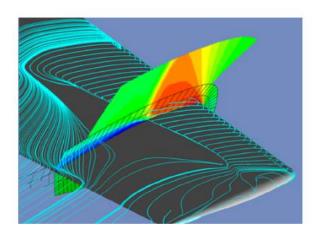
What if?

Several levels of fidelity of the same simulation are available

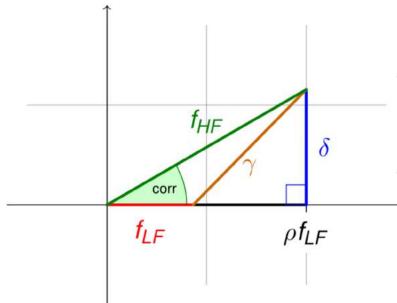
→ For example, in aerodynamics: Liflting line theory, Vortex lattice method, and RANS CFD code







Co Kriging



Additive formulation [Lewis 2000]

$$f_{HF}(x) = f_{LF}(x) + \gamma(x)$$

- Kennedy-O'Hagan [Kennedy 2001]

$$\begin{cases} f_{HF}(x) = \rho f_{LF}(x) + \delta(x) \\ f_{LF}(\cdot) \perp \delta(\cdot) \end{cases}$$

The addition of the term p makes the multi-fidelity learning more robust to poor correlation as well as differences in modelisation.

^{\$}Alexandrov, N., Lewis, R., Gumbert, C., Green, L., & Newman, P. (2000, January). Optimization with variable-fidelity models applied to wing design. In 38th Aerospace Sciences Meeting and Exhibit (p. 841).

Kennedy, M. C., & O'Hagan, A. (2001). Bayesian calibration of computer models. Journal of the Royal Statistical Society: Series B (Statistical Methodology), 63(3), 425-464.

Lam, R., Allaire, D. L., & Willcox, K. E. (2015). Multifidelity optimization using statistical surrogate modeling for non-hierarchical information sources. In 56th AIAA/ASCE/AHS/ASC Structures, Structural Dynamics, and Materials Conference (p. 0143).

→ It is also away to learn the difference between HF & LF ...

2nd European Workshop on MDO for Industrial Applications in Aeropautics

MFEGO 2 steps approach

• Most promising point: El-based criterion

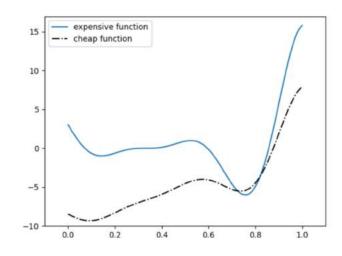
$$\mathbf{x}^* = \arg\max_{\mathbf{x}} \left(\mathrm{EI}(\mathbf{x}) \right)$$

• Choice of levels of enrichment: trade off information gain/cost

$$\mathbf{k}^* = \arg\max_{k \in (0, \dots, \ell)} \quad \frac{\sigma_{\mathrm{red}}^2(k, \mathbf{x}^*)}{\mathrm{cost}_{\mathrm{total}}(k)^2}$$



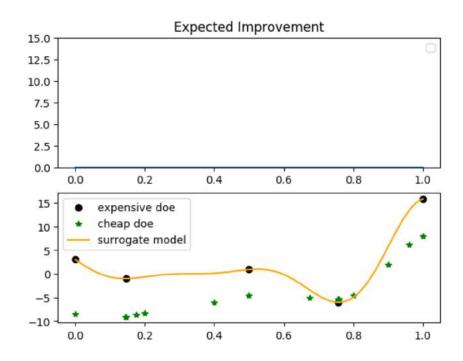
⇒ High-fidelity is used for Exploitation and model enhancement



$$f_{HF}(x) = (6x - 2)^2 \times \sin(2(6x - 2))$$

 $f_{LF}(x) = 0.5f_{HF} + 10(x - 0.5) - 5$

Results (Toy problem)

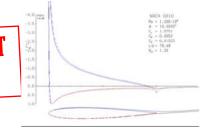


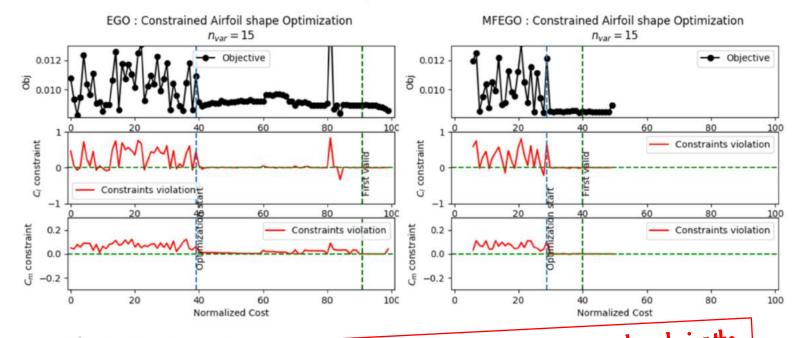
Cost ratio: 1/1000						
2	HF	LF	Cost			
MFEGO	3+2	6+9	5.015			
EGO	4+11	-	15			

Constrained Optimization



Estimated COST RATIO: 1/200





*https://web.mit.edu/drela/Public/web/xfoil/ \$ http://mdolab.engin.umich.edu MFEGO can speed up the Optimization process by reducing the calls to HF expensive code!

Conclusions on SMT

- « Reducing » industrial (**&costly**) simulation code is interesting to exchange data (without having access to the code) in a collaborative project (see AGILE...).
- Given its focus on **derivatives**, SMT is synergistic with the OpenMDAO framework. It can provide the derivatives that OpenMDAO requires from its components to compute the coupled derivatives of the multidisciplinary model.
- OpenMDAO/SMT is a natural framework for Bayesian/Surrogate based Optimization/Multifideility (Low DV number !)
- SMT core capabilities (KPLS) has been adapted for Surrogate based Optimization for mixed variables in OpenMDAO
- → See examples from Roy et al 2019

Roy, S., Crossley, W. A., Moore, K. T., Gray, J. S., & Martins, J. R. (2019). Monolithic Approach for Next-Generation Aircraft Design Considering Airline Operations and Economics. Journal of Aircraft, 56(4), 1565-1576.

Recent Papers on this topic



Bouhlel, M. A., Bartoli, N., Otsmane, A., & Morlier, J. (2016). Improving kriging surrogates of high-dimensional design models by Partial Least Squares dimension reduction. Structural and Multidisciplinary Optimization, 53(5), 935-952.

Bouhlel, M. A., Bartoli, N., Otsmane, A., & Morlier, J. (2016). An improved approach for estimating the hyperparameters of the kriging model for high-dimensional problems through the partial least squares method. Mathematical Problems in Engineering, 2016.



Bouhlel, M., Bartoli, N., Regis, R. G., Otsmane, A., & Morlier, J. (2018). Efficient global optimization for high-dimensional constrained problems by using the Kriging models combined with the partial least squares method. Engineering Optimization, 1-16





Bartoli, N., Lefebvre, T., Dubreuil, S., Olivanti, R., Priem, R., Bons, N, Martins, J. R. & Morlier, J. (2019). Adaptive modeling strategy for constrained global optimization with application to aerodynamic wing design. Aerospace Science and technology, 90, 85-102.

Bartoli, N., Meliani, M., Morlier, J., Lefebvre, T., Bouhlel, M. A., & Martins, J. (2019). Multi-fidelity efficient global optimization: Methodology and application to airfoil shape design. In AIAA Aviation 2019 Forum (p. 3236).







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 - 2. Examples
- 3. Synergy with SMT

4. Classrooms and add-ons

5. Conclusion and future works

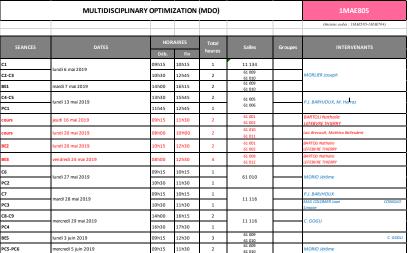
MDO courses & seminars



Thanks to
Imco Van
Gent,
TU Delft

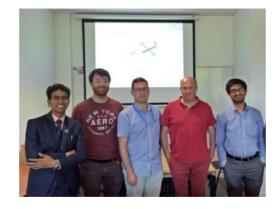






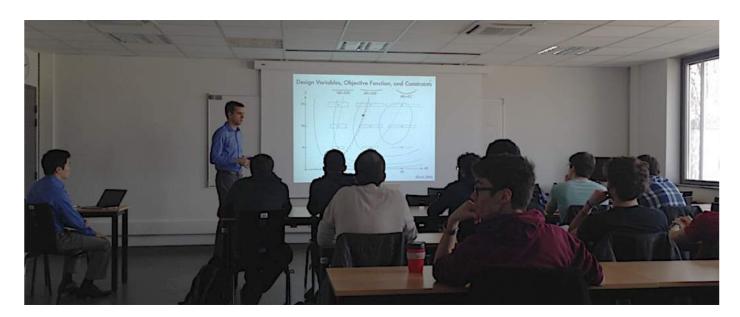
Sensitivity of finite element code
Continuous optimization (local/global)
Structural and topology Optimization
Response surface methods /DOE/SMT
Monte Carlo methods
Uncertainty propagation
Variance reduction
Bayesian Optimization
Reduced Order Modeling
MDA

Introduction to MDO+OpenMDAO (9h)



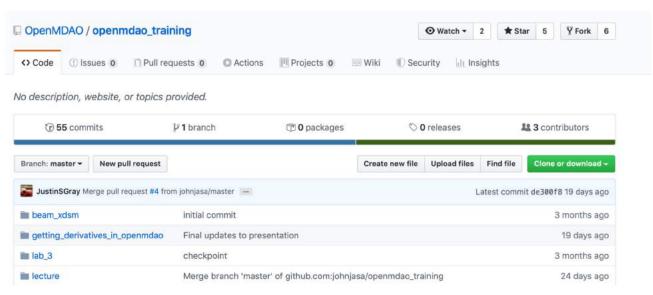
Since 2014 ... OpenMDAO course at SUPAERO

Supaero engineering 's program (15+ students) since 2014 Master MAE (25+ students) since 2016



Why OpenMDAO?

- Exists since 2010
- Stable since V2.0
- NASA team support (5+)
- Collaboration with UoM, Purdue

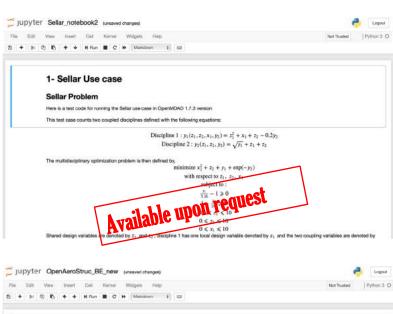


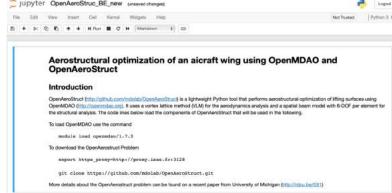
Jupyter Notebooks*

- Sellar
- SSBJ
- OAS
- Launcher design project based on FELIN https://github.com/M2CI-ONERA/FELIN

Short Courses for PhDs:

- Nice documentation for OM to start)!
- Coupling with DYMOS is really nteresting for us





Project-based learning: AGILE ACADEMY





S U P A E R O

WITH

ABOUT ISAE-SUPAERO

ACADEMICS

ADMISSION

from europe and north america.

In order to check if the AGILE paradigm was usable outside of the project, the AGILE partners create the AGILE Academy challenge. The aim was to perform the job of an aircraft design to AGILE tools. 36 students from 4 continents enter the challenge. They were divided into three February to july 2018, the three teams design the aircraft with the help of the partners.

At the end of the challenge, the winner team, composed by Andrew Jerayaj (Concordia University), Eznil Shakti Murugesan (Concordia University), marco picchi se University), Florian Sanchez (Bombardier and Concordia University), Karim Abusalem (Pisa University), Florian Sanchez (Bombardier and Concordia University), Karim Abusalem (Pisa University), Ronard (Pisa University), Ronard (Pisa University), Marine Ben Salah Bouhlel (Michigan University), designed a regional aircraft improved by so power sources.

They were invited to the AGILE Meeting in Naples to present their Work to all the agile Partne



https://www.agile-project.eu/agile-challenge





MID₂

Multidisciplinary optimization for Innovation: Design and Data

SUPAERO

Repositories 38

Packages









ONERA

THE FRENCH AEROSPACE LAB









Prof. J. Morlier joseph.morlier@isae-supaero.fr

40+ Students In MDO Courses At Master Level

In 2019 We Developed New Methodologies (or Applications) In:

Computational Structural Mechanics

#Gaussian Processes For Linear Elasticity

#Geometric Projection In Topology Optimization (AIRBUS and ICA)

#Topology Optimization For 3Dprinting (AIRBUS and ICA)

#High Resolution Topology Optimization (AIRBUS and ICA)

#Levet Set For Automatic Fiber Placement

#Eco Material Selection

#1D Refined FE Model In Dynamics

#IsoGeometric Analysis (LAMCOS and IMT)

Multidisciplinary Design Optimization

#HALE Ecodesign (CEDAR Chair) #Reusable Launchers (SACLAB Chair)

#Multifidelity Method with Gaussian Processes (ONERA and MDOlab)

#MDA Acceleration

#Codesign For Robust Flutter (AIRBUS)

#Trajectory Control

#Gaussian Processes and POD for coupled problem (ONERA)

#Aeroelastic for Scaled Aircraft (CEDAR Chair, ONERA and MDOlab)

#Hybrid Optimization (AIRBUS And IRT)

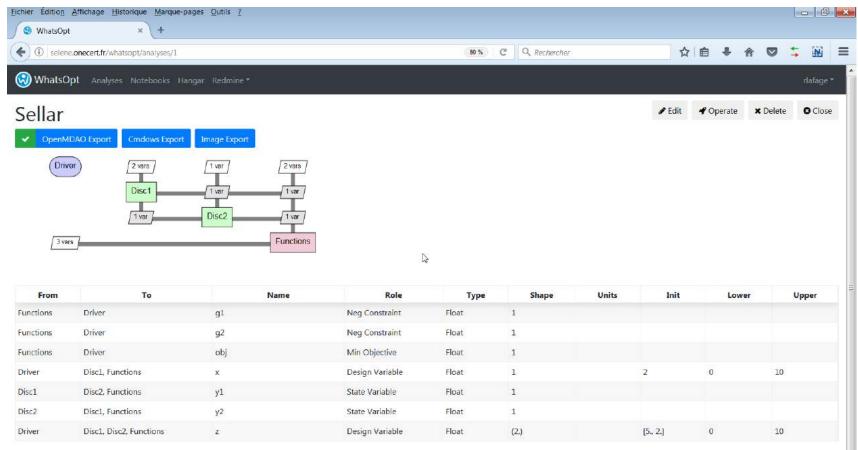
#BWB (CEDAR Chair, ONERA)

Thanks To All Supaero's Students (MAE, PIR, PhDs And Postdoc)

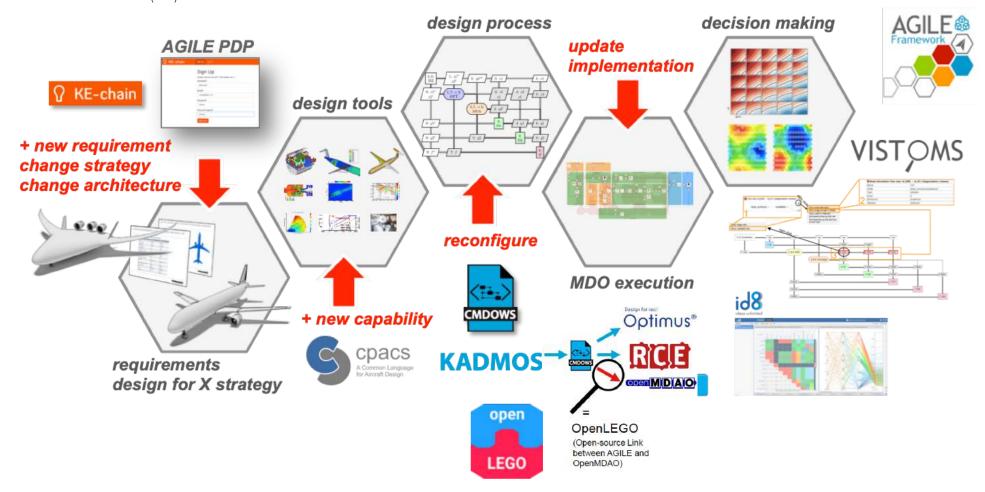


https://smt.readthedocs.io/en/latest/

ADD-ONS (1) https://github.com/OneraHub/WhatsOpt



ADD-ONS (2) http://www.agile-project.eu



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Conclusions

 MDO is the core of our Air/Craft Design researches at ONERA and SUPAERO→ SBO can facilitate the exploration of new concepts

EGO on SMT

• The multifidelity / Mixture of experts (MOE) options help us to speed the process (ongoing work)



- MDO concepts and technologies disseminated from university to aerospace industry
- → New generation of engineers with MDO skills
- Open questions:
- → Optimizer for hybrid design variables (continuous, discrete, categorical ...)?
- →Link with Manufacturing and MBSE

January 11th, 2019 - The follow-on project "AGILE 4.0" has been accepted by the European Commission.



AGILE 4.0 will start around mid-2019 and will extend the outcomes of the AGILE project to cover all the aspects of the development of complex aeronautical systems, including design, certification and manufacturing.

On Thursday MDO PhD day

...and in July 2020

```
# mixed continuous integer design variables
# optimal trajectory control
# {vehicle + mission} design
# Codesign ...
```

Minisymposium about Multifidelity Optimization during the joint WCCM/ECCOMAS events. (reference MS442 Multifidelity Optimization https://www.wccm-eccomas2020.org/frontal/MSList.asp)

The conference will take place in Paris (19th to 24th of July 2020). More details about the conference is available on the website

https://www.wccm-eccomas2020.org/frontal/introduction.asp



MDO PhD day - 21 November 2019 - ISAE-SUPAERO

Time	Session	Title	Presenter
08:30 - 09:00	Arrival & Coffee		2
09:00 - 09:10	Welcome & Introduction		Grégoire Casalis (ISAE – SUPAERO)
09:10 - 09:30	MDO Applications Chair: T. Lefebvre	"Exploration of a Blended Wing-Body concept featuring distributed electric propulsion with gradient optimization techniques"	Alessandro Sgueglia (ONERA / ISAE - SUPAERO)
09:30 - 09:50		"Multidisciplinary Design Optimization of a Reusable Lunar Vehicle "	Laurent Beauregard (ICA / ISAE - SUPAERO)
09:50- 10:10		"MDO for representative sub-scale flight Testing"	Akshay Raju Kulkarni Anne-Liza Bruggemai (TUD)
10:10 - 10:30		"Tool path optimization for free form surfaces machining"	Mahfoud Herraz (ENAC/ICA)
10:30 - 10:50		Coffee Break	K 1025432 (C. 6843)
10:50 - 11:10	MDO Aero-structure focus Chair: J. Morlier	"NeOPT: structural optimization in the conceptual design phase"	Francesco Toffol (POLIMI)
11:10 - 11:30		"HALE multidisciplinary design optimization with focus on Eco-Material selection"	Edouard Duriez (ISAE - SUPAERO)
11;30 - 11:50		"Transonic flight and movables loads modelling for preliminary aeroservoelastic MDO of a wing box"	Paul Lancelot (TUD)
11:50 - 12:10		"Some computational aspects of active flutter suppression for co-design"	Emeline Faisse (ISAE – SUPAERO / AIRBUS)
12:10 - 12:30		"High-fidelity aerodynamic shape optimization of wind turbine blades"	Mads Holst Aagaard Madsen (DTU)
12:30 - 14:00	Lunch & Coffee		1 1000000
14:00 - 14:20	MDO Methods Chair: Y. Diouane	"Recent MDO Lab activities"	Neil Wu (UM - MDO Lab)
14:20 - 14:40		"Optimizing System Architectures by Leveraging Collaborative MDO"	Jasper Bussemaker (DLR)
14:40 - 15:00		"Inclusion of discrete technological choices in design optimization process within a Bayesian framework"	Julien Pelamatti (ONERA)
15:00 - 15:20		"On a feasibility criterion for aircraft efficient global multidisciplinary optimization"	Rémy Priem (ONERA / ISAE - SUPAERO)
15:20 - 15:40		"Robust multidisciplinary design optimization applied to long range aircraft under operational uncertainty"	Nicolas Peteilh (ENAC)
15:40 - 16:00	Conclusion J. Morlier (ISAE – SUPAERO)		
16:00	Coffee Break		
	End of the Day		

NEW CHALLENGE (0) with IRT SE for Airbus

Structural Multidisciplinary Optimization manuscript No. (will be inserted by the editor)

Continuous



e.g.: area, thickness

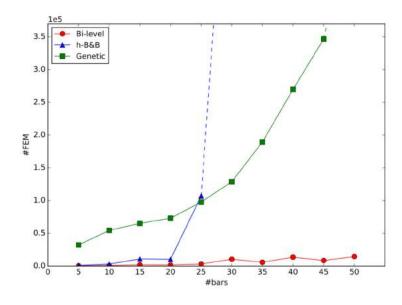
Non-ordered categorical



e.g.: cross-section, material

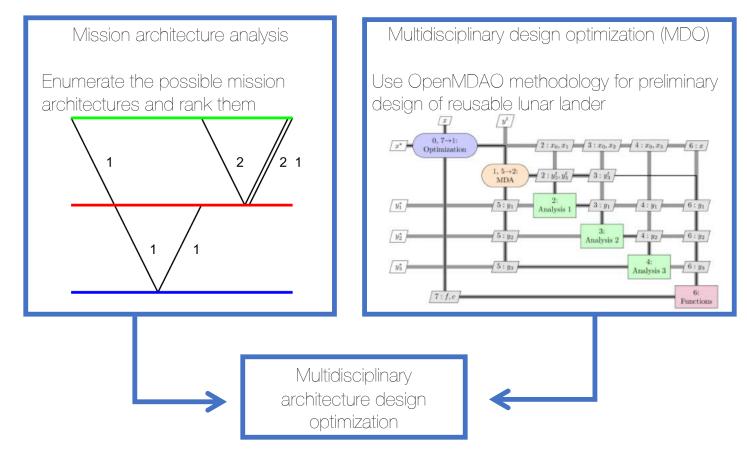
A bi-level scalable methodology for mixed categoricalcontinuous structural optimization problems

Pierre-Jean Barjhoux · Youssef Diouane · Stéphane Grihon · Dimitri Bettebghor · Joseph Morlier



NEW CHALLENGE (1) with AIRBUS DS/ Ariane Group



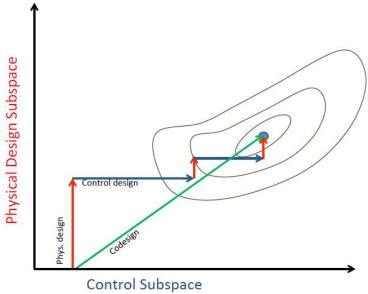


NEW CHALLENGE (2) with AIRBUS (Aircraft)

Navigate in physical and control design subspaces simultaneously.



Tailor structural/mechanical/control system designs: system optimality



Deshmukh, A. P., & Allison, J. T. (2016). Multidisciplinary dynamic optimization of horizontal axis wind turbine design. Structural and Multidisciplinary Optimization, 53(1), 15-27.

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