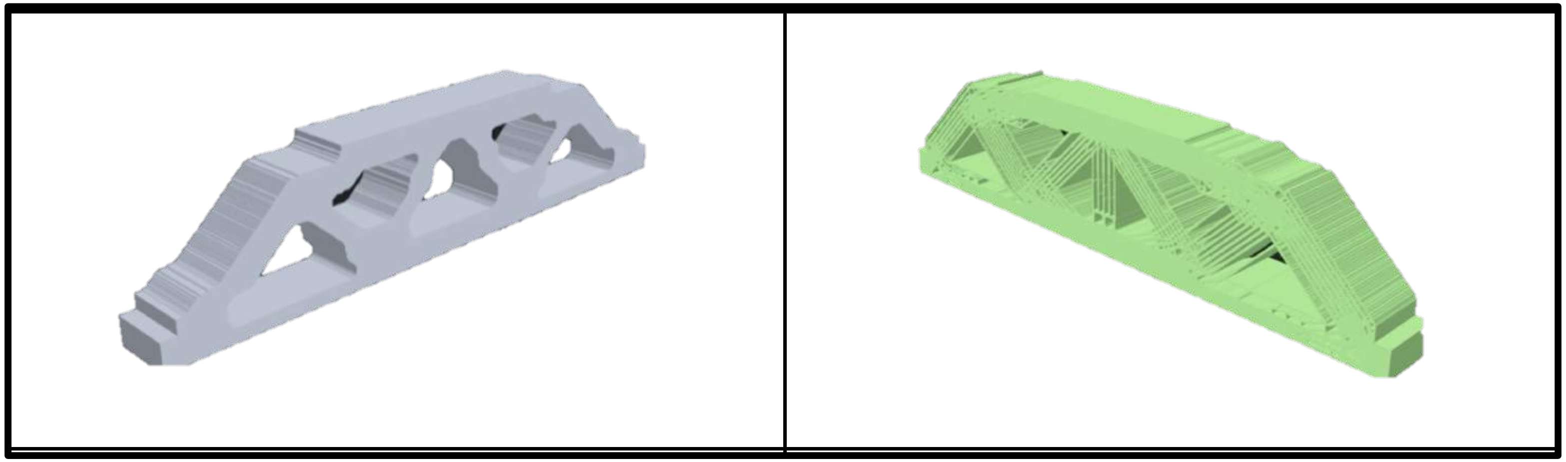
AE4ASM521

- Additive Manufacturing



Professor Joseph Morlier

Institut Supérieur de l'Aéronautique et de l'Espace SUPAERO

INSTITUT CLÉMENT ADER

https://ica.cnrs.fr/en/author/jmorlier/

Au programme

Duration	Description	Agenda
15’	Design Optimization	Refresh
15’	Computing Derivatives	Mathematical Background
15’	Topopt review	Need a break ?
15’	Current CAD-CAE	Quiz
15’	GGP for ALM	Our research
15’	Ecodesign	That’s new

Fondation ISAE-SUPAERO
Reconnue d’utilité publique

ANR
PLAN D’ACTION 2021

CONSTRUCTION DE L’ESPACE EUROPEEN DE LA RECHERCHE ET ATTRACTIVITE INTERNATIONALE

Programme : « *Montage de Réseaux Scientifiques Européens ou Internationaux* »
- Edition 2021, Vague 1 -

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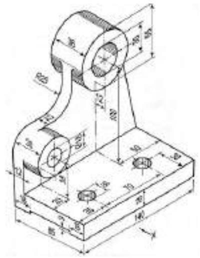
Current CAD-CAE

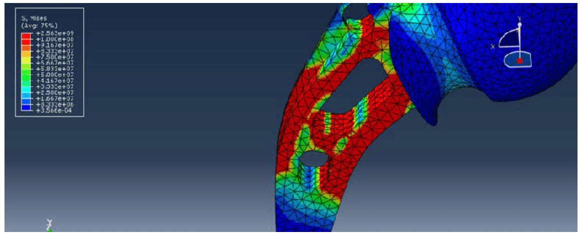
Practical Engineering Skills

CAD design

(engineering drawings)

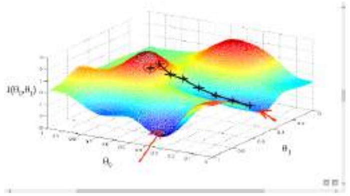


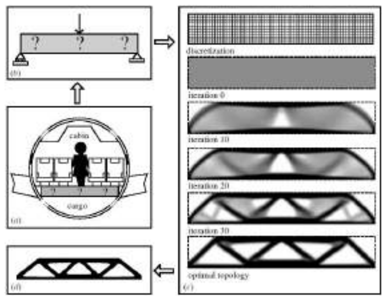
Finite Element Simulations



(stress analysis)

Gradient descent optimization
(TopOpt)





Altair

Additive Manufacturing
(future of industrial standards of manufacturing)

Software list

https://www.topology-opt.com/software-list
abaqus, nastran, ansys, comsol, ls dyna etc…

https://www.materialise.com/en/cases/materialise-3-matic-makes-topology-optimization-more-attractive

Mechanical Learning of Additive Manufacturing Parts

- Highlights of MATLS 2H04A (2018) – Structure Materials Design Project

McMaster University

Compression tests of students' designs
(video click to play: crushed samples will disappear)

Sessional Instructor: Dr. Bosco (Hiu Ming) Yu, PhD
2018

Mechanical Learning of Additive Manufacturing Parts

- Highlights of MATLS 2H04A (2018) – Structure Materials Design Project

Mechanical Learning of Additive Manufacturing Parts

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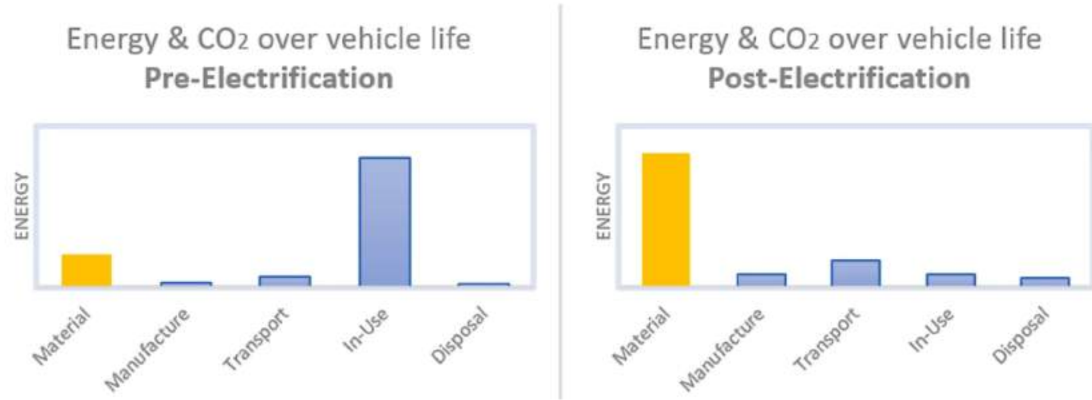
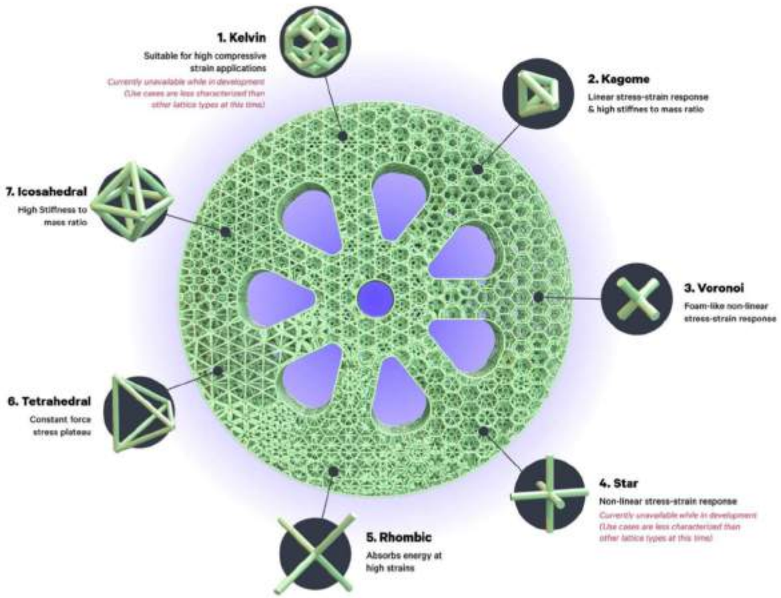
Material/Process as new design variables

Eco Material selection
Eco Process selection



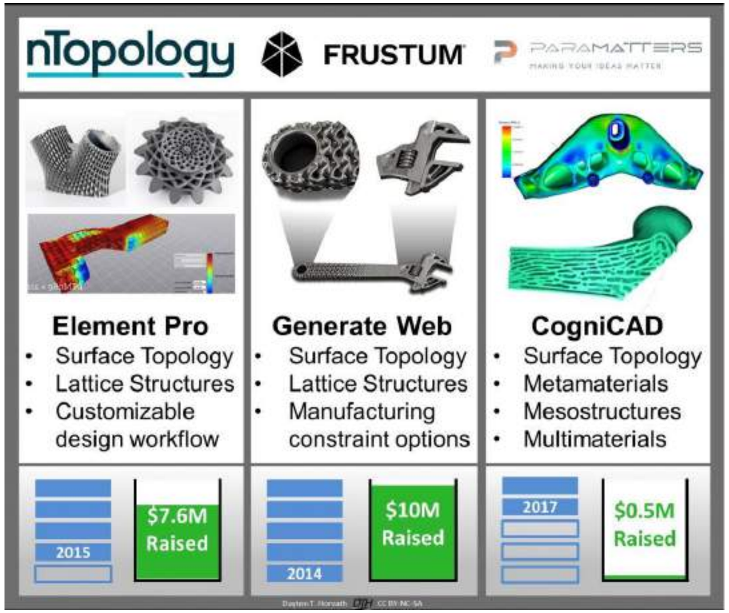


Unit cell design (anisotropy)
Digital materials

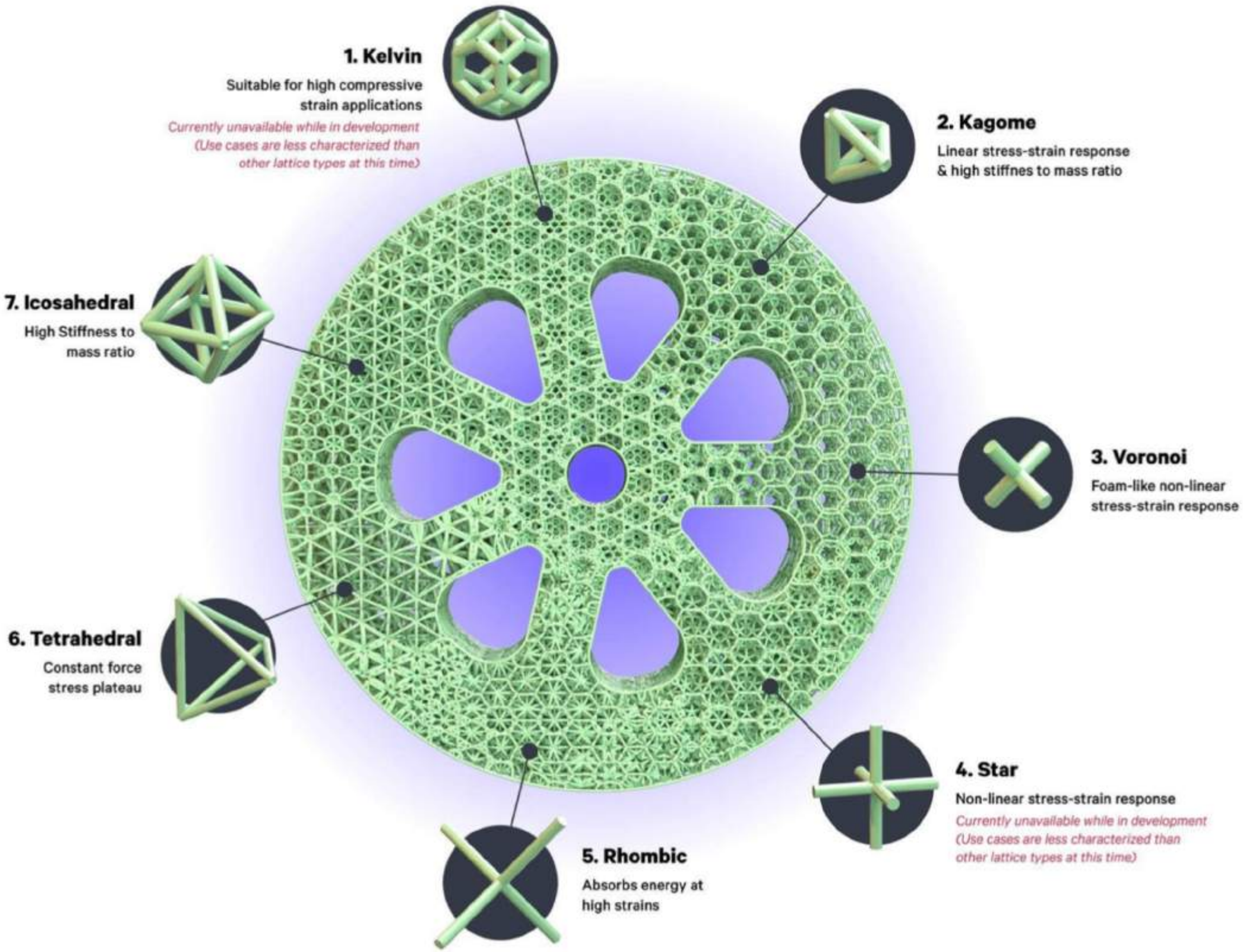
Softwares for **hierarchical design**

Conventional CAD programs do not work well
New players are emerging

Examples:
- ntopology (see case studies): https://ntopology.com/
- additiveflow: https://www.additiveflow.com/
- Hyperganic
- ParaMatters: https://paramatters.com/
- Fusion 360 (Autodesk)



Carbon example



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GGP for ALM

Reproducible Research

- https://www.topopt.mek.dtu.dk
- https://www.top3d.app

Giving Back with SUPAERO's students

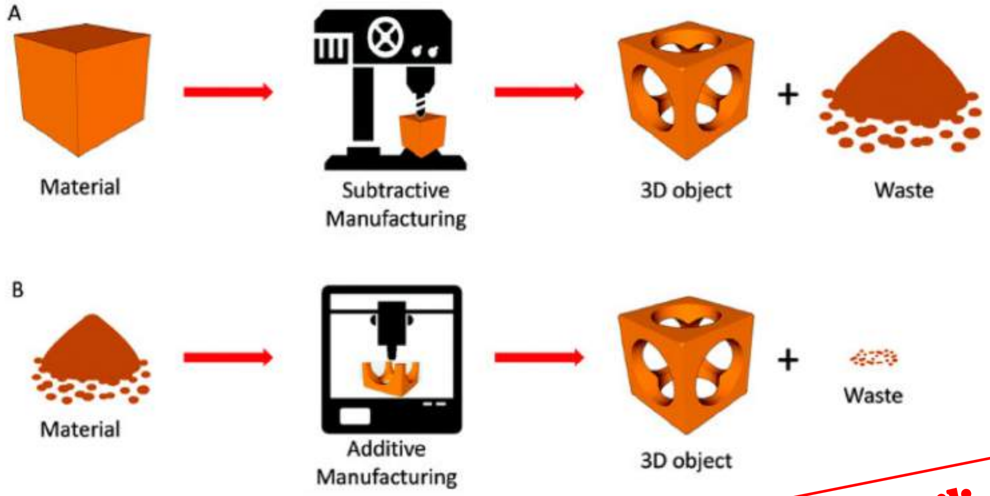
Prof's Impact increases, Students are becoming Phds, or R&T engineers (these new researches feed the courses&RP)

Students produce research papers and obtain new skills !

At master Level: Prof teaches elective courses+ Research Projects with students

Students are PhD Ready in CSM/MDO

Why Metallic 3D printing?

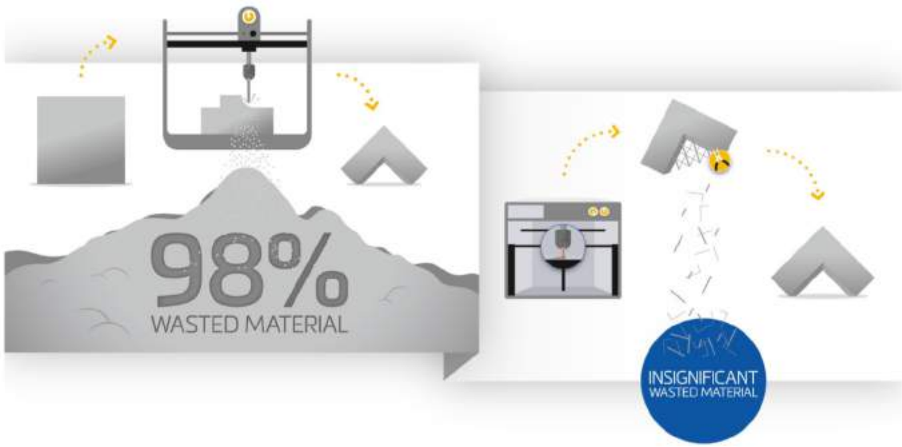


+Near 100% material utilization
+Recyclability
+LCA of 3D printing machine
+Monitoring

'Buy to-Fly' Ratio

- Traditional subtractive manufacturing machining techniques often result in a costly
- imbalance between the weight of raw material required to make a specific component, and the weight of the component itself — a relationship more commonly referred to (from its aerospace heritage) as the 'Buy-to-Fly' ratio.

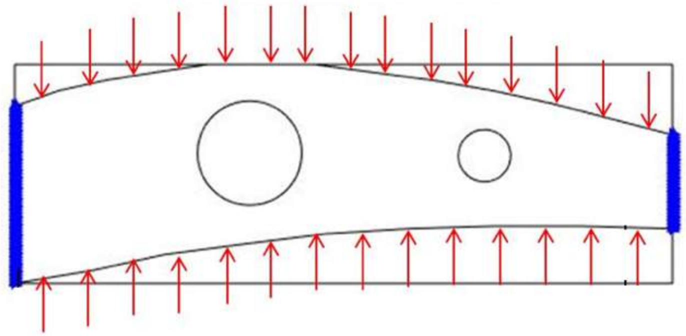
https://www.materialise.com/sites/default/files/resources/Whitepaper_Buy-to-Fly-Ratio_E.pdf

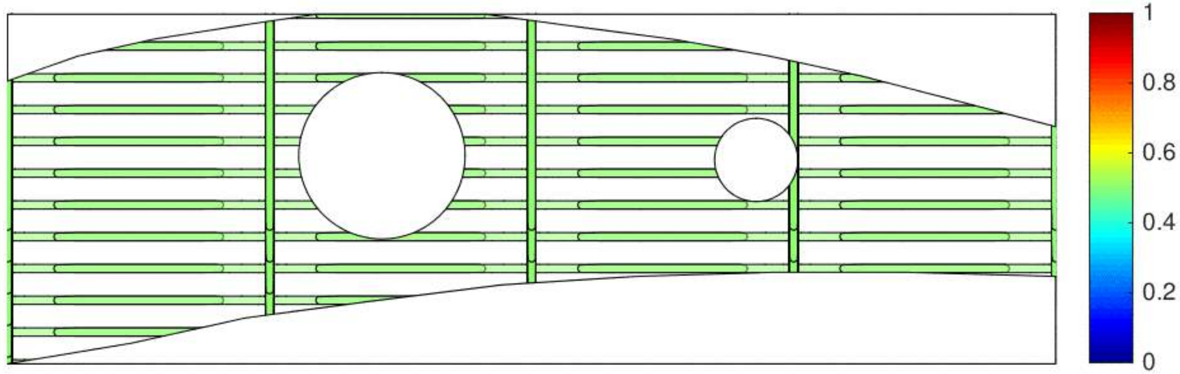


Reducing environmental impact

Near 100% material utilization, a high degree of material recyclability and reduced energy consumption (almost no additional tooling and less power required to move/manipulate lighter parts for product construction) has significant environmental benefits **whilst also generating cost-efficiencies.**

A typical Aerostructures, a simple **design**





use GGP

https://github.com/topggp/blog

State of the art

Design variables update

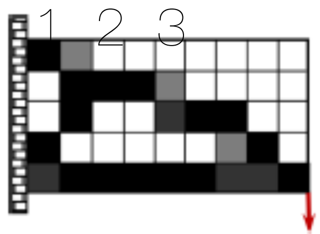
Interpretation

Model update
Density, Young modulus

Density based

variables : material density

$$x_1 = 1$$
$$x_2 = 0.5$$
$$x_3 = 0$$
$$\ldots$$

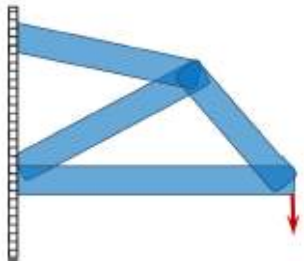


Innovative approach to help engineering solution identification : Components are placed in design space according to variables and material density are derived accordingly.

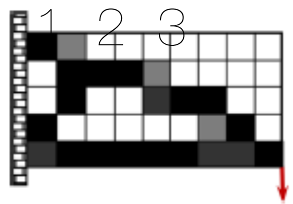
Lagrangian approaches

Variables : geometrical data

$$x_1 = Position$$
$$x_2 = Length, Height \ldots$$



Projection

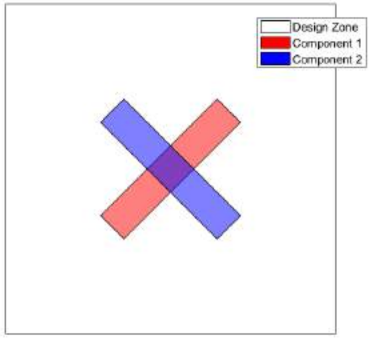


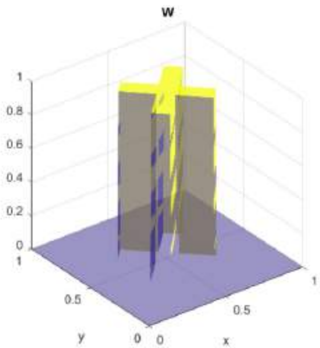
[12] Zhang, Weisheng, Jian Zhang, and Xu Guo. "Lagrangian description based topology optimization—a revival of shape optimization." Journal of Applied Mechanics 83.4 (2016): 041010.

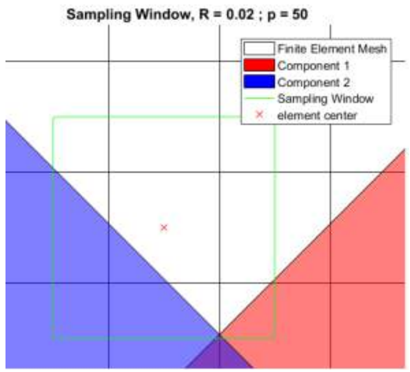
Design is made of engineering bricks like: beam, plate, brick….

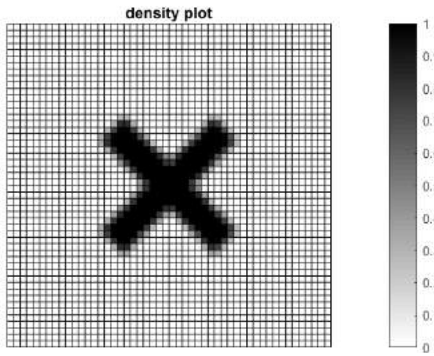
Generalized Geometric Projection

Geometric features → Characteristic functions → Generalized Geometric Projection → Fixed mesh model update









[Coniglio et al. 2019]

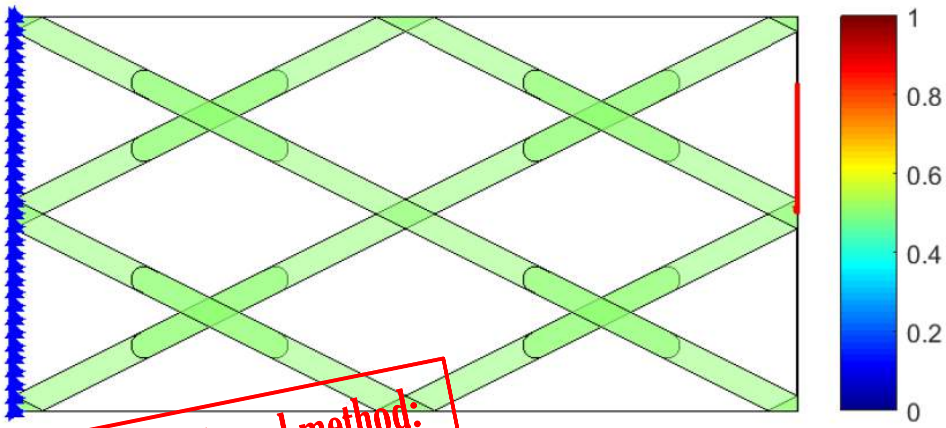
$$\mathbf{D}(\{X_g\}, \boxed{p}, \boxed{R}) = \{\{X\} \in \mathbb{R}^{d_g} \mid \| \{X\} - \{X_g\} \|_{2p} \leq R\}$$

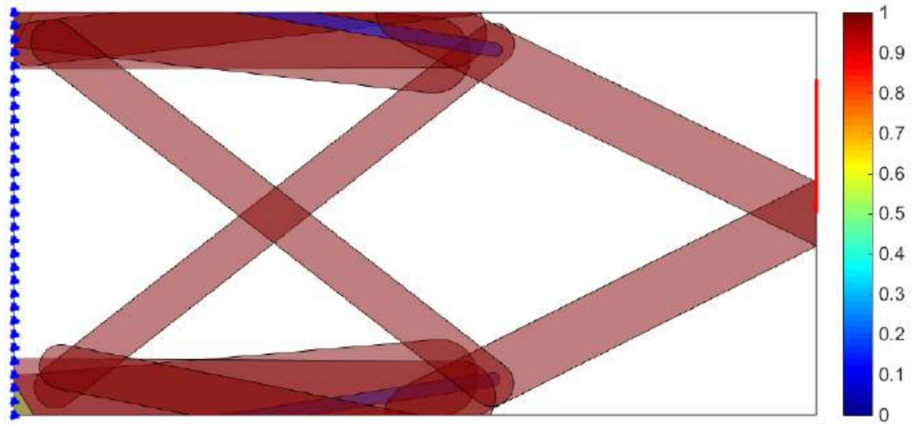
$$\delta_i^{el}(\boxed{W_i}, p, R) = \frac{\int_{\mathbf{D}(\{X_g^{el}\},p,R)} W_i(\{X\}, \{X_i\}, \{r\}) d\Omega}{\int_{\mathbf{D}(\{X_g^{el}\},p,R)} d\Omega}$$

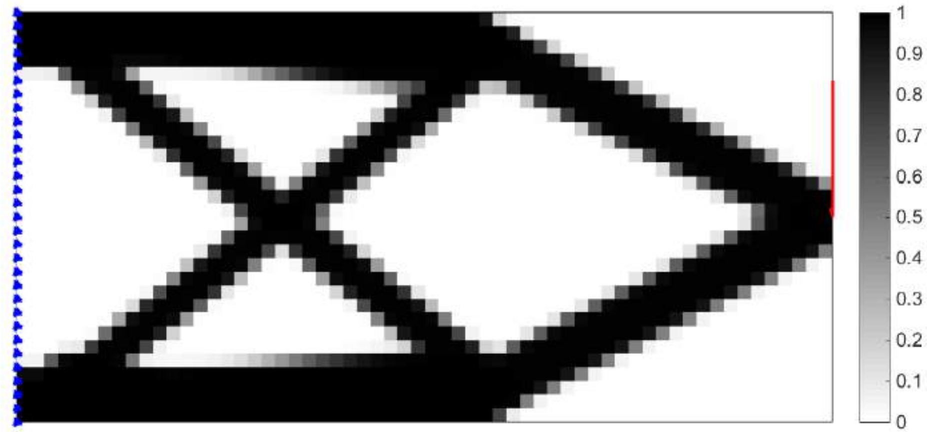
$$E^{el} = \boxed{\mathbb{M}}(\{\delta^{el}\}_c, E, E_{min}, \kappa)$$

$$\rho^{el} = \boxed{\mathbb{V}}(\{\delta^{el}\}_v, \kappa)$$

Generalized **G**eometry **P**rojection (GGP)





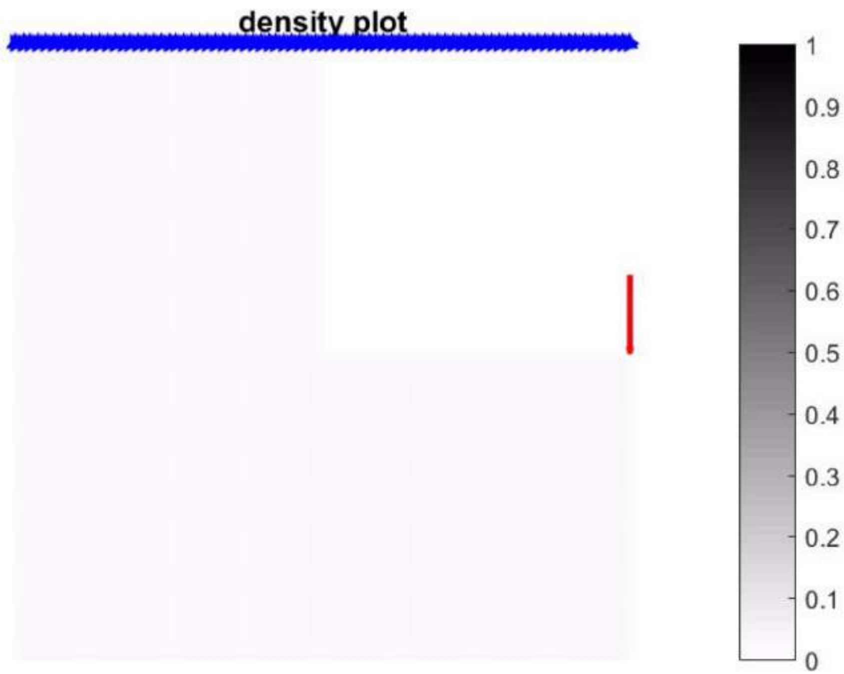


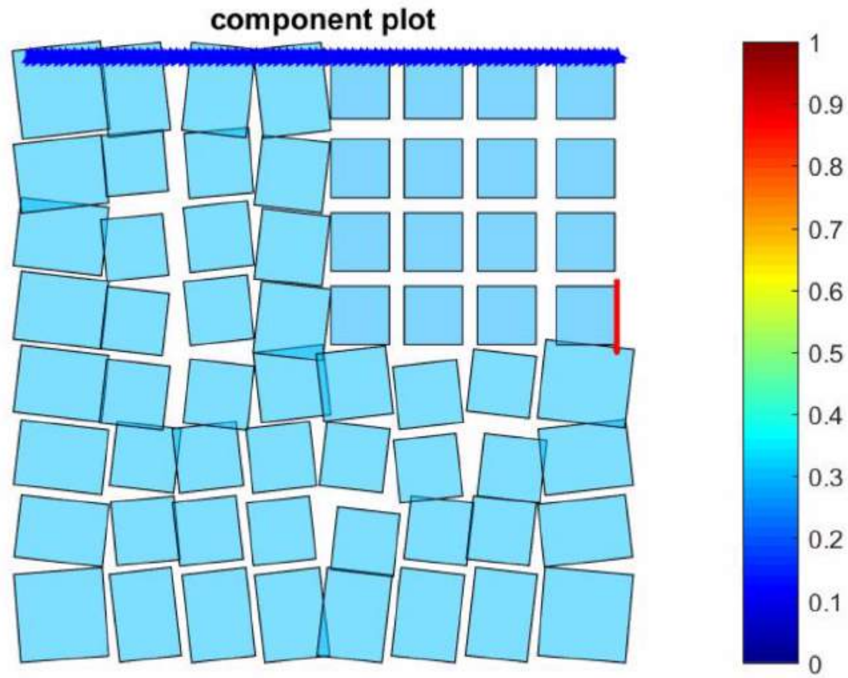
For every Gradient based method:
X0 sensitivity

$$\begin{cases} \min_{\{x\}} C = \{U\}^T\{F\} \\ s.t. \\ V = \dfrac{\sum_{el=1}^{N} \rho^{el}}{N} \leq V_0 \\ \{l_b\} \leq \{x\} \leq \{u_b\} \end{cases}$$

Results MNA, 8*8*6=384 design variables minC st Volfrac=0,4

At the end, explicit assembly of components!

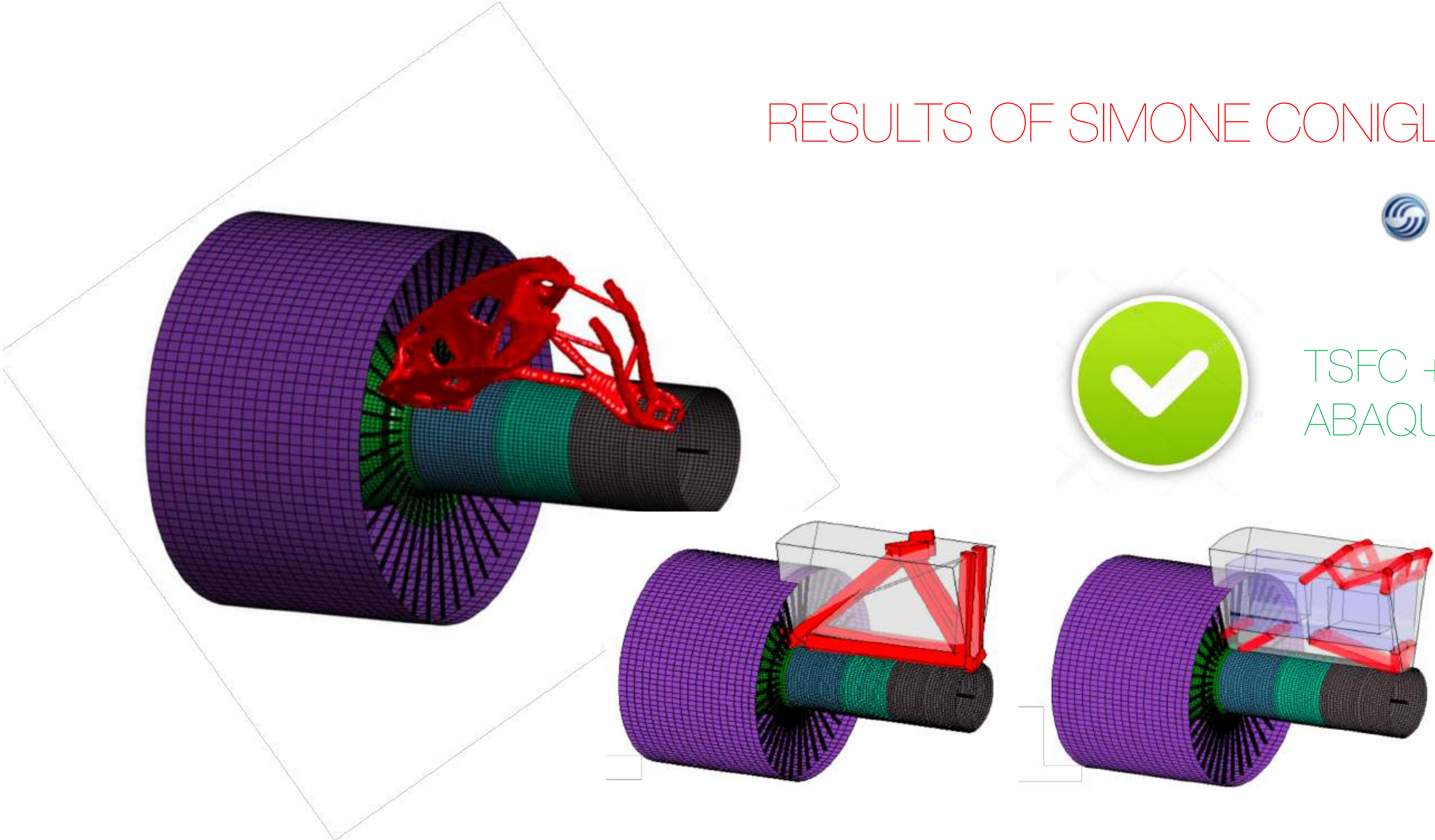




Bionic SIMP vs EXPLICIT TRUSS vs EXPLICIT BOX

RESULTS OF SIMONE CONIGLIO PHD

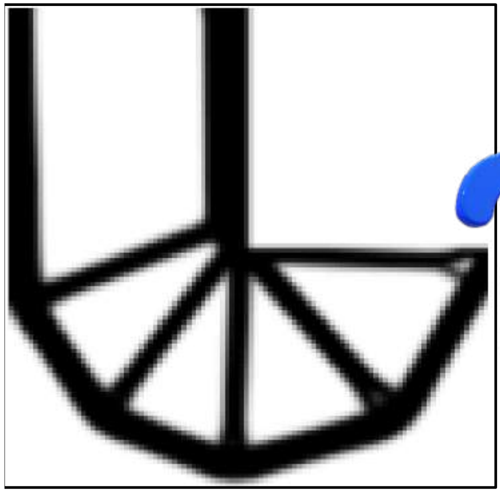
AIRBUS



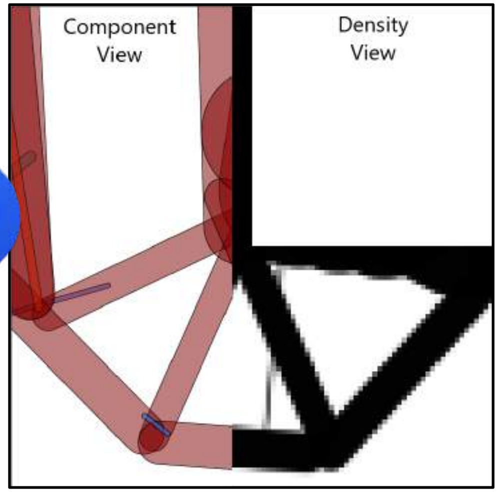
TSFC + STRESS
ABAQUS REANALYSE

GGP For ALM?

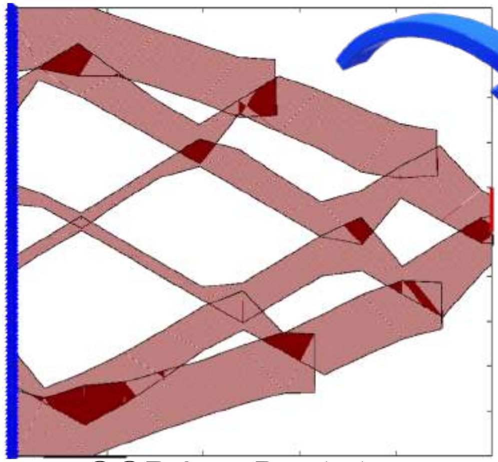
Prof, How can I do that?



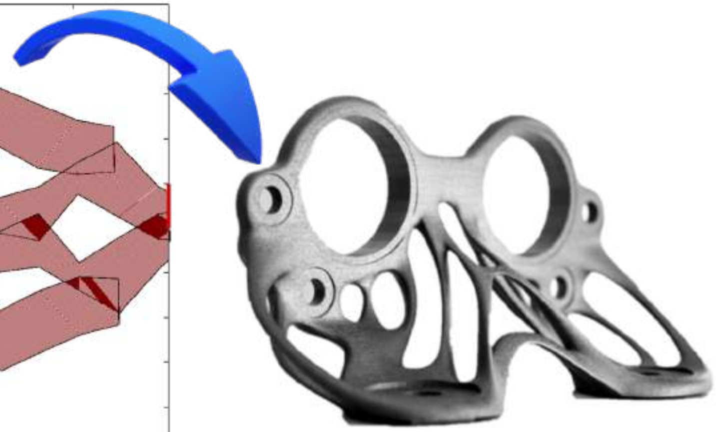
SIMP



GGP



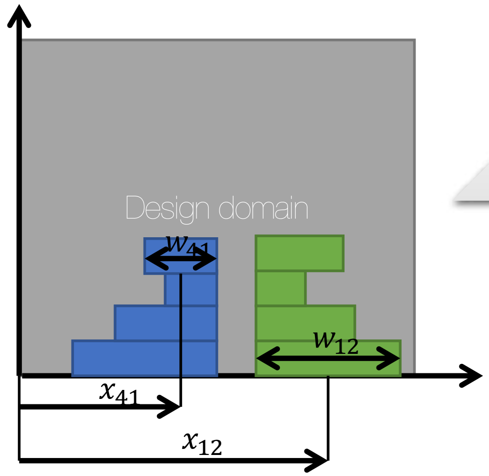
GGP for 3D printing



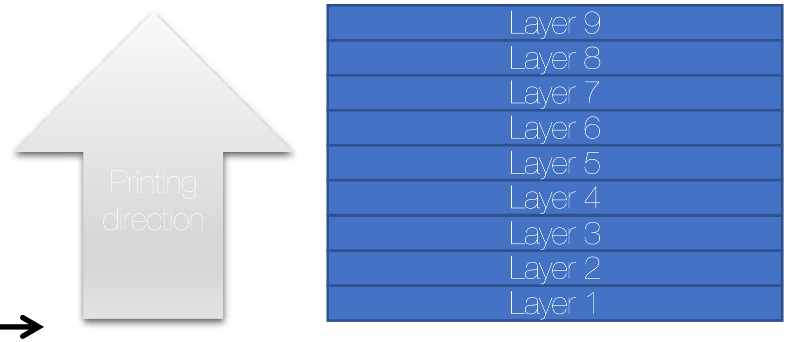
3D Printed part

ALM based on explicit topopt

A solution is determined by its manufacturing process: (in this case printing path)



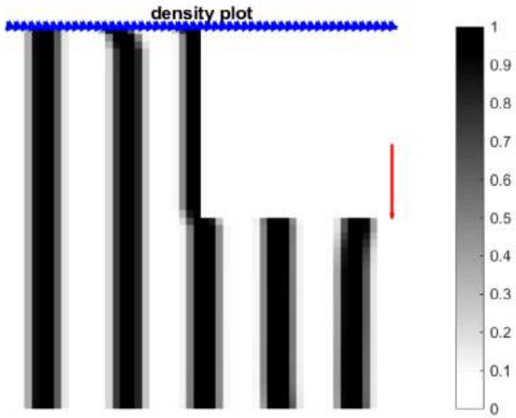
Printing process is discretized in n_l layers

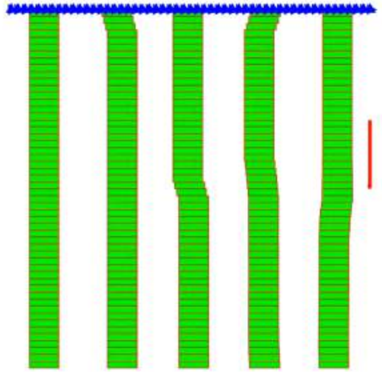


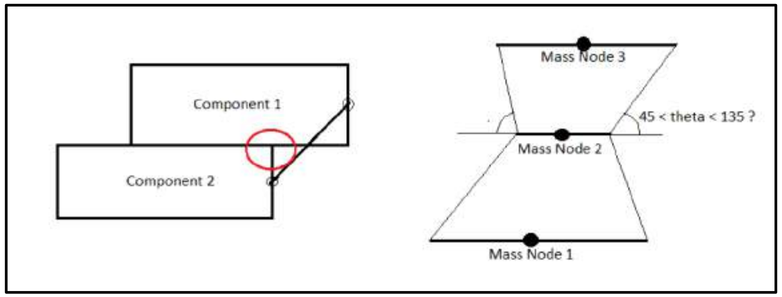
- MNA Components are replaced by printed branches
- Design variables will be printed branch position and width per layer: x_{li}, w_{li}
- For each layer a projection is made to get the solid model modulus

ALM based on explicit topopt

$$
\begin{cases}
\min_X c = F^T \cdot U \\
s.t. \\
\sum_{i=1}^{N} \rho_i - v_f N \le 0 \\
\theta_l \le \theta \quad \le \pi - \theta_l
\end{cases}
$$





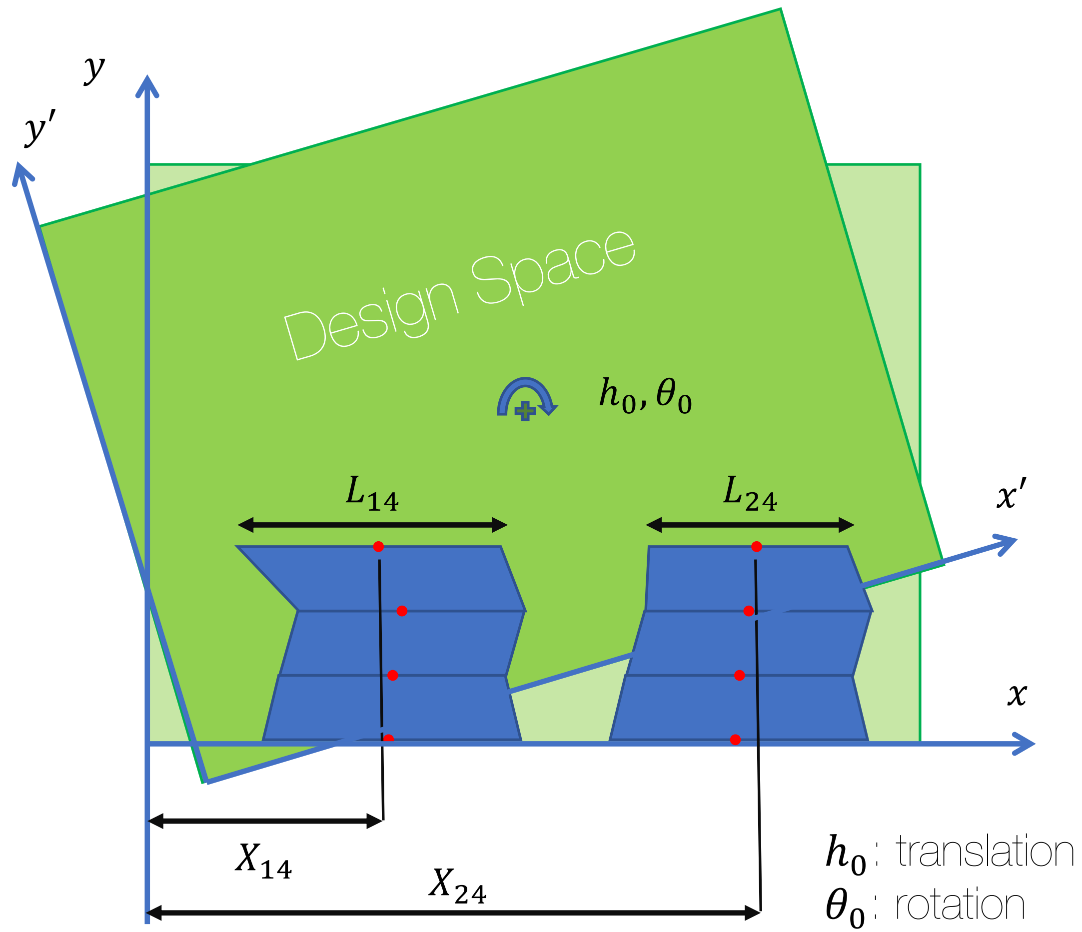


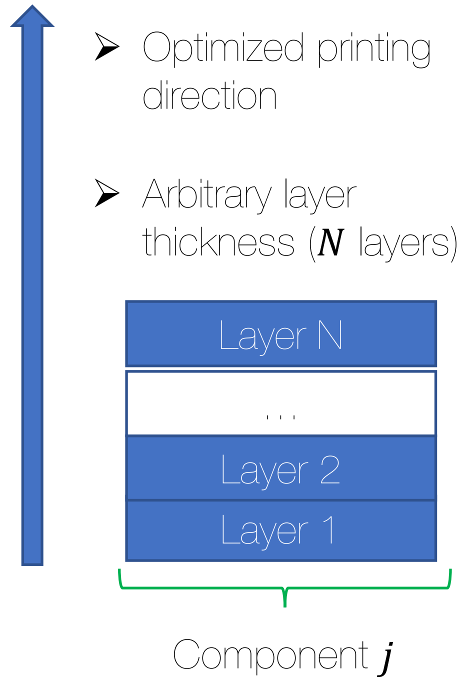
$N_x = N_y = 52$
$v_f = 0.4$
$5\ printing\ components$
$18\ printing\ intervals$
$5 \times 18 \times 2\ design\ variables$

Current approaches

	Check on	Overhang angle	Bridge length	Optimal printing **plane**	Comment
SIMP [Leary et al. 2014]	Boundaries	Yes	No	No	Additional **iterations**
AM Filter (SIMP-based**)** [Langelaar 2015]	Densities	Yes	No	No	One constraint **per element**
Level-set [Allaire et al. 2017]	Boundaries	Yes	Yes	No	Implicit **constraints**
MMV [Guo et al. 2017]	Boundaries	Yes	No	No	
MMC [Xian et al. 2019]	Components angles	Yes	No	Yes	Difficult **quality check**

ALM based GGP: Last Results





- Optimized printing direction
- Arbitrary layer thickness (N layers)

Problem Statement

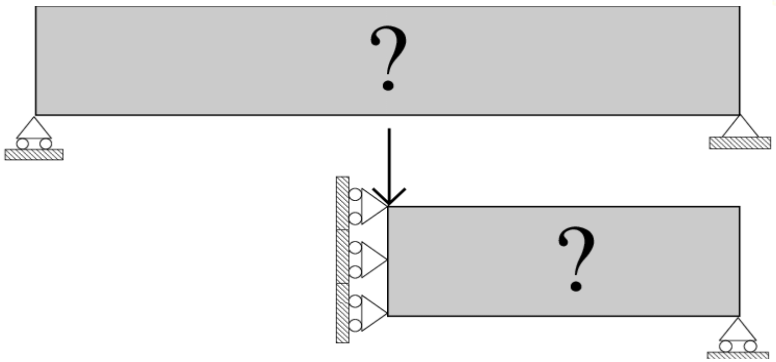
$$
\begin{cases}
min & & C(X, U_f) \\
s.t.: & & V \le V_0 \\
& \theta_{ij} \le \theta_{max} & \forall\, i = 1, \dots, N \quad j = 1, \dots, M \\
& BL_{ij} \le BL_{max} & \forall\, i = 1, \dots, N \quad j = 1, \dots, M
\end{cases}
$$

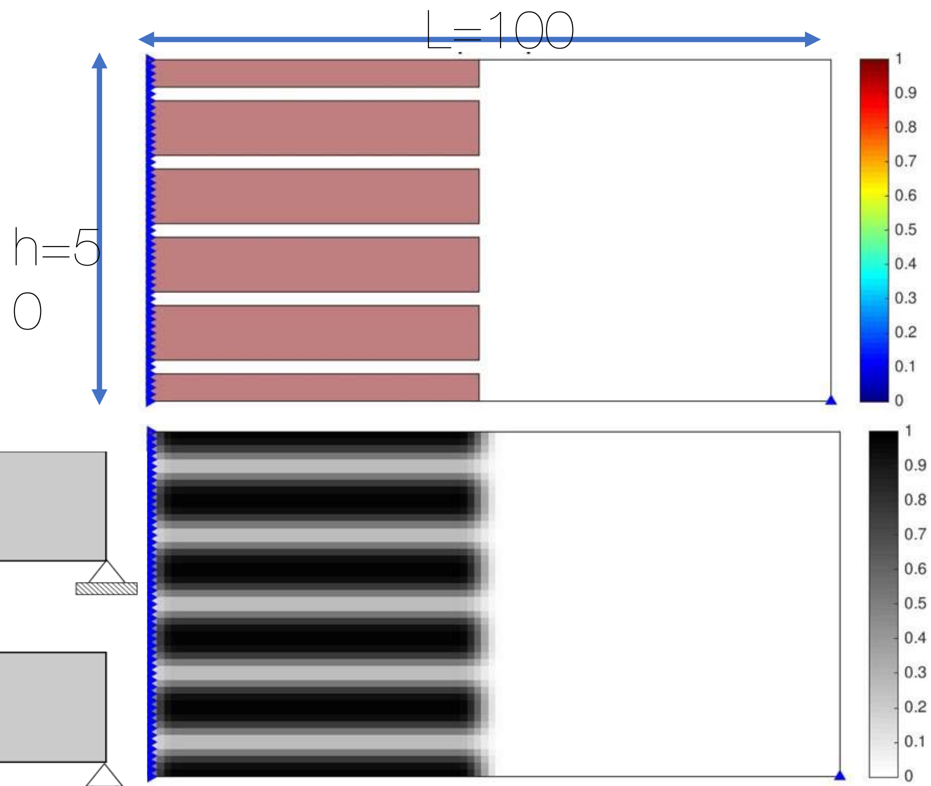
- N layers per component
- N+1 segments per component
- M components
- 2 features per segment (X_k, L_k)
- 2 features per component (h_j, m_j)
- 2 global features (h_0, θ_0)

$\Rightarrow$ $2M(N+2)+2$ design variables

MBB Results: convergence

Parameter	Quantity
Mesh	50×100
V_0	0.3
θ_{max}	$45°$
$BL_{max}/2$	25
N (layers)	18
M (components)	6
Design variables	242





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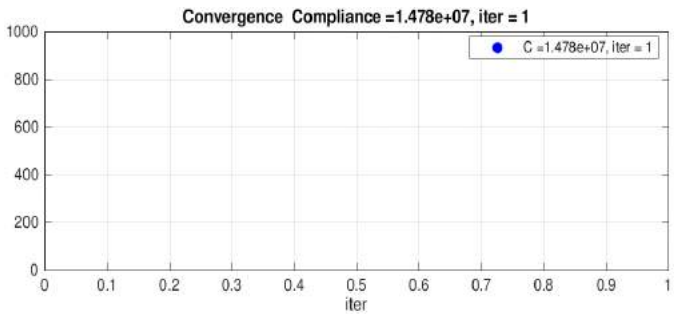
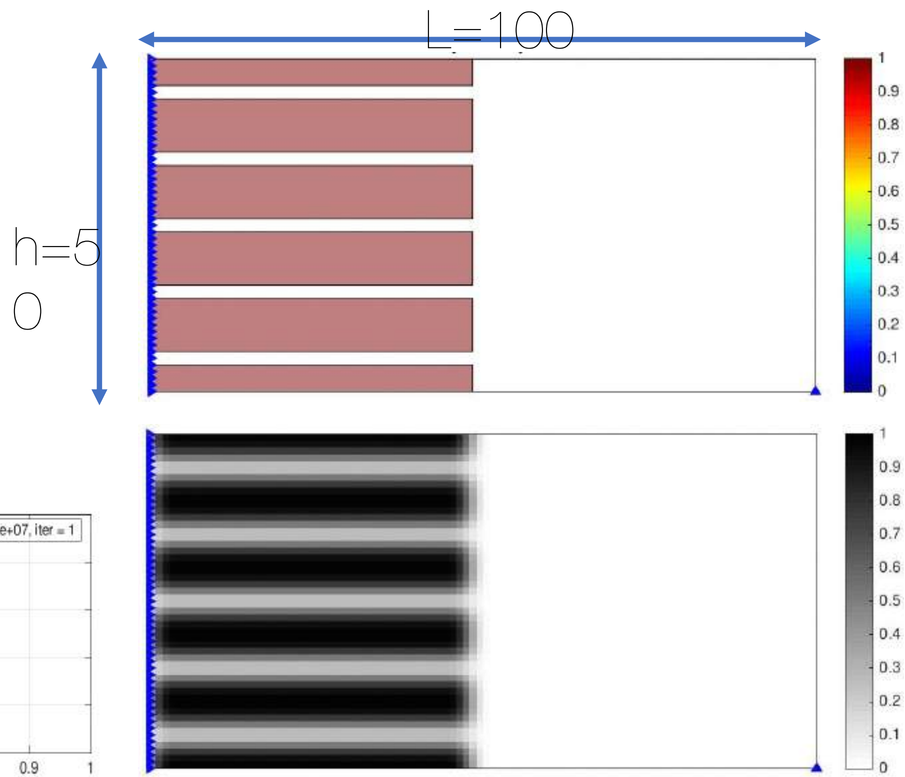


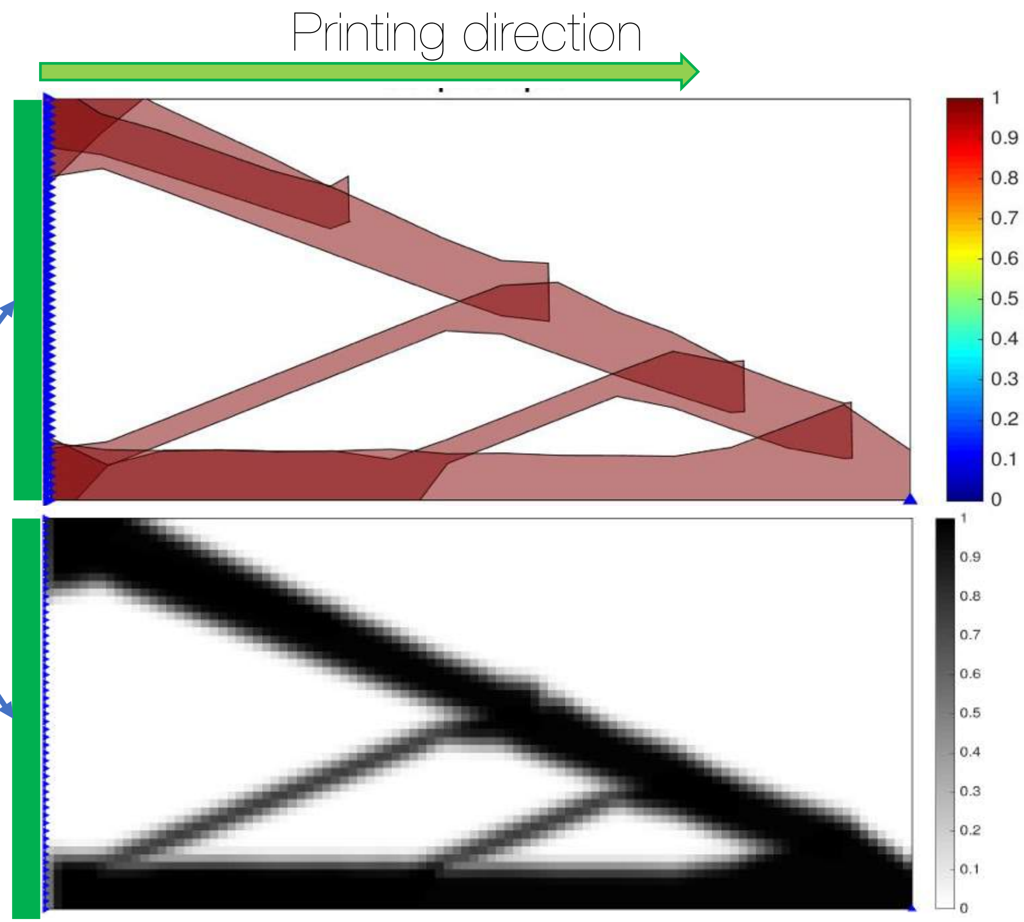


MBB Final Results

Compon
ent plot

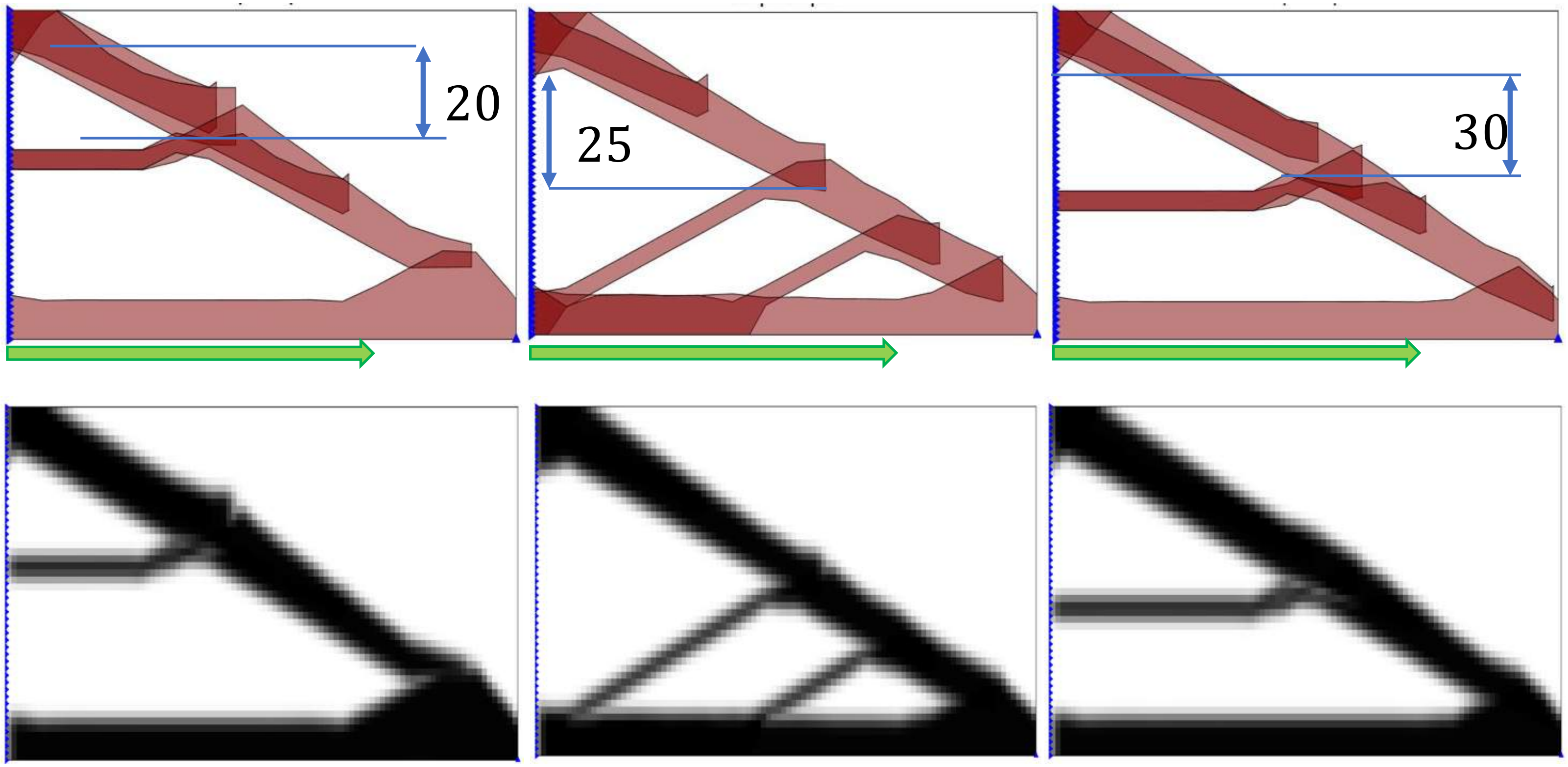
Printin
g plan

Densiti
es
plot



Bridge length variation

$BL_{max}/2 = 20$ $BL_{max}/2 = 25$ $BL_{max}/2 = 30$



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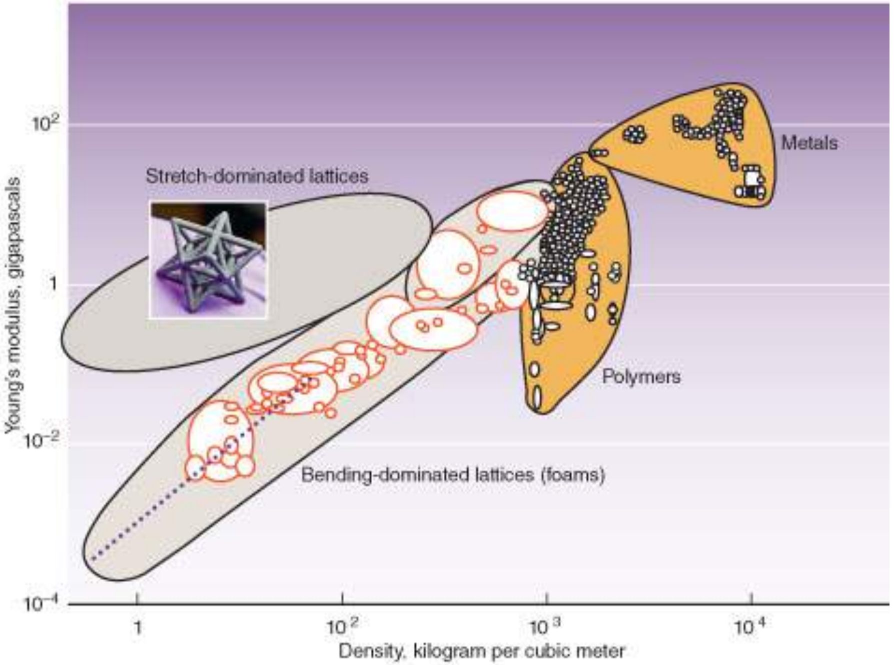
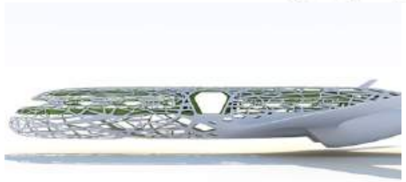
Ecodesign

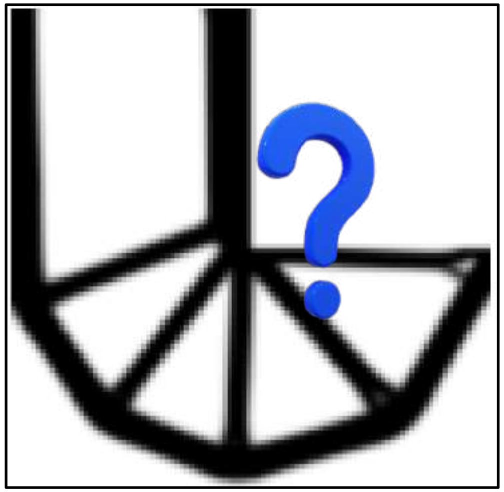
DIGITAL MATERIALS

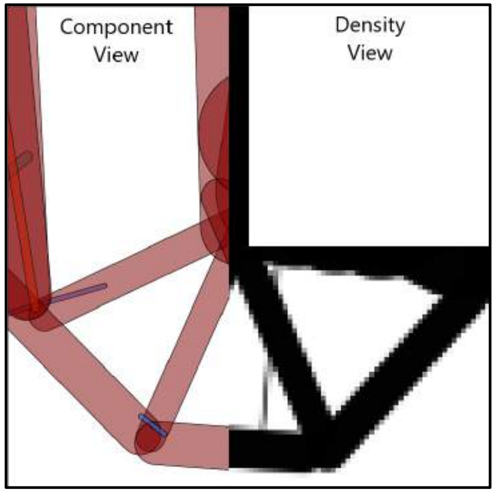


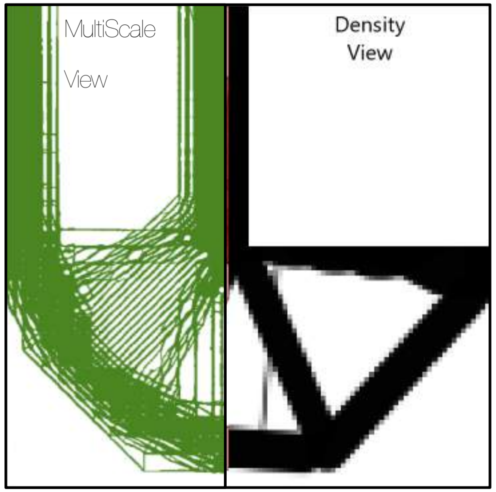


Chris Spadaccini (llnl,USA)
"By controlling the architecture of a microstructure, we can create materials with previously unobtainable properties in the bulk form."

How to do Eco efficient Ultralight Structures?

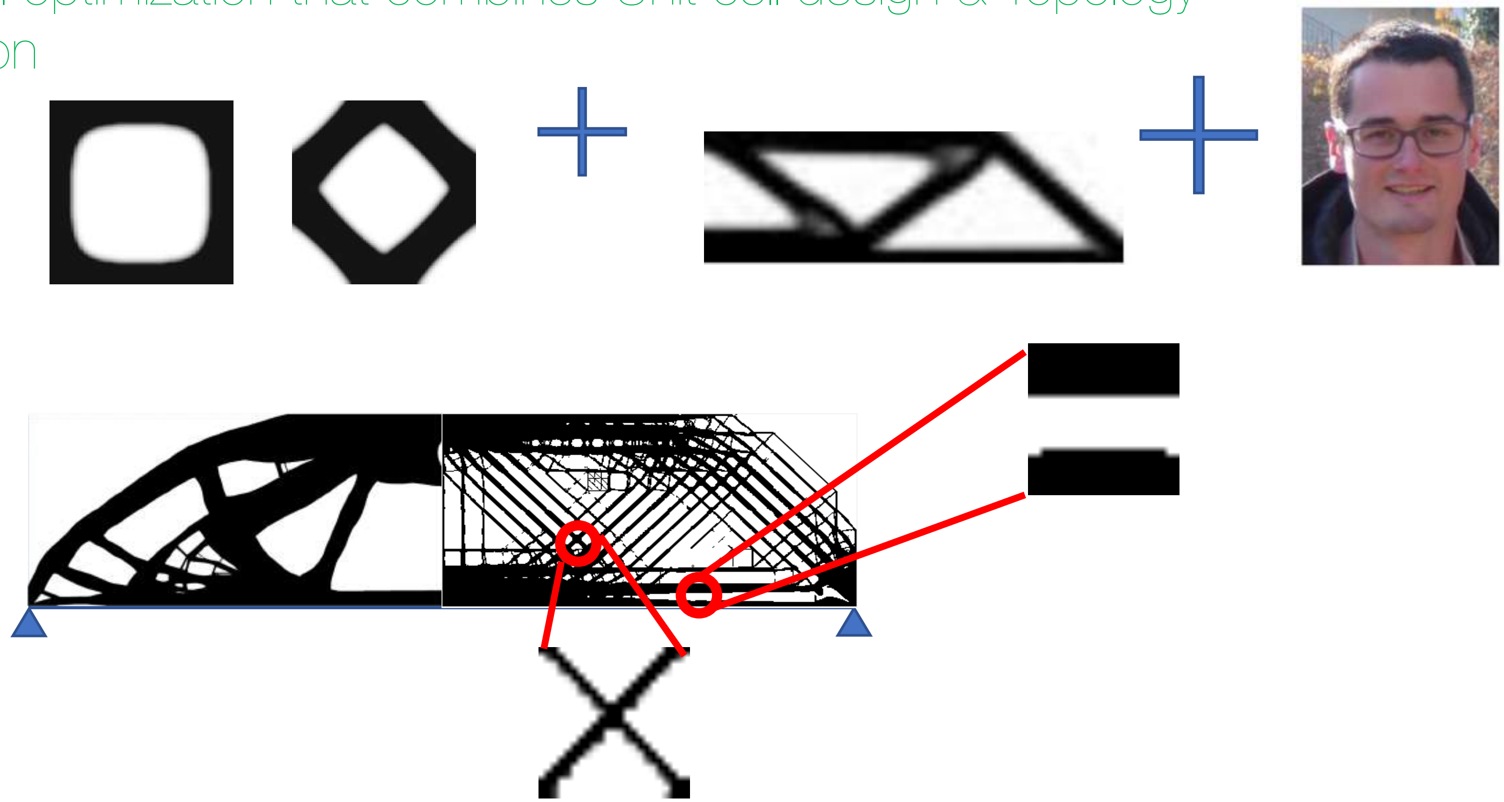




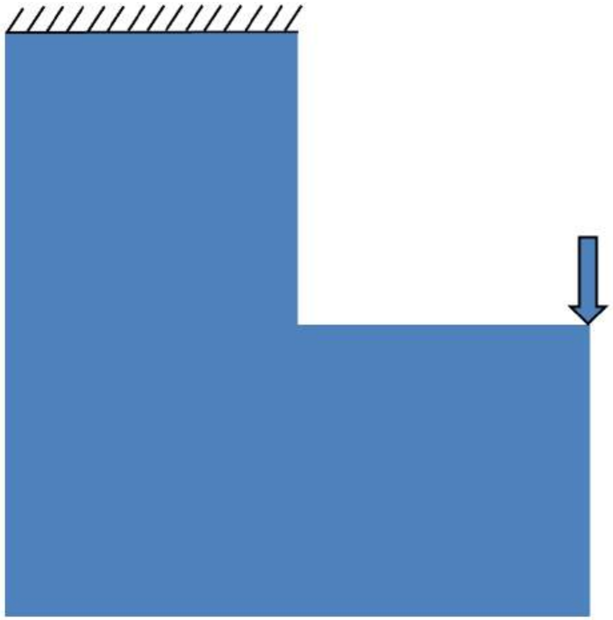


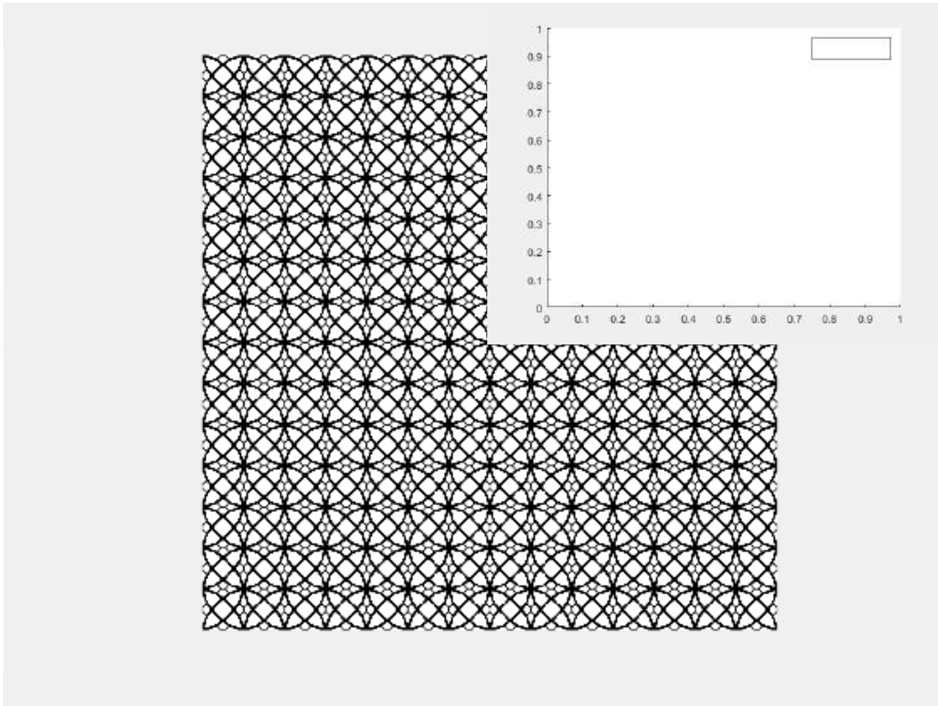
Multi-scale TO

A two level optimization that combines Unit cell design & Topology Optimization



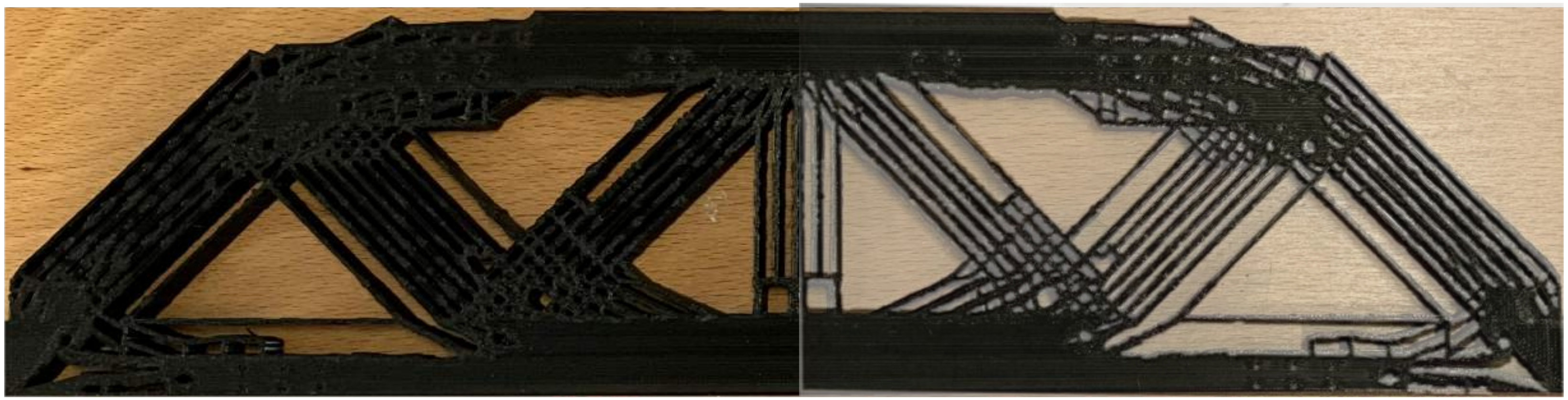
EMTO on L-shape (cellular /digital materials)





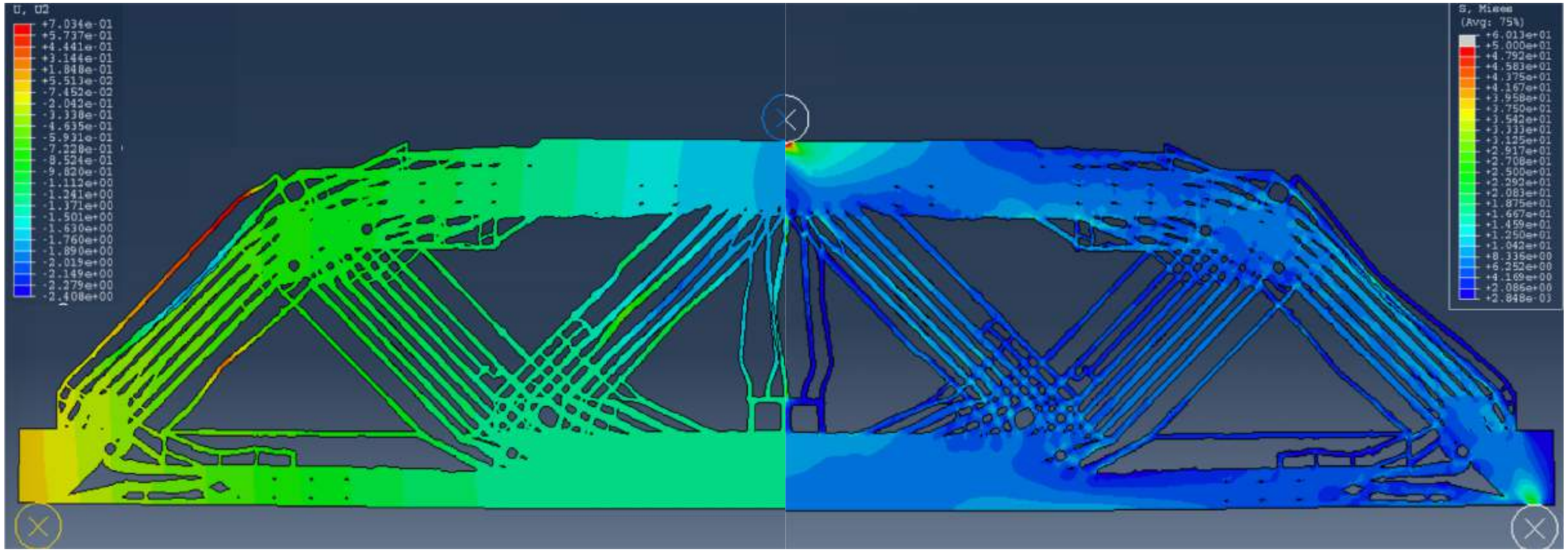
https://github.com/mid2SUPAERO/EMTO

EMTO 3D printed (3pts bending)



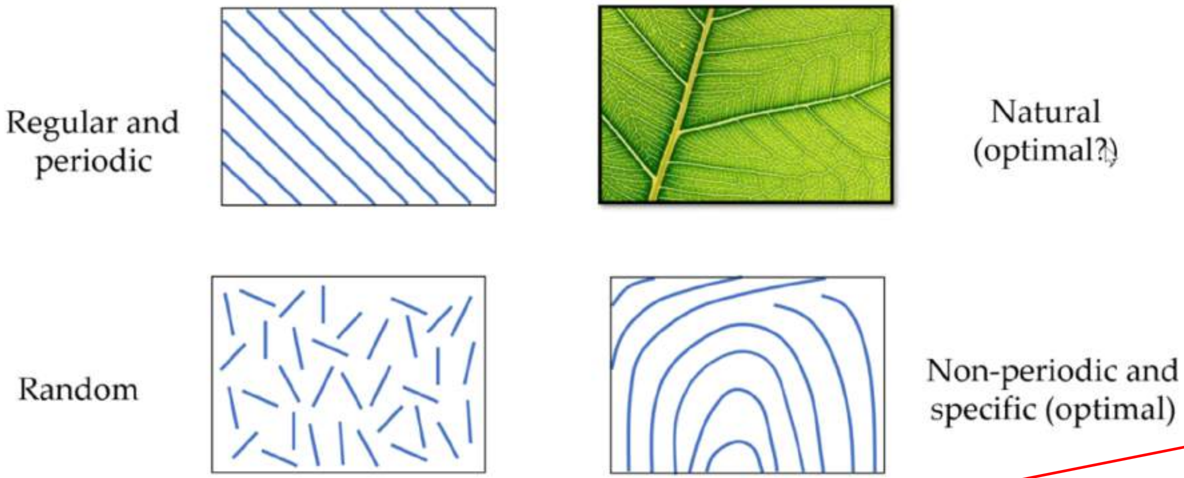
EXPERIMENTATION

EMTO 3pts bending (disp vs stress)



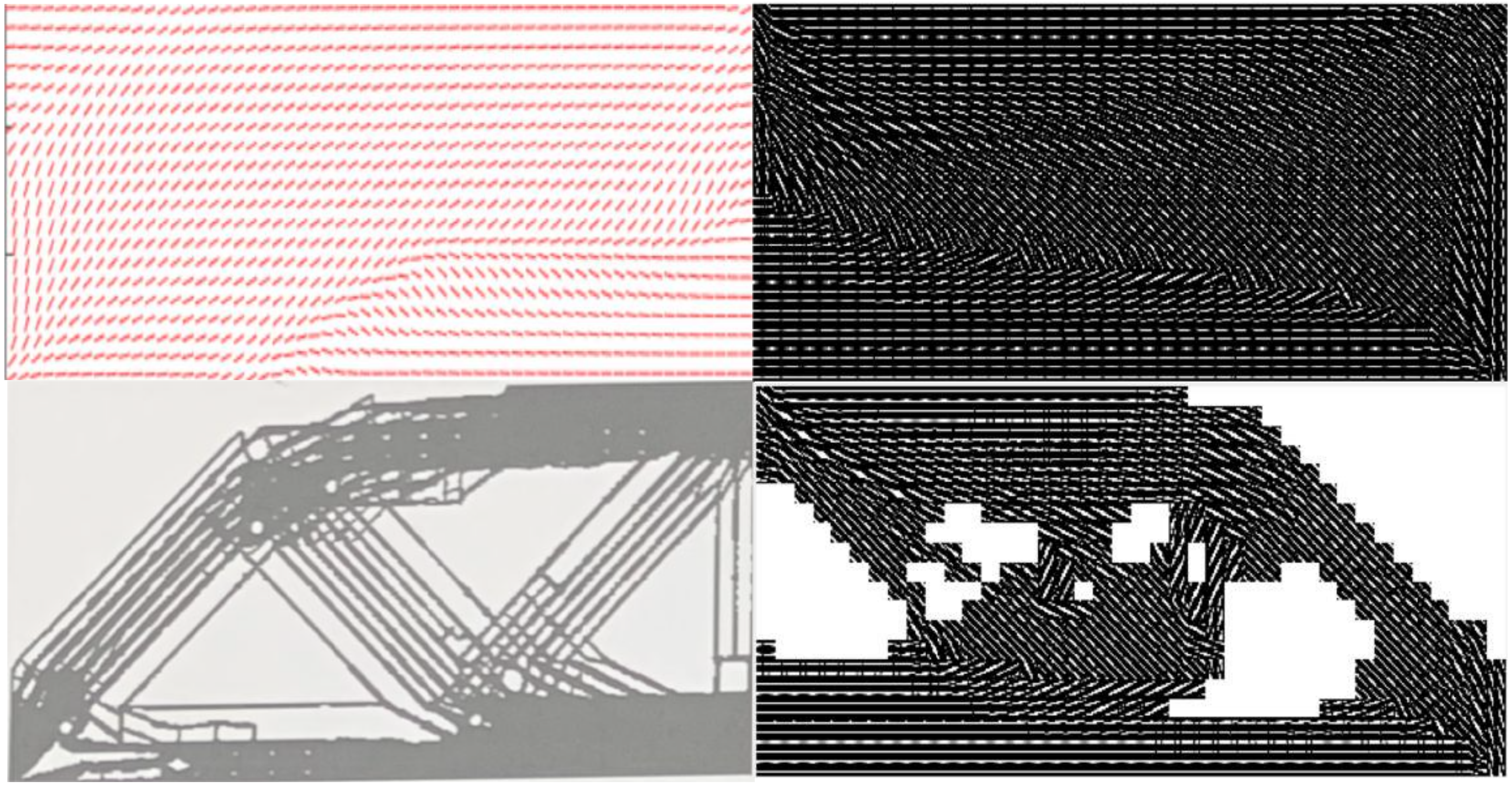
ABAQUS REANALYSE

Why Composites 3D printing?



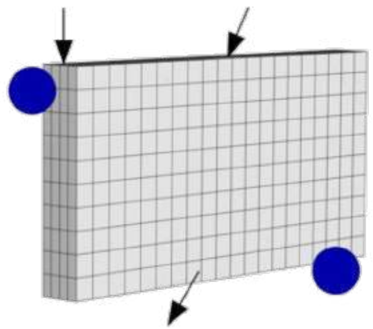
+ Automatic Fiber Placement + eco-fiber/resin selection +Monitoring

EMTO 3pts bending (Fiber placement)

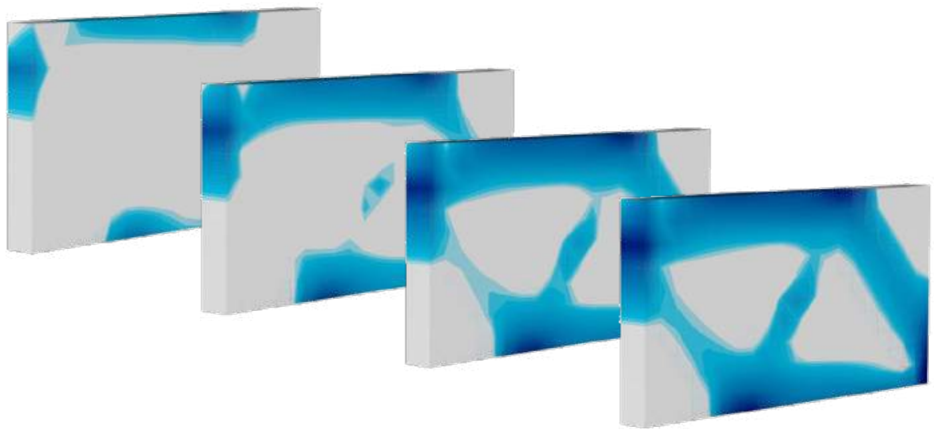


Ecodesign + Topology Optimization?

Topology Optimization



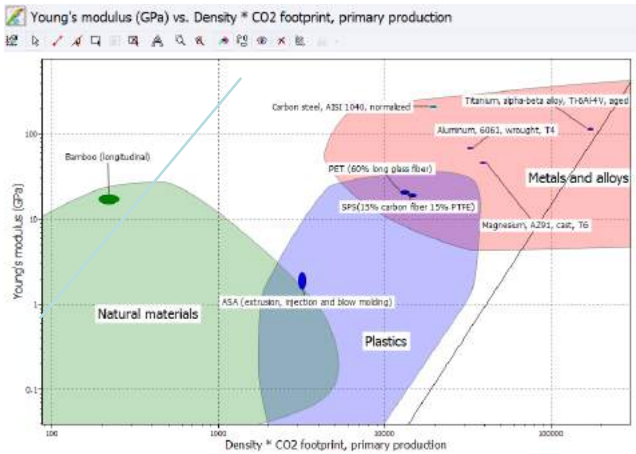
Inputs: Material, BCs and Loading



Outputs: design of a „stiff" bicycle frame

CO2 footprint minimization (Ashby's method)

Inputs: Type of Structures, materials



Outputs: Optimal material (bamboo) with optimal Design

#Generalized Ashby's theory compatible with TopOpt

#All In One problem is a MDO problem !!!

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www.elsevier.com/locate/procedia

32nd CIRP Design Conference

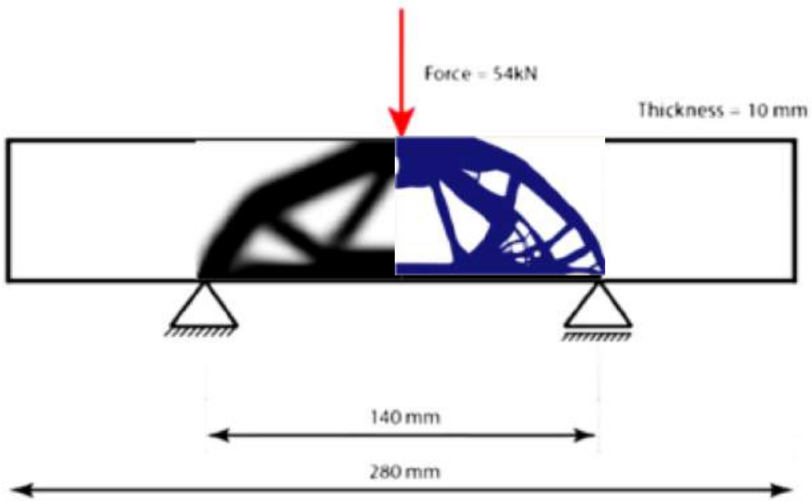
Ecodesign with topology optimization

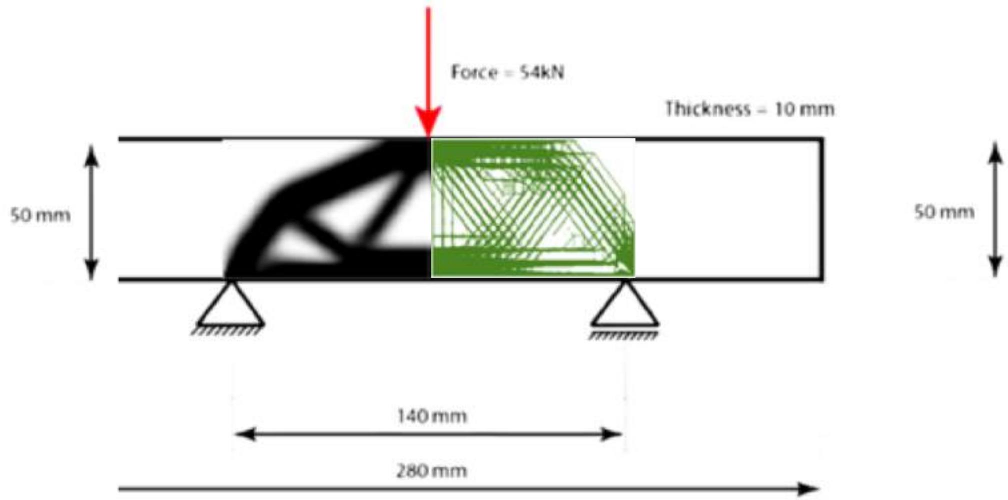
Edouard Duriez*[a], Joseph Morlier[a], Catherine Azzaro-Pantel[b], Miguel Charlotte[a]

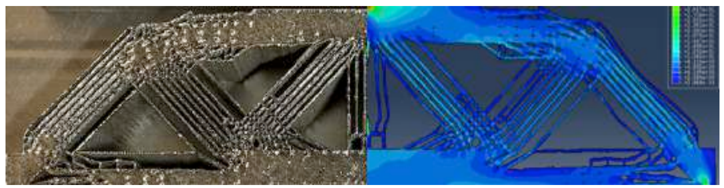
How to **ECO**design tomorrow's structures?

Prof. Joseph Morlier, Edouard Duriez, Miguel Charlotte, Catherine Azzaro-Pantel

Print it , Test it







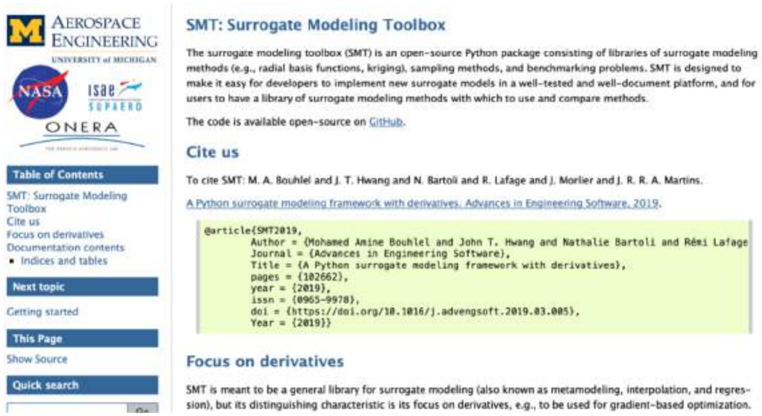
Conclusions

Researcher view (Reproducible Research)

- https://www.topopt.mek.dtu.dk
- https://www.top3d.app

- https://github.com/topggp/blog
- https://github.com/mid2SUPAERO/EMTO

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



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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.

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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.

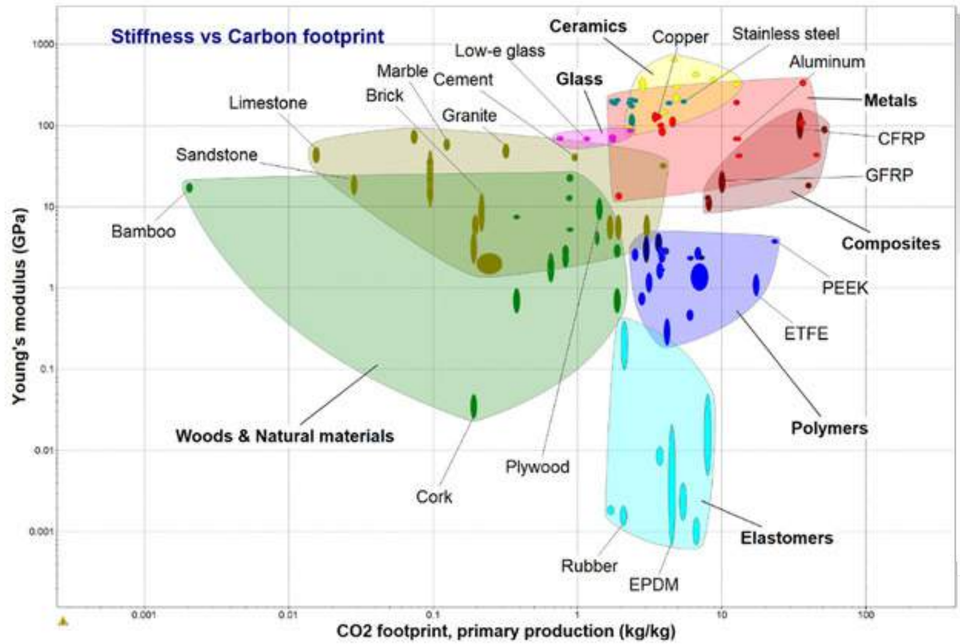
```
@article{SMT2019,
        Author = {Mohamed Amine Bouhlel and John T. Hwang and Nathalie Bartoli and Rémi Lafage
        Journal = {Advances in Engineering Software},
        Title = {A Python surrogate modeling framework with derivatives},
        pages = {102662},
        year = {2019},
      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

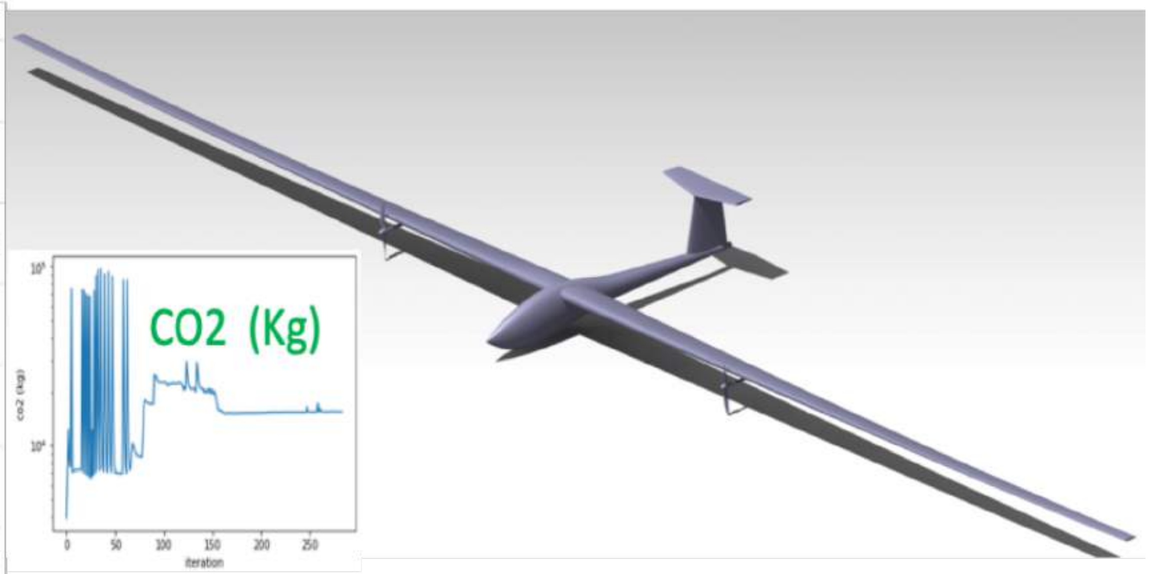
Focus on derivatives

SMT is meant to be a general library for surrogate modeling (also known as metamodeling, interpolation, and regression), but its distinguishing characteristic is its focus on derivatives, e.g., to be used for gradient-based optimization.

Min Mass or Min CO2?

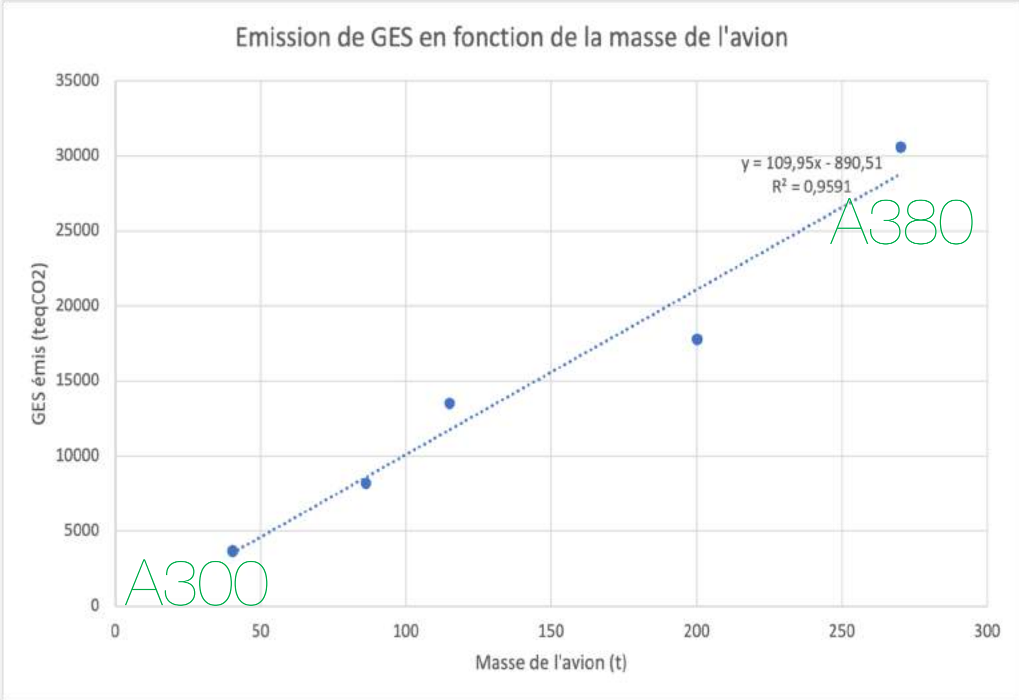
Trade-off between use phase (young's modulus and density) and production phase (CO2 footprint)





EcoHale Project @SUPAERO

First Order



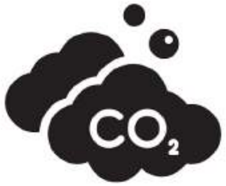
At the first order

min {mass} is close to min {CO2}

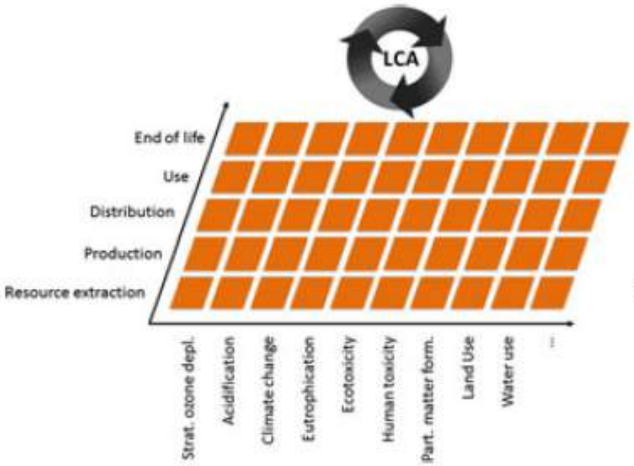
SUSTAINABLE ULTRALIGHT STRUCTURE BASED ON DIGITAL FABRICATION AND ARCHITECTED MATERIALS:

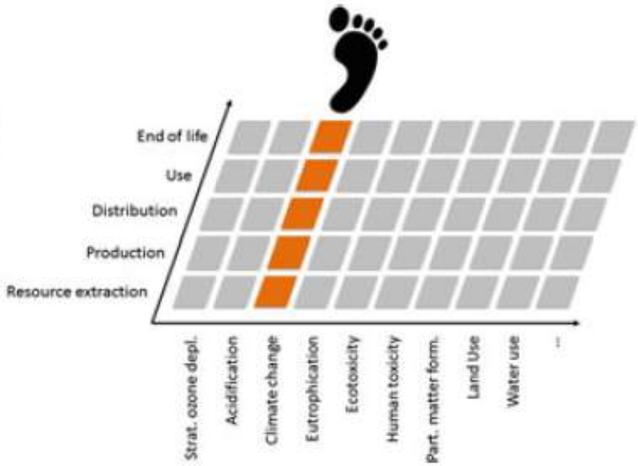
EcoDesign for Additive Manufacturing •Topology Optimization •Multidisciplinary Design Optimization •Surrogate Modeling

Materials + Manufacturing Process:=

Carbon Footprint	Energy requirement	Water withdrawal	Generation of waste
			

→ Take environmental issues into account from the design phase and throughout the life cycle (LCA)





Perspectives

→ Multiobjective formulation CO_2 versus Cost

→ Natural Fiber / Resin Eco-selection

→ Material performance prediction by AI (e.g. recycled composites)

Popularization

https://www.linkedin.com/pulse/possible-build-aircraft-wing-lego-joseph-morlier/?articleId=6627240732975480832

https://www.tripadvisor.fr/LocationPhotoDirectLink-g187529-d574612-i349532022-Museum_of_Natural_Science_Museo_de_Ciencias_Naturales-Valencia_Province_o.html

Is it possible to build an aircraft wing in LEGO® ?

Publié le 17 février 2020 Modifier l'article | Voir les stats

joseph morlier
Professor in Structural and Multidisciplinary Design Optimization, ... any idea?

5 articles

To review

https://github.com/jomorlier/mdo_ml_21

(MDO_ML 2021)

https://www.youtube.com/watch?v=geptVGimkYY&list=PL_TG7DdVYSp_fcwKATnwPRnJHGOyaiPPX&index=9&t=2443s

https://ica.cnrs.fr/author/jmorlier/

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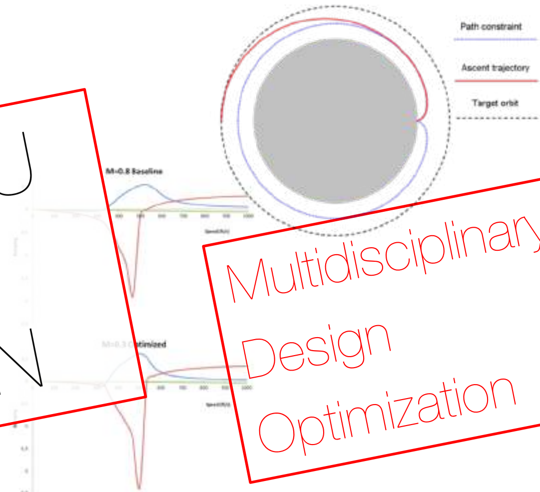
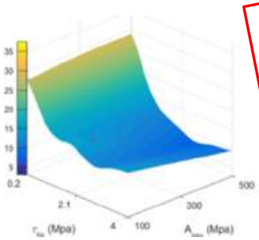
Structural Optimization & Ecodesign

AIRBUS

THANK YOU for Your ATTENTION

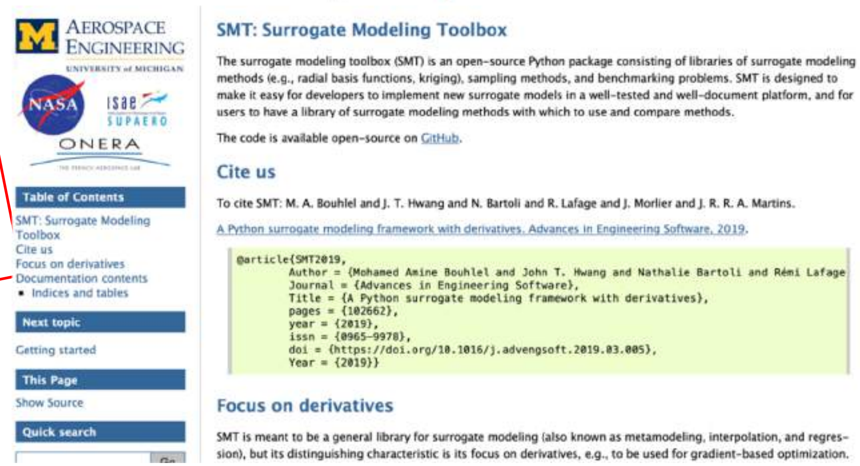


Multidisciplinary Design Optimization



#AI4E Artificial Intelligence For Engineers

https://github.com/SMTorg/smt



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