

MDO_ML_21

Multidisciplinary Optimization and Machine Learning for Engineering Design

19 July 2021 – 5 August 2021

<https://mdoml2021.ftmd.itb.ac.id/>

Jointly organized by



香港科技大學
THE HONG KONG
UNIVERSITY OF SCIENCE
AND TECHNOLOGY

Design for Additive Manufacturing:
Topology Optimization
Prof. Joseph Morlier



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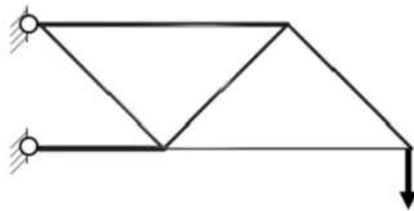
Part 2

Topology Optimization

Topology is the Key to Sustainability?



Truss "A"



Truss "B"

Strength $\frac{V_A}{V_B} = 27\% \text{ More}$

Deflection $\frac{V_A}{V_B} = 60\% \text{ More}$

History

- Homogenization of Microstructures was introduced by mathematics in the 1970s.
- First paper by Martin Bendsoe (Technical University of Denmark) and Noboru Kikuchi (University of Michigan) in 1988

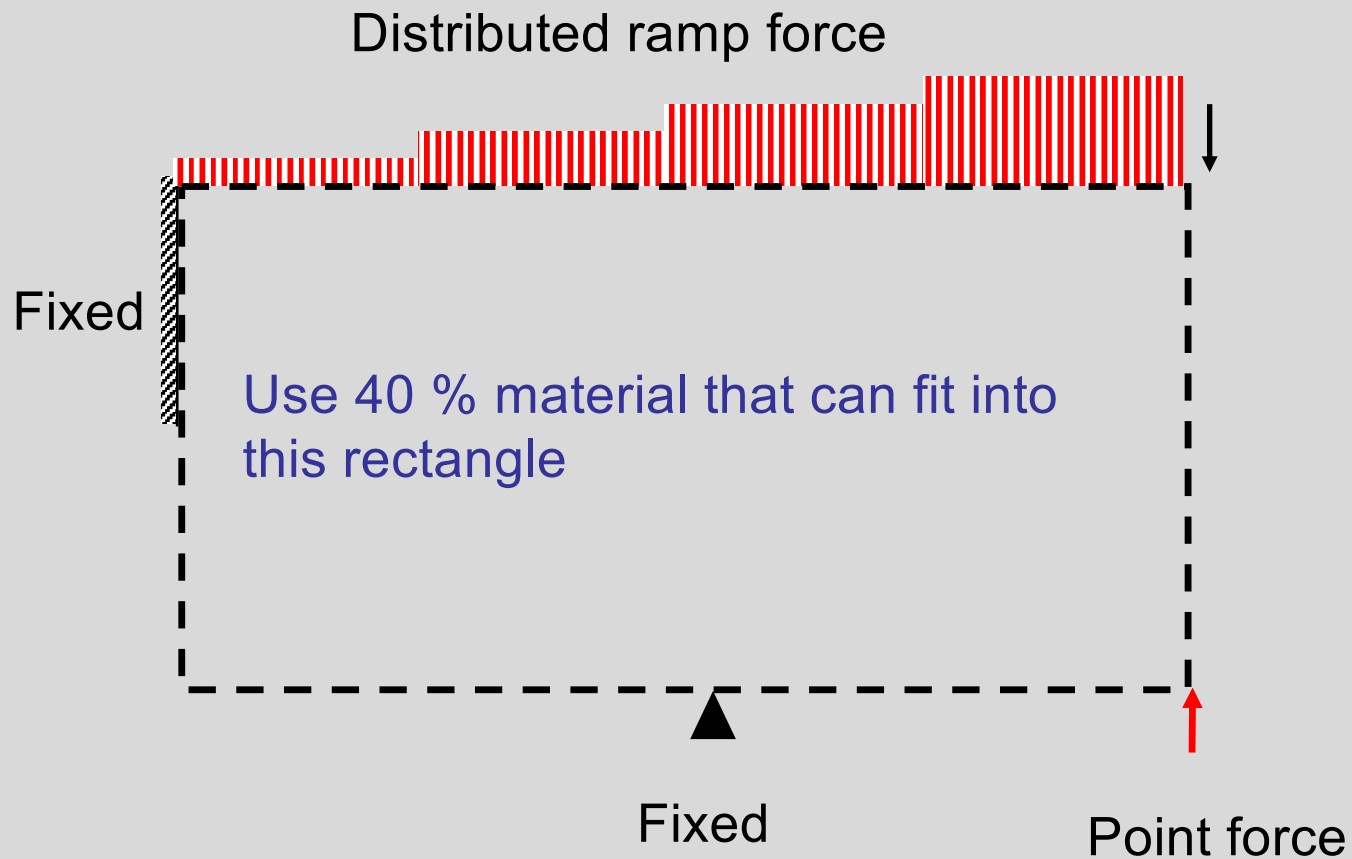
A topology optimisation problem can be written in the general form of an optimization problem as

$$\min_{\rho} F = F(\mathbf{u}(\rho), \rho) = \int_{\Omega} f(\mathbf{u}(\rho), \rho) dV$$

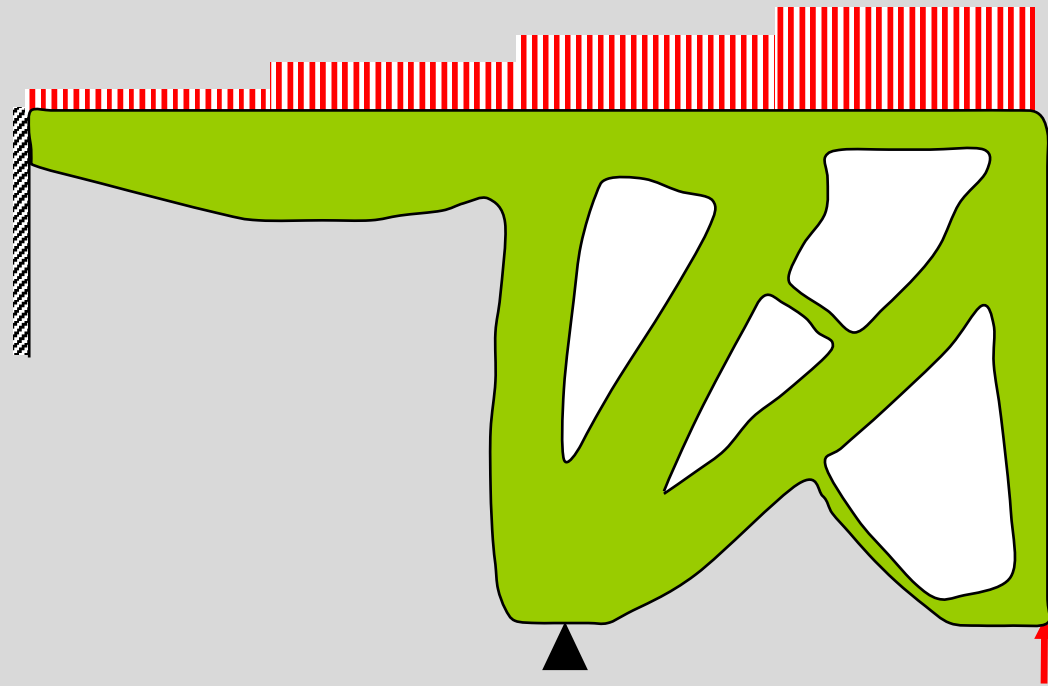
subject to

- $\rho \in \{0, 1\}$
- $G_0(\rho) = \int_{\Omega} \rho(\mathbf{u}) dV - V_0 \leq 0$
- $G_j(\mathbf{u}(\rho), \rho) \leq 0$ with $j = 1, \dots, m$

Stiff structure for your specifications



Stiff structure for your specifications



Well-Known example

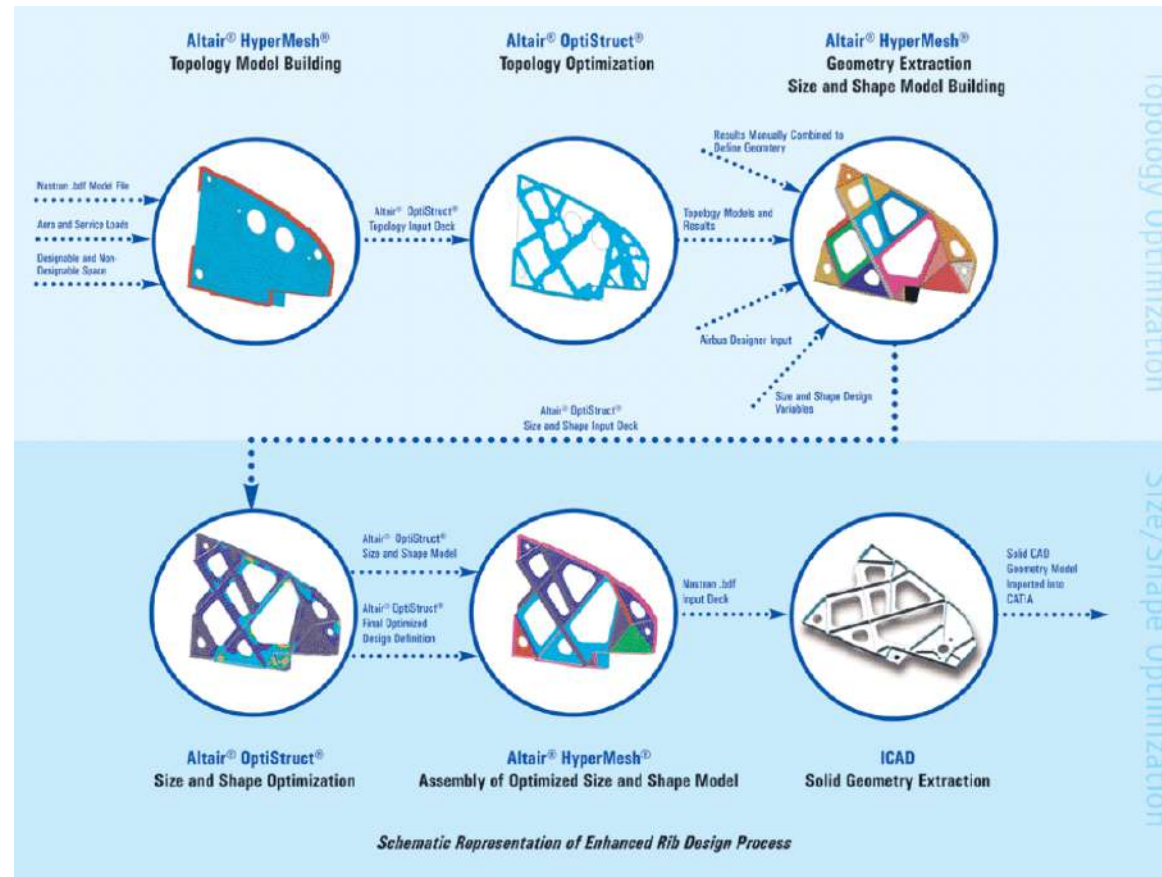
- Wing stiffening ribs of Airbus A380:



- Objective: reduce weight
- Constraints: stress, buckling

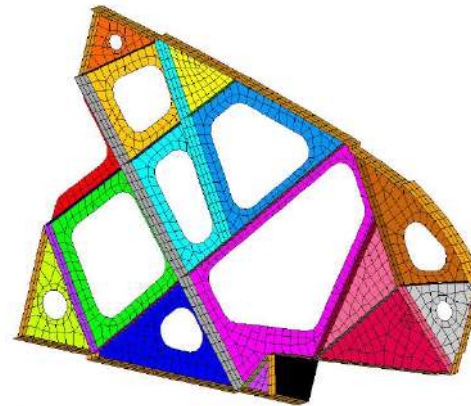
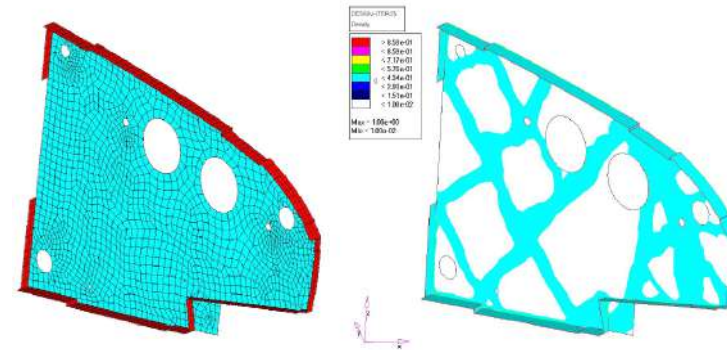
Topology and shape optimization

This process
will be
shortcutted in
Part4

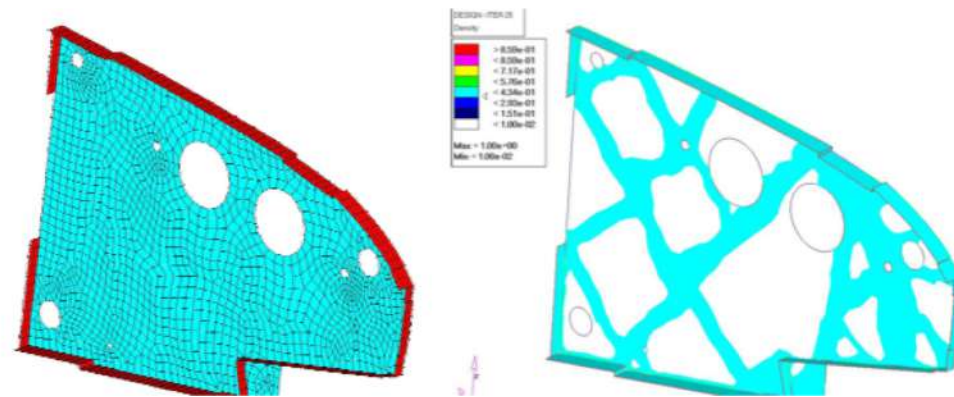


Airbus A380 example (cont.)

- Topology optimization:
- Sizing / shape optimization:

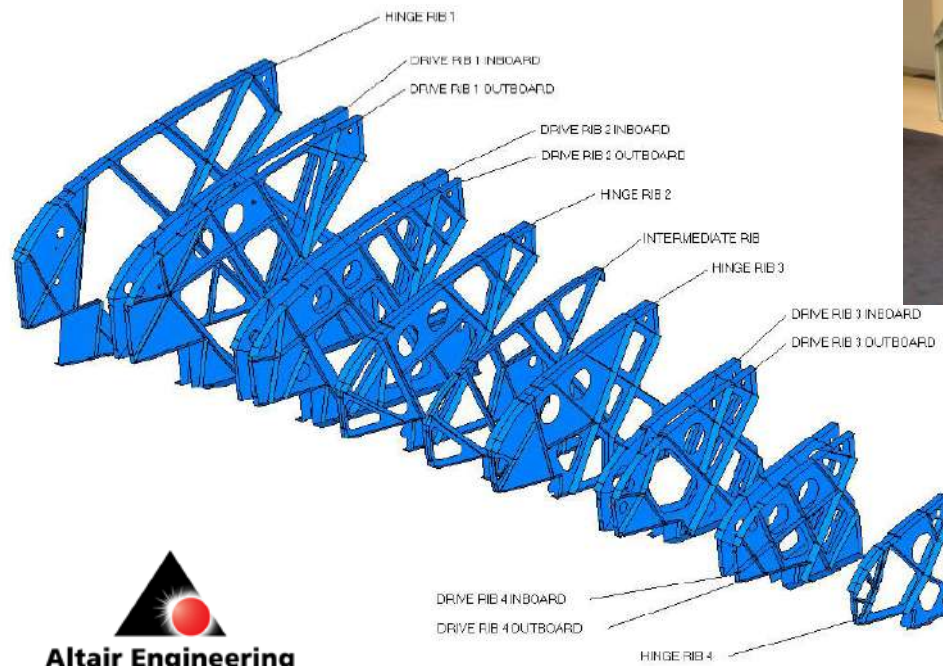


Finally...



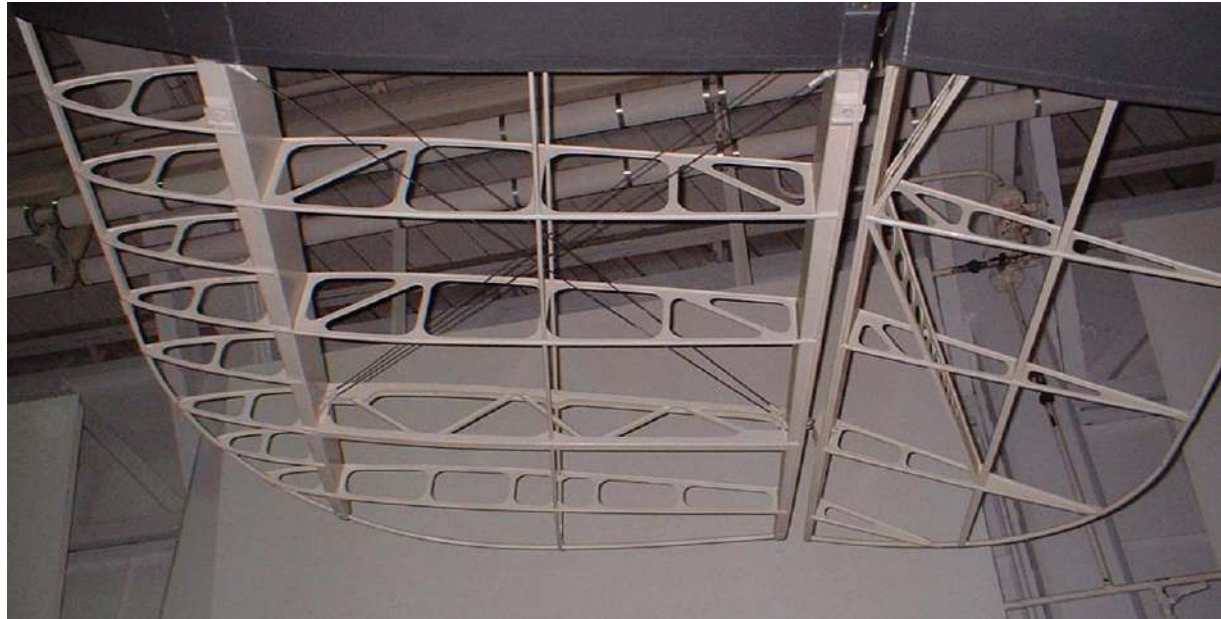
Airbus A380 example (cont.)

- Result: 500 kg weight savings!



Is this really a discovery?

Supermarine Southampton, 1925



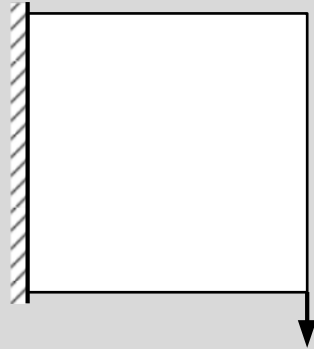
System approach automates the process!

Industrial problems

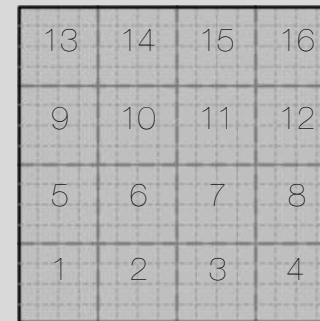
- TopOpt: Preliminary phases of a project
- The idea is to find the best path of stiffening in a given volume of matter.
- The mass is only found where it is needed, which is a good starting point for optimization of shape or dimensioning.
- The structure with the best static behavior.
- The paths of internal forces identified are those which help to rigidify the structure as well as possible

Inside the material ?→ DMO

Maximum stiffness in the plane of a plate by selecting the best orientations of fibers



Loads and boundary conditions



Design model with 4 * 4 patches

Table 4 Material properties

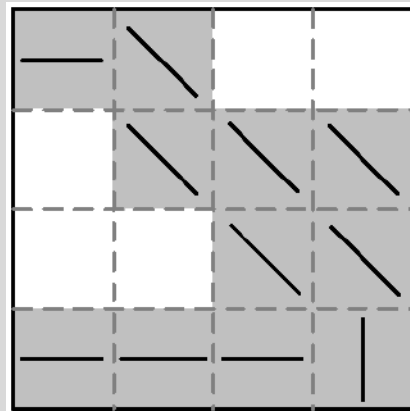
E_x	E_y	G_{xy}	ν_{xy}
146.86GPa	10.62GPa	5.45GPa	0.33

Table 3 Orientations

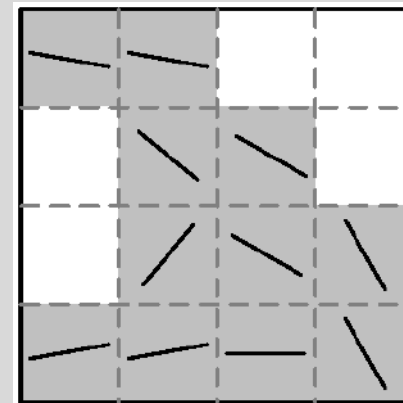
Number of material phases (m)	Number of design variables for each region (m_v)	Discrete orientation angle ($^\circ$)
4	2	90/45/0/-45
9	4	80/60/40/20/0/-20/-40/-60/-80
12	4	90/75/60/45/30/15/0/-15/-30/-45/-60/-75

Discrete Material Optimization: exemple

Topological optimization: vacuum + composite laminate
Volume constraints: $V < 11/16$



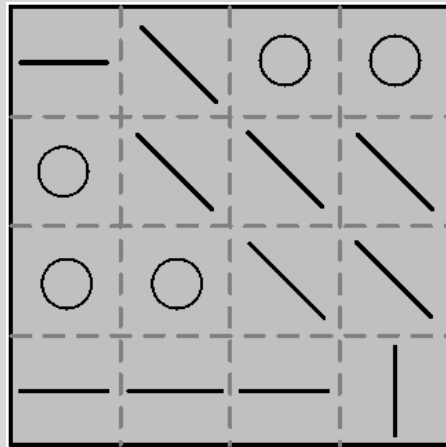
4 orientations
90/45/0/-45



18 orientations
90/80/70/60/50/40/30/20/10/0/
-10/-20/-30/-40/-50/-60/-70/-80

Discrete Material Optimization: exemple

Use of both glass fibers and foam
Limitation of the number of domains occupied by the fiber of glass

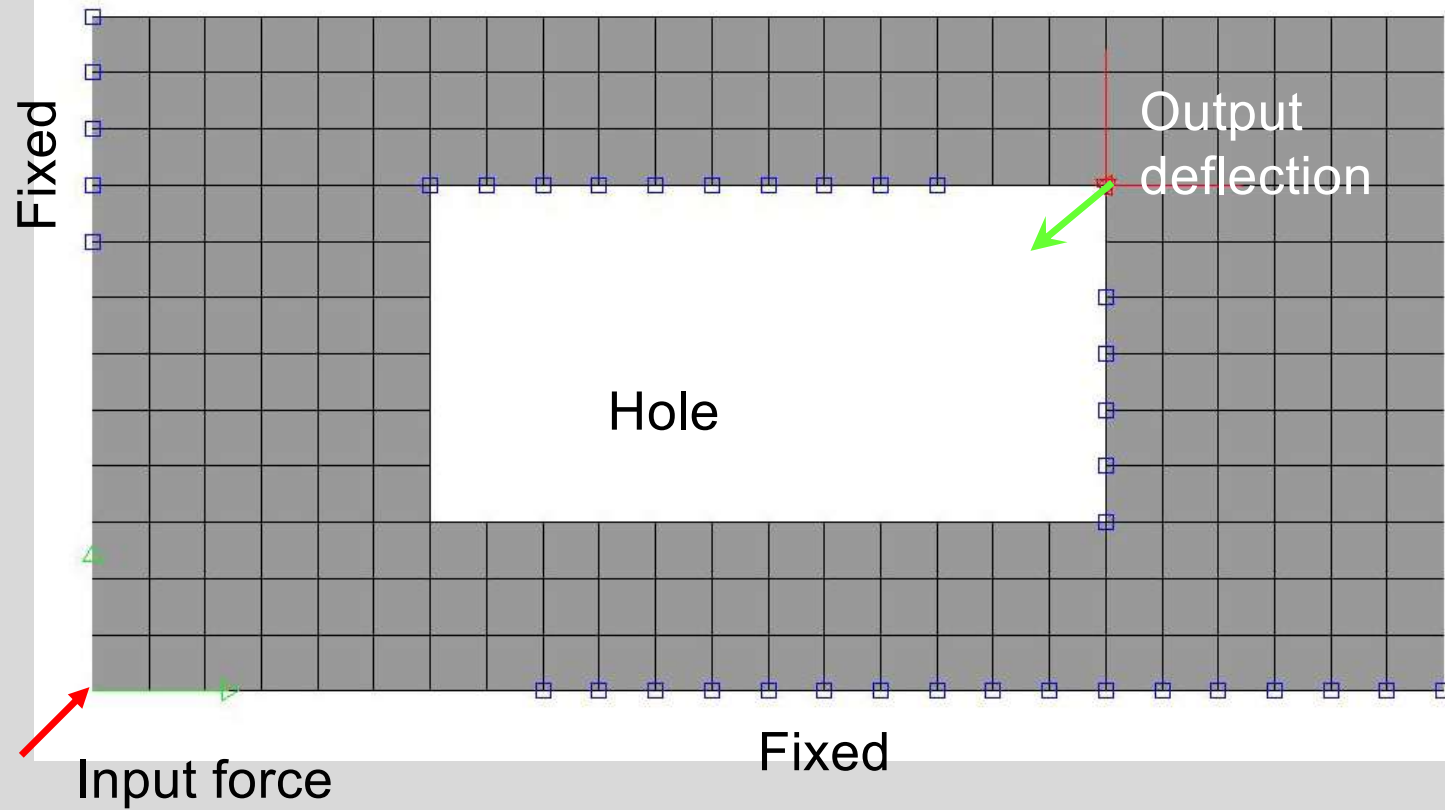


Optimization result of the square
plate under vertical force with
volume constraint

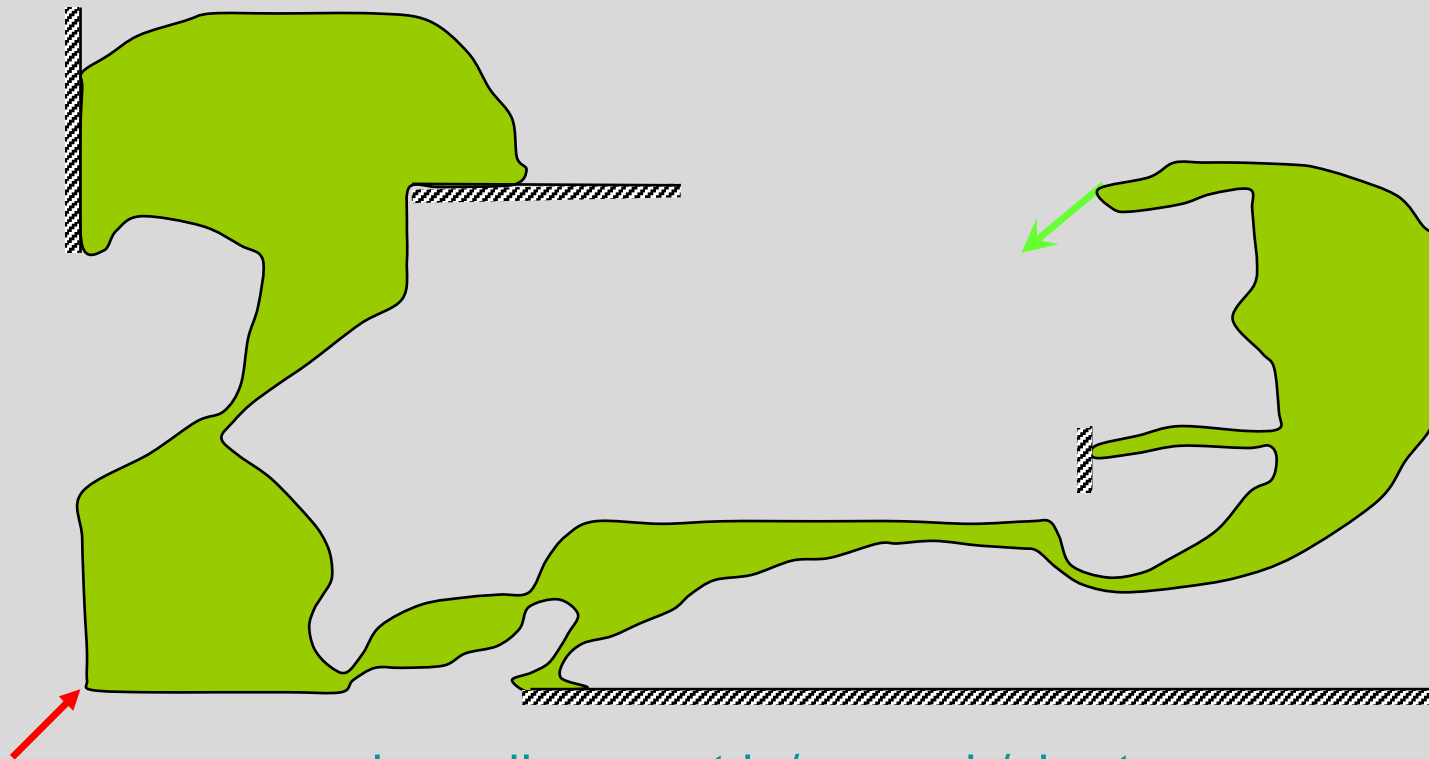
Glass-epoxy with 4 orientations
(90/45/0/-45) and polymer-foam

Compliant mechanism to your specifications

Use 30 % material



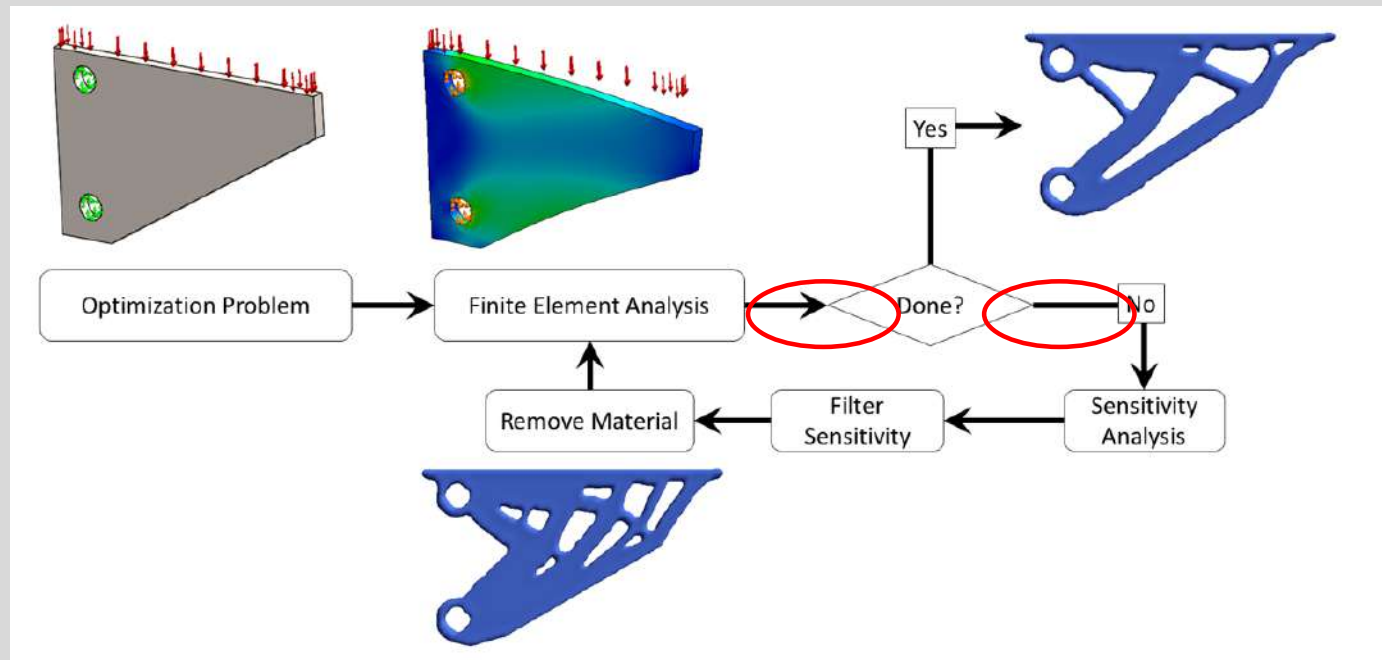
Compliant mechanism to your specifications



www.mecheng.iisc.ernet.in/~suresh/shortcourse

TopOpt relies on FEA

Online computation: <http://www.cloudtopopt.com>

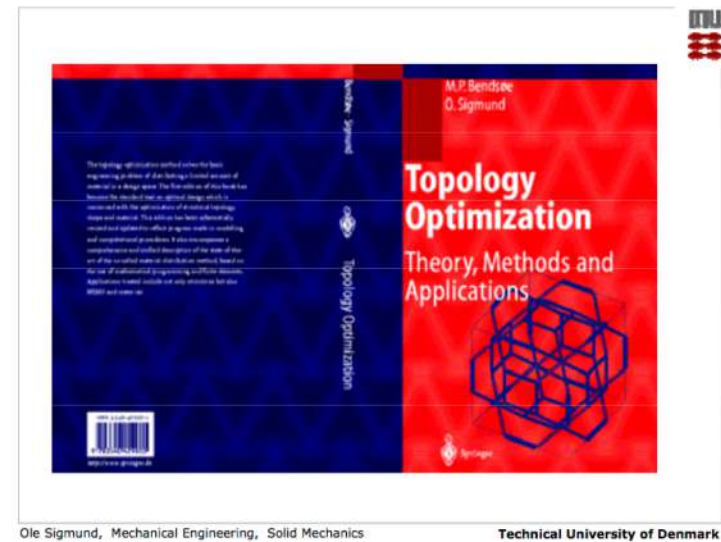
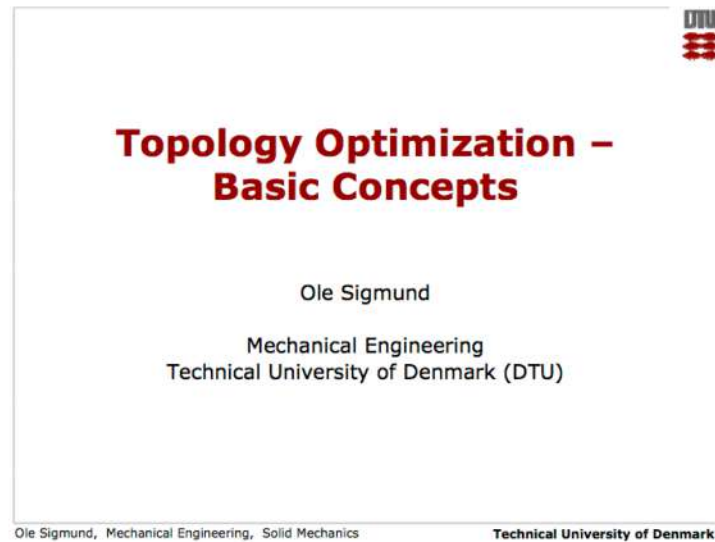


One pioneer, **SIMP** (Solid Isotropic Material with **Penalization**)

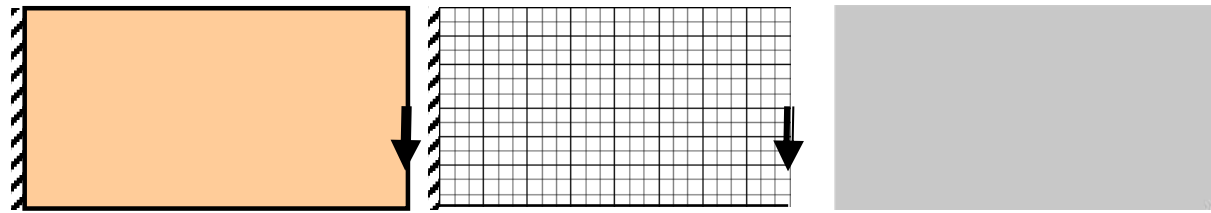
~~Homogenization~~

~~Level Set~~

~~Evolutionary~~



SIMP: Solid Isotropic Material with Penalization



Min Compliance

$$v = 0.5v_0$$

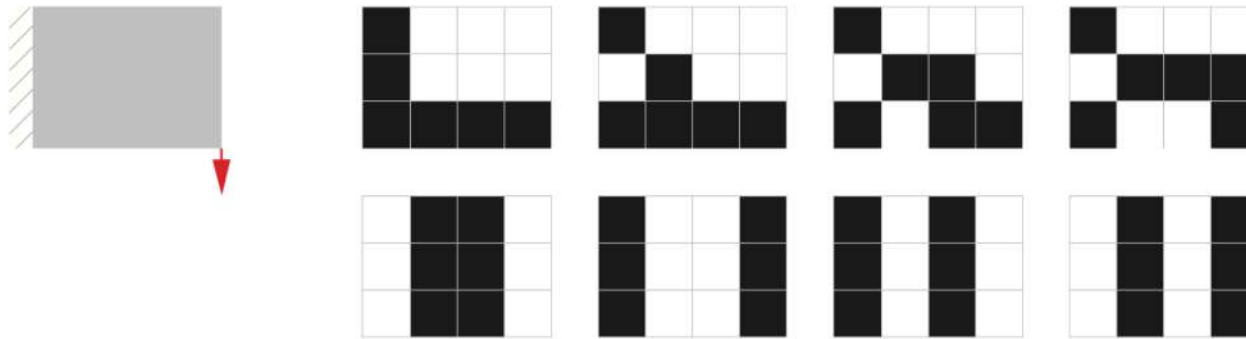
$0 < \rho_e \leq 1$: 'PseudoDensity'

Min Compliance

$$\sum_{\rho_e} \rho_e v_e = 0.5v_0$$

Pixels

- Finding a solution by checking all the possible combinations is impossible since the number of topologies nT increases exponentially with the number of finite elements n
- $nT = 2^n$,



The legal (top) and some illegal (bottom) topologies with 4 by 3 elements

Division into elements (pixels or voxels) and binary decision for each
or example 10,000 elements --> 210,000 possible configurations!

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Pixels?

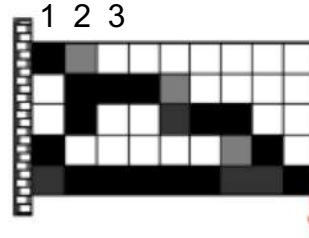
When the size of the FE model is increasing, the SIMP optimization problem is ... increasing



Chris Columbus et al, Pixels, movie 2015



Intuitive Problem? Quadratic Form



$$\begin{aligned} x_1 &= 1 \\ x_2 &= 0.5 \\ x_3 &= 0 \\ &\dots \end{aligned}$$

- Objective function; Strain energy

$$\min c(\mathbf{x}) = \mathbf{U}^T \mathbf{F} = \mathbf{U}^T \mathbf{K} \mathbf{U} \quad \text{with} \quad x_e = \frac{\rho_e}{\rho_0} \quad (4)$$

with $\mathbf{K} = \mathbf{K}_0 \sum_{e=1}^N x_e^p$ one can write:

$$\min c(\mathbf{x}) = \sum_{e=1}^N (x_e)^p \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e \quad \text{Scalar} \quad (5)$$

- Constraints: mass target

$$\frac{V(\mathbf{x})}{V_0} = f = \text{const} \Leftrightarrow \sum_{e=1}^N V_e x_e - V_0 f = 0 = h(\mathbf{x}) \quad \text{Scalar}$$

$$0 < \rho_{\min} \leq \rho_e \leq 1$$

$$\min c(\mathbf{x}) = \sum_{e=1}^N (x_e)^p \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e$$

Quadratic Form

$$\mathbf{x} \in \mathbb{R}^{m \times 1}, \mathbf{A} \in \mathbb{R}^{m \times m}$$

$$\text{Quadratic form : } \mathbf{x}^T \mathbf{A} \mathbf{x}$$

$\mathbf{x}^T \mathbf{A} \mathbf{x}$ is a scalar value.

$$\begin{array}{c} \downarrow \quad \searrow \quad \searrow \\ (1 \times m) \times (m \times m) \times (m \times 1) \rightarrow 1 \times 1 \end{array}$$

K is linked through E and x_e

Rozvany, G.I.N. , Zhou, M., and Gollub, M. (1989). Continuum Type Optimality Criteria Methods for Large Finite Element Systems with a Displacement Constraint, Part 1. *Structural Optimization* 1:47-72.

$$\mathbf{K} = \mathbf{K}_0 \sum_{e=1}^N x_e^p \quad x_e = \frac{\rho_e}{\rho_0}$$

What is p ,
(simP)????????

- **But HOW ??**

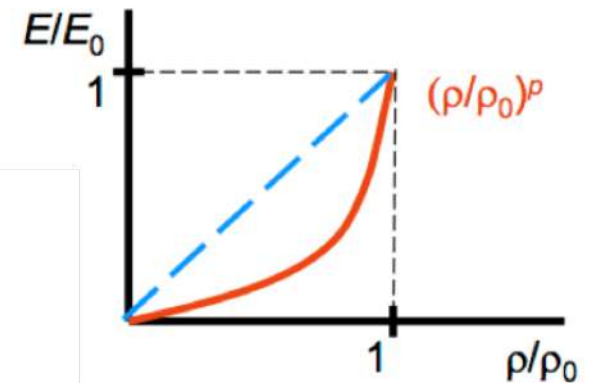
Avoid intermediate densities !

Solid Isotropic Material with Penalization (SIMP)

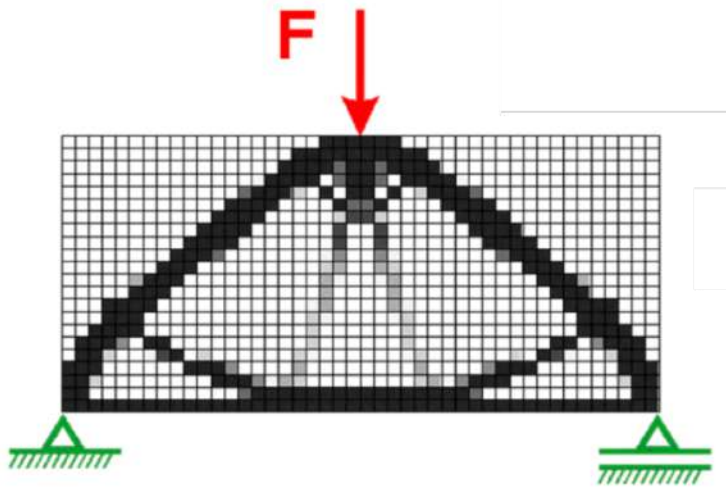
$$E(x) = E_{min} + (E_0 - E_{min})x^p$$

p is the penalty parameter to push densities to black (1) and white (0).

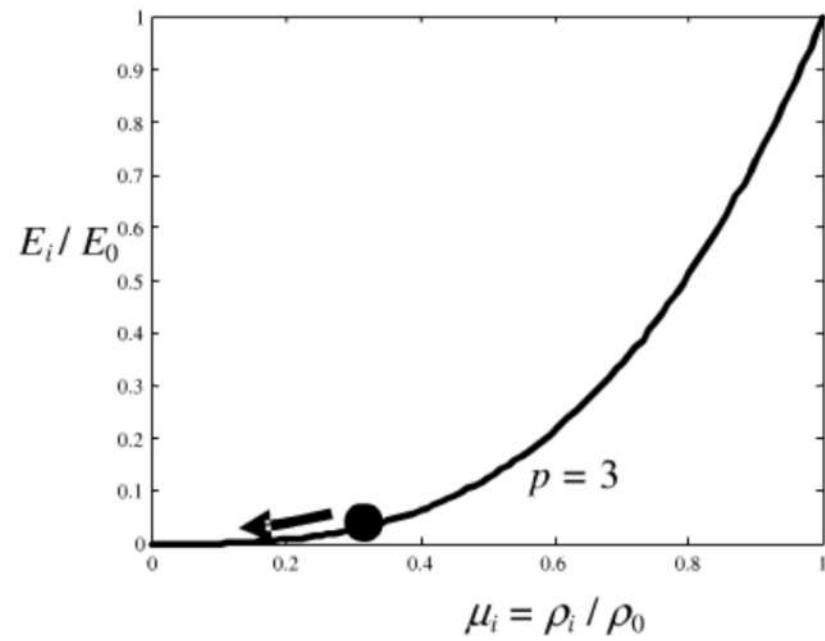
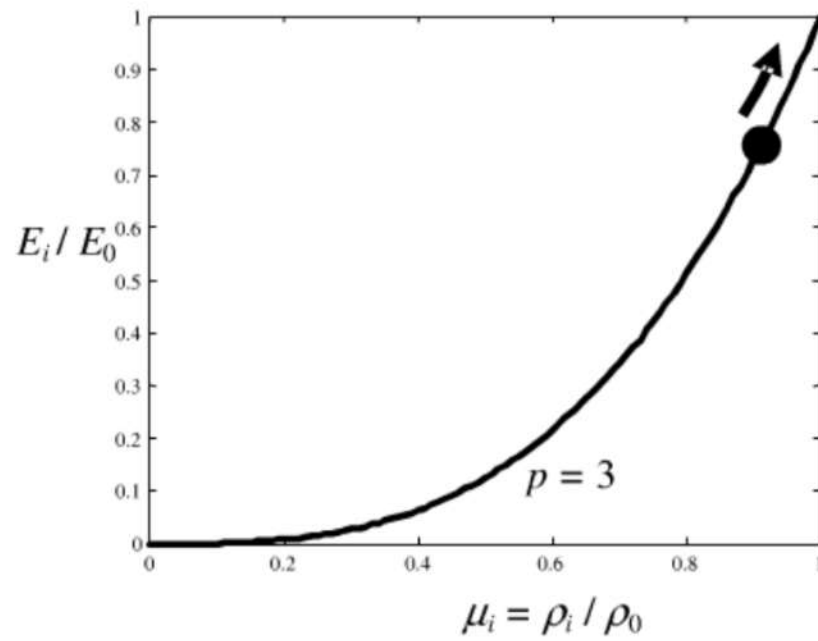
E_{min} is a small value that avoid stiffness matrix singularity



Penalization for altering stiffness locally



Let's take $p=3$



Penalty parameter in the SIMP method: Proof

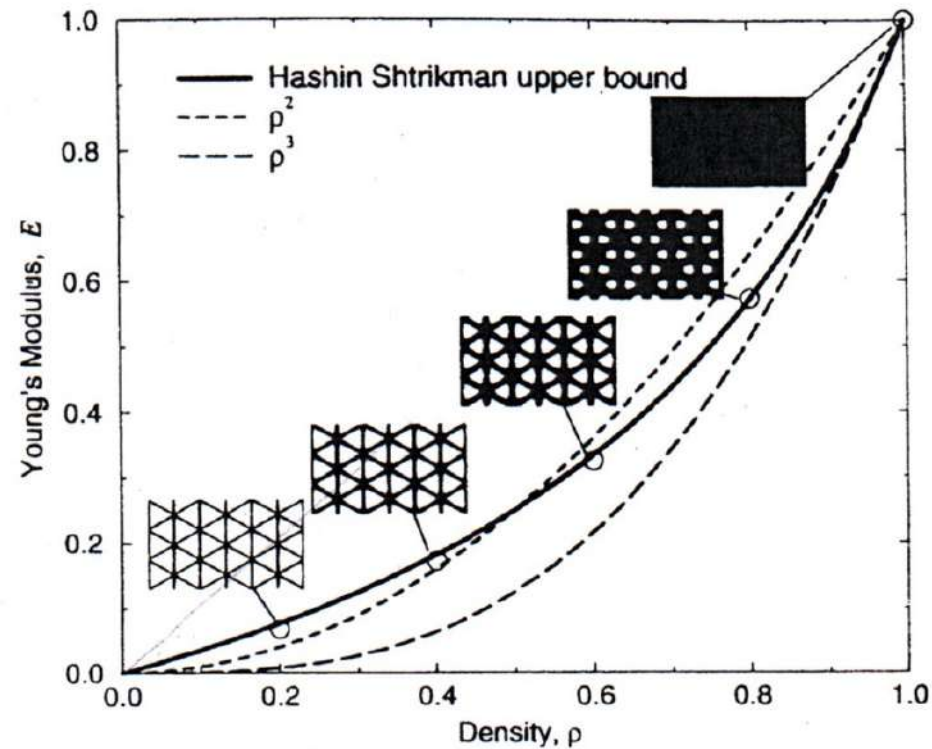
Hashin-Shtrikman bounds

$$0 \leq E \leq \frac{\rho E^0}{3 - 2\rho}$$

Therefore,

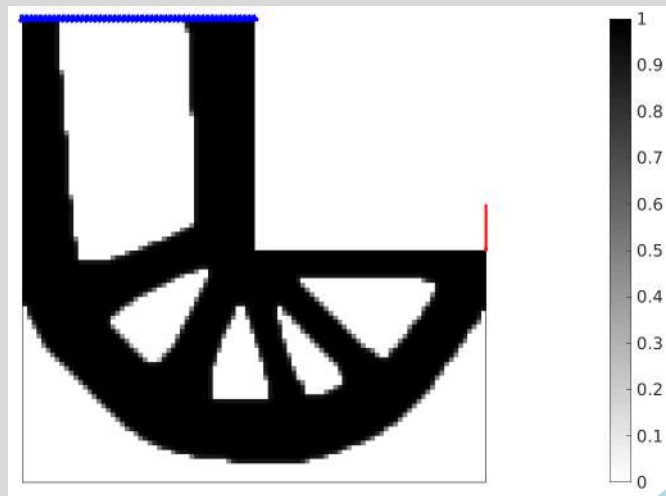
$$\rho^p E^0 \leq \frac{\rho E^0}{3 - 2\rho}$$

$$\Rightarrow p \geq 3$$

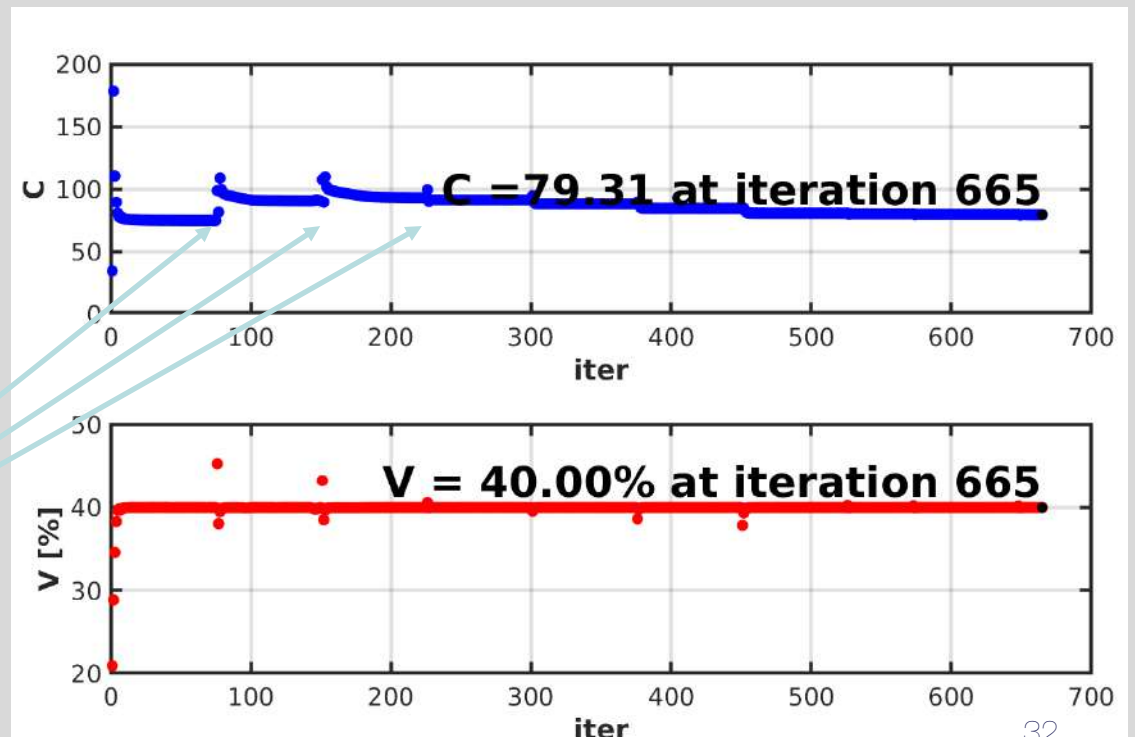


Bendsøe, M.P. and Sigmund, O., "Material Interpolation Schemes in Topology Optimization," *Archives in Applied Mechanics*, Vol. 69, (9-10), 1999, pp. 635-654.

Penalty parameter in the SIMP method: Continuation Methods



Increase p
Reduce the filter radius
Increase β



REALLY Nice idea !

1. Transform discrete variables continuously (TO USE gradient-based algorithms)
2. Find an objective function with "cheap" derivatives (we will see this later)

Others formulations

$$\begin{aligned} \min_{\boldsymbol{\mu}} \max_{l=1,\dots,nc} C_l &= \mathbf{F}_l^T \mathbf{q}_l \\ \sum_i \mu_i V_i &\leq \bar{V} \\ 0 < \underline{\mu}_i \leq \mu_i &\leq 1 \end{aligned}$$

- If several load cases nc
- we can minimize the maximal compliance
- with \mathbf{q}_l obtained by solving $\mathbf{K}\mathbf{q}_l = \mathbf{F}_l$

$$\begin{aligned} \min_{\boldsymbol{\mu}} \sum_i \mu_i V_i \\ q_j \leq \bar{q}_j \quad j=1,\dots,m \\ 0 < \underline{\mu}_i \leq \mu_i \leq 1 \end{aligned}$$

- Prescribed displacement
- we can minimize the volume (mass)
- wrt amplitude at node j inferior to a certain displacement

Others formulations

$$\begin{aligned} & \max_{\underline{\mu}} \min_{k=1, \dots, n_f} \omega_k \\ & \sum_i \mu_i V_i \leq \bar{V} \\ & 0 < \underline{\mu}_i \leq \mu_i \leq 1 \end{aligned}$$

- Eigensolver to obtain the stiffest structure at a certain volfrac

→ wrt a vibration ccriteria

linear buckling, thermal, thermoelasticity etc

This is not Sauron's eye

DOI. 10.1137/070699822

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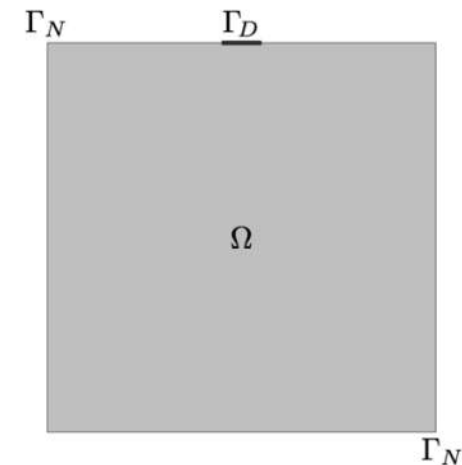
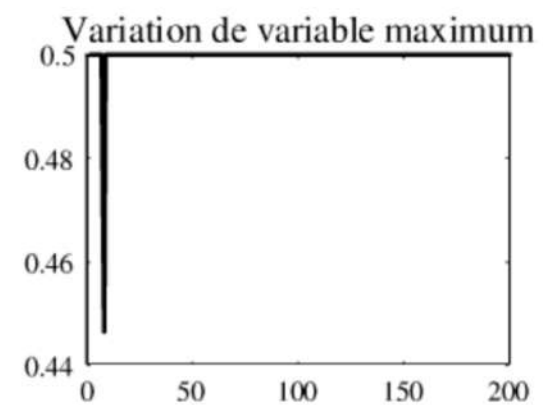
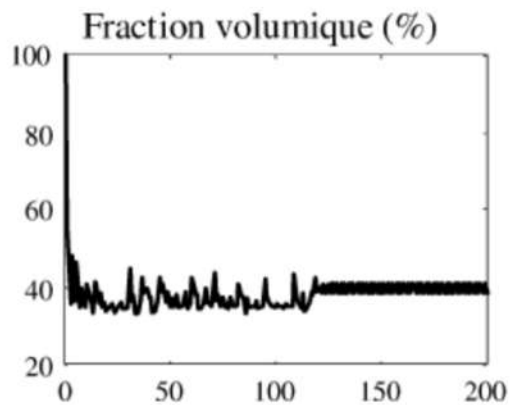
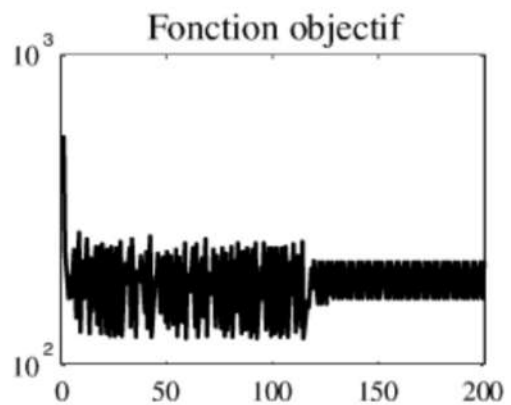
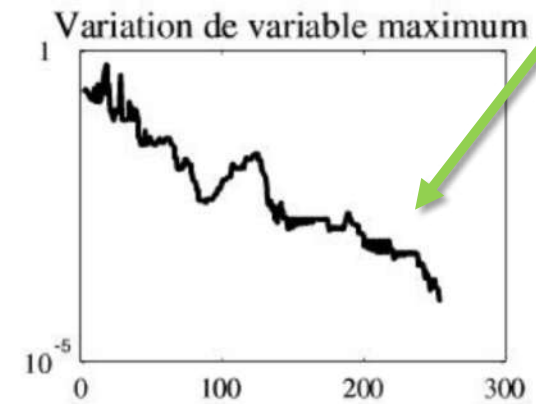
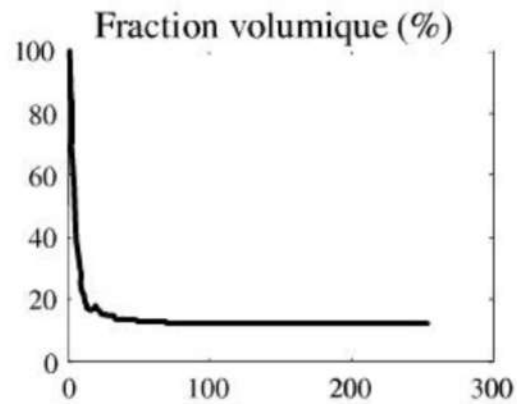
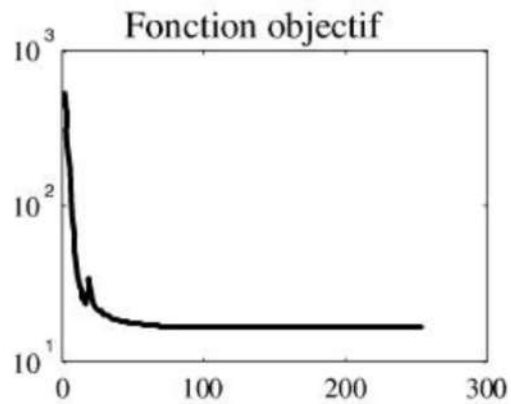


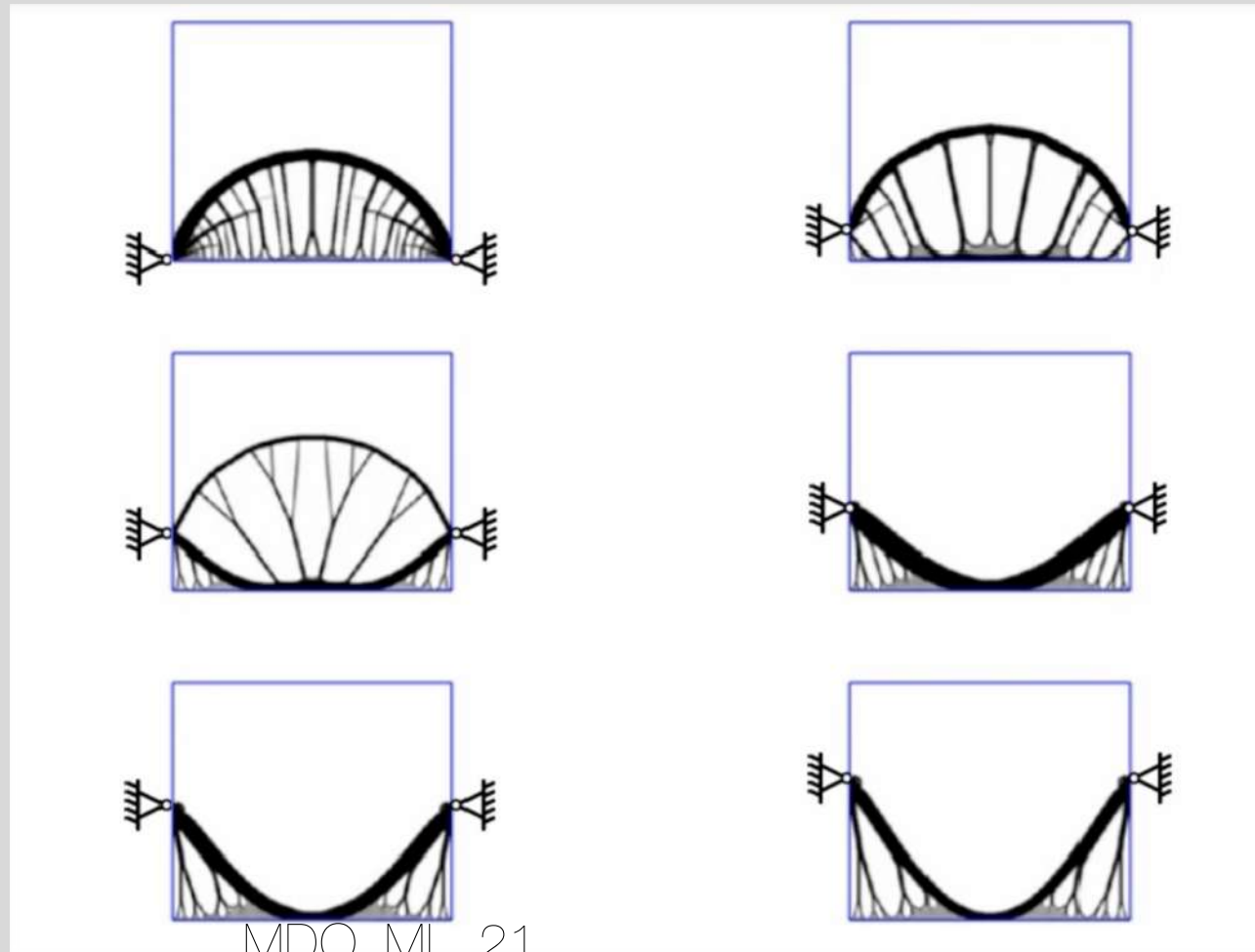
Fig. 2.1 The problems consist of finding the distribution within Ω of two materials with different heat conduction properties in order to obtain a temperature field that is as even as possible.

Which is the best optimizer? why ?



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Small changes in BCs ...



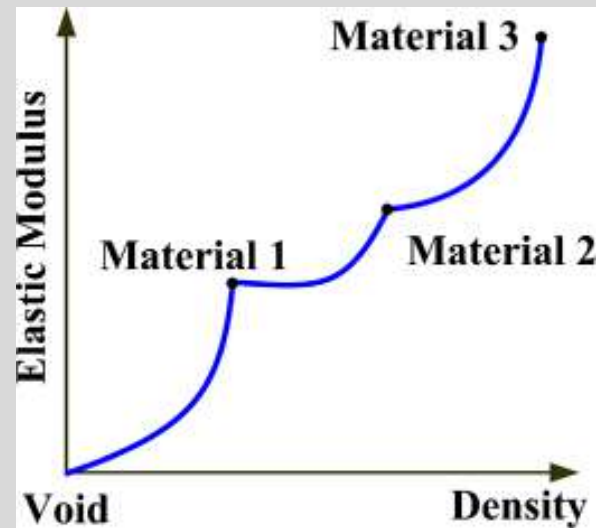
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MULTIMATERIALS

- Solid Isotropic Material with Penalization (SIMP)

- $E_e(\rho_e) = A_E * \rho_e^p + B_E$,

$$\rho_e \in [\rho_i, \rho_{i+1}] , \quad A_E = \frac{E_i - E_{i+1}}{\rho_i^p - \rho_{i+1}^p} , \quad B_E = E_i - A_E * \rho_i^p$$



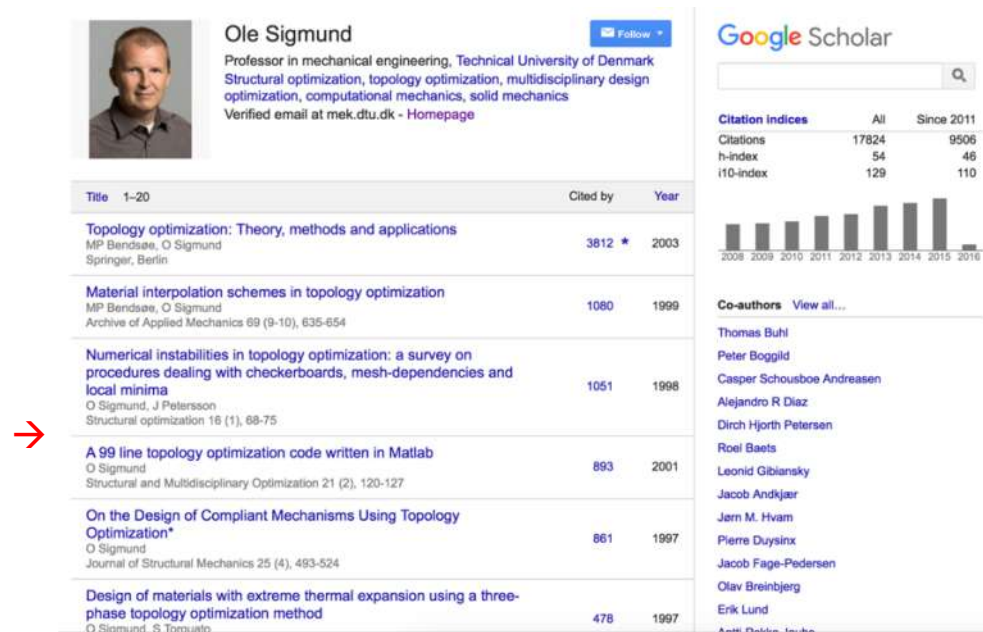
Zuo, W., & Saitou, K. (2016). Multi-material topology optimization using ordered SIMP interpolation. *Structural and Multidisciplinary Optimization*, 55(2), 477-491. doi:10.1007/s00158-016-1513-3

BUT ...IN PRACTICE?

Educational article:

O. Sigmund , A 99 line topology optimization code written in Matlab Struct Multidisc Optim 21, 120–127 Springer-Verlag 2001

Heuristic formulation (intuitive method of optimisation, but with no convergency proofs) to update x_e by bi-section algorithm



History (1988, Bendsoe)

A topology optimization problem based on the power-law approach, where the objective is to minimize compliance can be written as

$$\left. \begin{array}{l} \min_{\mathbf{x}}: \quad c(\mathbf{x}) = \mathbf{U}^T \mathbf{K} \mathbf{U} = \sum_{e=1}^N (x_e)^p \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e \\ \text{subject to:} \quad \frac{V(\mathbf{x})}{V_0} = f \\ \quad \quad \quad : \quad \mathbf{K} \mathbf{U} = \mathbf{F} \\ \quad \quad \quad : \quad \mathbf{0} < \mathbf{x}_{\min} \leq \mathbf{x} \leq \mathbf{1} \end{array} \right\}, \quad (1)$$

where \mathbf{U} and \mathbf{F} are the global displacement and force vectors, respectively, \mathbf{K} is the global stiffness matrix, \mathbf{u}_e and \mathbf{k}_e are the element displacement vector and stiffness matrix, respectively, \mathbf{x} is the vector of design variables, \mathbf{x}_{\min} is a vector of minimum relative densities (non-zero to avoid singularity), N ($= \mathbf{nelx} \times \mathbf{nely}$) is the number of elements used to discretize the design domain, p is the penalization power (typically $p = 3$), $V(\mathbf{x})$ and V_0 is the material volume and design domain volume, respectively and f (`volfrac`) is the prescribed volume fraction.

Compliance minimization self adjoint

- Compliance is the opposite of stiffness

$$C = \mathbf{f}^T \mathbf{u} = \mathbf{u}^T K \mathbf{u}$$

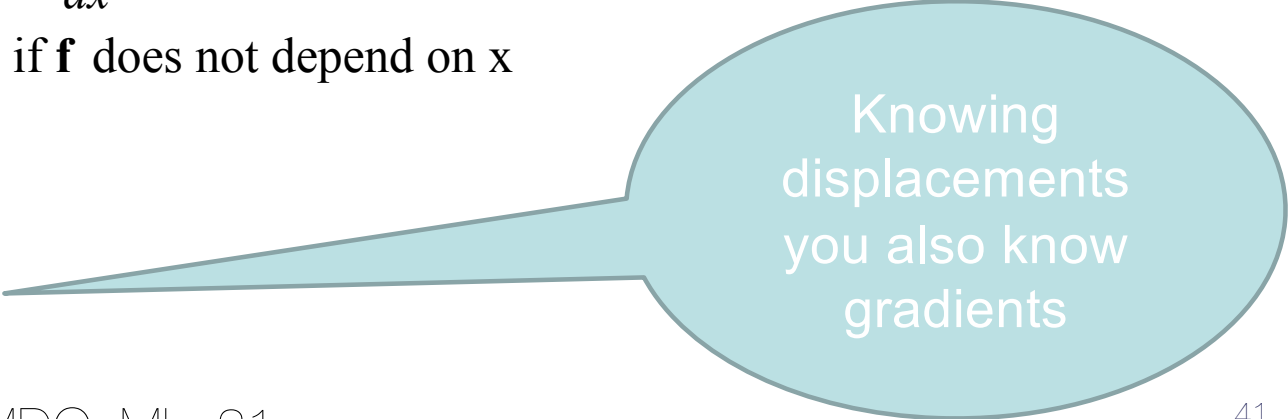
- Inexpensive derivatives (use chain rule)

$$\frac{dC}{dx} = 2\mathbf{u}^T K \frac{d\mathbf{u}}{dx} + \mathbf{u}^T \frac{dK}{dx} \mathbf{u}$$

But since $K\mathbf{u} = \mathbf{f}$ if \mathbf{f} does not depend on x

$$K \frac{d\mathbf{u}}{dx} = -\frac{dK}{dx} \mathbf{u}$$

$$\frac{dC}{dx} = -\mathbf{u}^T \frac{dK}{dx} \mathbf{u}$$



Knowing
displacements
you also know
gradients

Density design variables

Need a DEMO ? see next slides

- Recall $\frac{dC}{dx} = -\mathbf{u}^T \frac{dK}{dx} \mathbf{u}$

- For density variables

$$\frac{dC}{d\rho^e} \propto -\mathbf{u}^T \rho^{p-1} K^e \mathbf{u}$$

- Want to increase density of elements with high strain energy and vice versa
- To minimize compliance for given weight can use an optimality criterion method.

$$C = f^T u$$

f is not dependent on x

$$\frac{\partial C}{\partial x} = \frac{\partial f^T}{\partial x} u + f^T \frac{\partial u}{\partial x} = 0 + f^T \frac{\partial u}{\partial x} \quad (1)$$

starting by expressing ^{the derivative of} $K u = f$

$$\text{it gives } \frac{\partial K}{\partial x} u + K \frac{\partial u}{\partial x} = \frac{\partial f}{\partial x} = 0$$

$$\Leftrightarrow \boxed{K \frac{\partial u}{\partial x} = - \frac{\partial K}{\partial x} u}$$

$$(1) \quad \frac{\partial C}{\partial x} = f^T \frac{\partial u}{\partial x} = u^T \left(K \frac{\partial u}{\partial x} \right) = u^T \left(- \frac{\partial K}{\partial x} u \right) = - u^T \frac{\partial K}{\partial x} u = \frac{\partial C}{\partial x}$$

or by expressing $C = f^T u$

as $C = u^T K u$ (quadratic form) (2)

$$dC = \frac{\partial C}{\partial u} du + \frac{\partial C}{\partial x} dx \Rightarrow \boxed{\frac{dC}{dx}} = \frac{\partial C}{\partial u} \frac{du}{dx} + \frac{\partial C}{\partial x}$$

from (2) $\frac{\partial C}{\partial u} = 2u^T K \frac{du}{dx}$ \rightarrow

(Recap $\frac{\partial u^T K u}{\partial u} = 2u^T K$)

it comes

$$\begin{aligned} \frac{dC}{dx} &= 2u^T \left(K \frac{du}{dx} \right) + u^T \frac{\partial K}{\partial x} u \\ &= -2u^T \frac{\partial K}{\partial x} u + u^T \frac{\partial K}{\partial x} u = \end{aligned}$$

$$\boxed{-u^T \frac{\partial K}{\partial x} u} = \frac{\partial C}{\partial x}$$

Matlab Code (the tutorial will be in python)

```

x(1:nely,1:nelx) = volfrac; % INITIALIZE

loop = 0; change = 1.;
while change > 0.01 % START ITERATION While Xk+1>>Xk
    loop = loop + 1;
    xold = x;
    [U]=FE(nelx,nely,x,penal); % FE-ANALYSIS
    [KE] = lk;
    c = 0.;

    for ely = 1:nely
        for elx = 1:nelx
            n1 = (nely+1)*(elx-1)+ely;
            n2 = (nely+1)* elx +ely;
            Ue = U([2*n1-1;2*n1; 2*n2-1;2*n2; 2*n2+1;2*n2+2; 2*n1+1;2*n1+2],1);
            c = c + x(ely,elx)^penal*Ue'*KE*Ue; % OBJECTIVE FUNCTION
            dc(ely,elx) = -penal*x(ely,elx)^(penal-1)*Ue'*KE*Ue; % SENSITIVITY ANALYSIS
        end
    end
end

```

Sensitivity

$$\frac{\partial c}{\partial x_e} = -p(x_e)^{p-1} u_e^T k_0 u_e$$

Update rule

► OPTIMALITY CRITERIA METHOD

$$B_e = \frac{-\frac{\partial c}{\partial x_e}}{\lambda \frac{\partial V}{\partial x_e}}$$

$$\begin{cases} \max(x_{\min}, x_e - m) & \text{if } x_e B_e^\eta \leq \max(x_{\min}, x_e - m), \\ x_e B_e^\eta & \text{if } \max(x_{\min}, x_e - m) < x_e B_e^\eta < \min(1, x_e + m) \\ \min(1, x_e + m) & \text{if } \min(1, x_e + m) \leq x_e B_e^\eta, \end{cases}$$

Element Stiffness Matrix

```
function [KE]=lk
%Element Stiffness Matrix

E = 1.;
nu =1/3.;
k=[ 1/2-nu/6 1/8+nu/8 -1/4+nu/12 -1/8-nu/8
   -1/4-nu/12 -1/8+3*nu/8 ... nu/6 1/8-3*nu/8];
KE = E/(1-nu^2)* ...
[ k(1) k(2) k(3) k(4) k(5) k(6) k(7) k(8)
  k(2) k(1) k(8) k(7) k(6) k(5) k(4) k(3) k(3) k(8) k(1) k(6)
  k(7) k(4) k(5) k(2) k(4) k(7) k(6) k(1) k(8) k(3) k(2) k(5)
  k(5) k(6) k(7) k(8) k(1) k(2) k(3) k(4) k(6) k(5) k(4) k(3)
  k(2) k(1) k(8) k(7) k(7) k(4) k(5) k(2) k(3) k(8) k(1) k(6)
  k(8) k(3) k(2) k(5) k(4) k(7) k(6) k(1)];
```

FEM Analysis (2D mesh is invariant wrt to homotheties)

```
function [U]=FE(nelx,nely,x,penal)
[KE] = lk;
K = sparse(2*(nelx+1)*(nely+1), 2*(nelx+1)*(nely+1));
F = sparse(2*(nely+1)*(nelx+1),1); U = zeros(2*(nely+1)*(nelx+1),1);
for elx = 1:nelx
    for ely = 1:nely
        n1 = (nely+1)*(elx-1)+ely;
        n2 = (nely+1)* elx +ely;
        edof = [2*n1-1; 2*n1; 2*n2-1; 2*n2; 2*n2+1; 2*n2+2; 2*n1+1; 2*n1+2];
        K(edof,edof) = K(edof,edof) + x(ely,elx)^penal*KE;
    end
end
F(2*(nelx+1)*(nely+1),1)=-1;
fixeddofs=union([1,2],[2*nely+1:2*(nely+1)]);
alldofs = [1:2*(nely+1)*(nelx+1)];
freedofs = setdiff(alldofs,fixeddofs);
% SOLVING
U(freedofs,:) = K(freedofs,freedofs) \ F(freedofs,:);
U(fixeddofs,:)= 0;
```

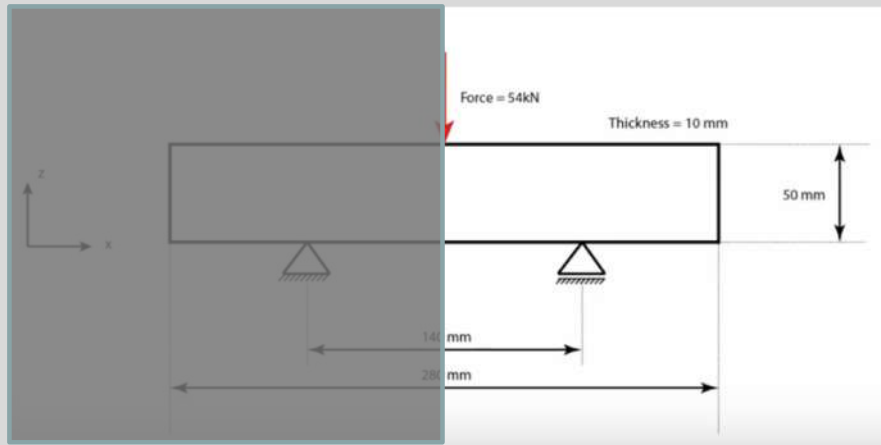
TO HAVE REAL DISPLACEMENT (see tutorial on github)

- 1) Choose consistent units **N, mm, MPa** for example (*Remember Nastran Course*)
- 2) Put the real Young's modulus $E=210 \times 10^3$ **MPa** for example;
- 3) Multiply the unit load by true amplitude F for example 54×10^3 **N** ;
- 4) Multiply the elementary stiffness matrix by the thickness (**mm**)
- 5) 2D mesh is invariant wrt to homotheties ; Need to check that n_{elx} and n_{ely} are related to the true value for example **140 and 50 mm**
- 6) Apply the BCs

The compliance unit is **mJ**.

An engineering example (tutorial available on Github)

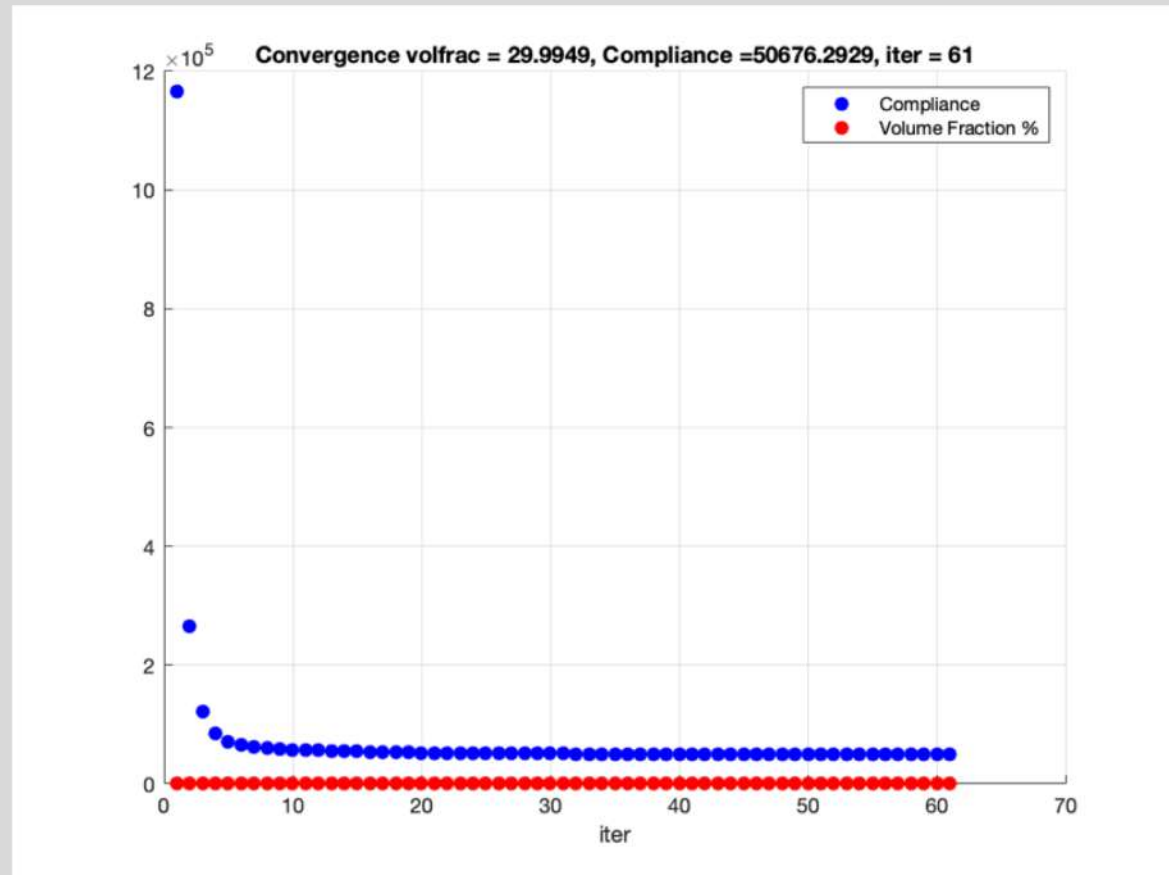
- Search the optimal 2D topology using symmetry
--> modify top88.m



`top88_ptBENDING(140, 50, 0.3, 3, 2, 2)`

MDO_ML_21

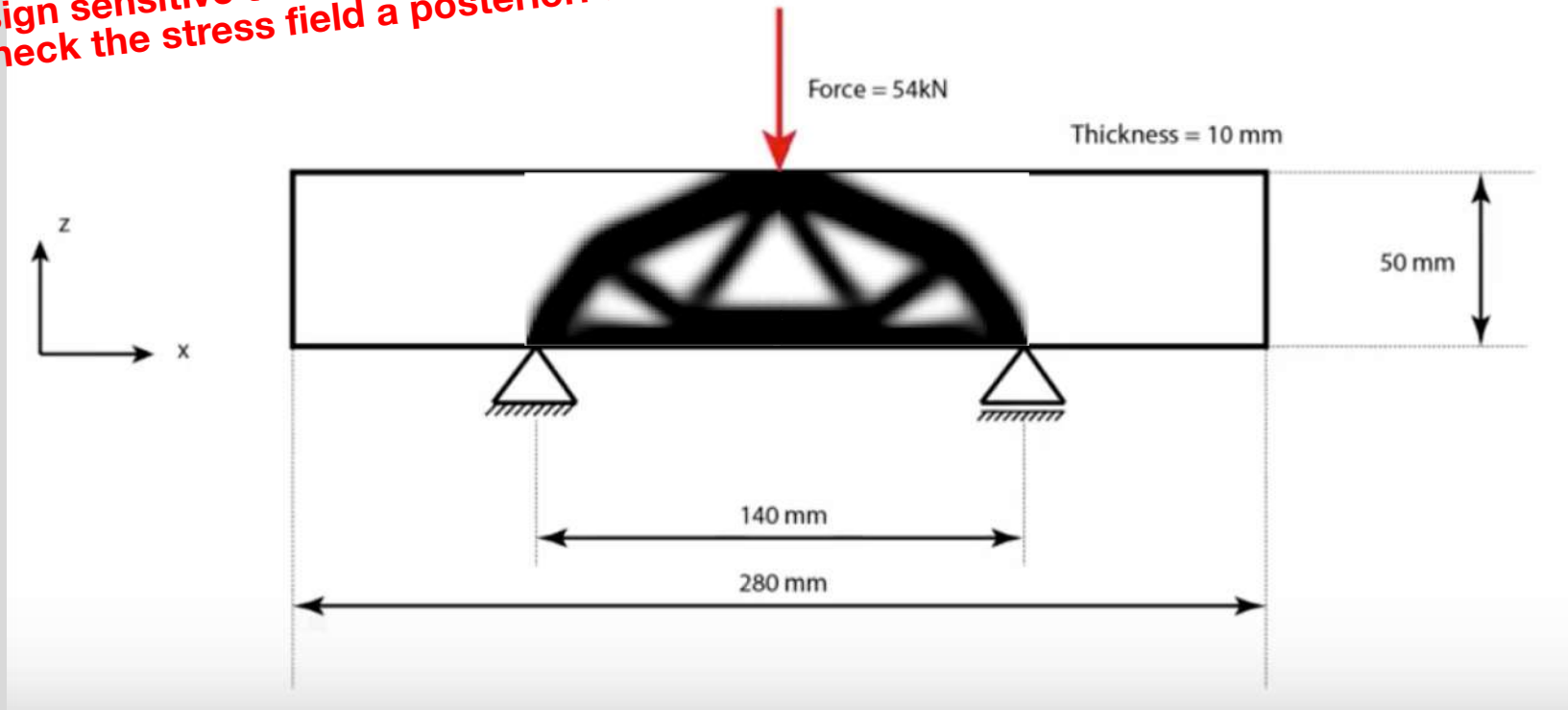
2 Outputs, X^* + feasibility/optimality graph



MDO_ML_21

Outputs

Need to add constraint max displacement ?
Need to produce a Pareto front design vs volfrac?
How do you experimentally introduce the external force?
Is the design sensitive to BCs size ?
Do you check the stress field a posteriori ?



MDO ML_21

```

1 %%% AN 88 LINE TOPOLOGY OPTIMIZATION CODE Nov, 2010 %%%
2 function x=top88(nelx,nely,volfrac,penal,rmin,ft)
-
3 %% MATERIAL PROPERTIES
4 E0 = 1;
5 Emin = 1e-9;
6 nu = 0.3;
-
7 %% PREPARE FINITE ELEMENT ANALYSIS
8 A11 = [12 3 -6 -3; 3 12 3 0; -6 3 12 -3; -3 0 -3 12];
9 A12 = [-6 -3 0 3; -3 -6 -3 -6; 0 -3 -6 3; 3 -6 3 -6];
10 B11 = [-4 3 -2 9; 3 -4 -9 4; -2 -9 -4 -3; 9 4 -3 -4];
11 B12 = [ 2 -3 4 -9; -3 2 9 -2; 4 9 2 3; -9 -2 3 2];
12 KE = 1/(1-nu^2)/24*([A11 A12;A12' A11]+nu*[B11 B12;B12' B11]);
13 nodenrs = reshape(1:(1+nelx)*(1+nely),1+nely,1+nelx);
14 edofVec = reshape(2*nodenrs(1:end-1,1:end-1)+1,nelx*nely,1);
15 edofMat = repmat(edofVec,1,8)+repmat([0 1 2*nely+[2 3 0 1] -2 -1],nelx*nely,1);
16 iK = reshape(kron(edofMat,ones(8,1))',64*nelx*nely,1);
17 jK = reshape(kron(edofMat,ones(1,8))',64*nelx*nely,1);
18 % DEFINE LOADS AND SUPPORTS (HALF MBB-BEAM)
19 F = sparse(2,1,-1,2*(nely+1)*(nelx+1),1);
20 U = zeros(2*(nely+1)*(nelx+1),1);
21 fixeddofs = union([1:2*(nely+1)],[2*(nelx+1)*(nely+1)]);
22 alldofs = [1:2*(nely+1)*(nelx+1)];
23 freeddofs = setdiff(alldofs,fixeddofs);
24 %% PREPARE FILTER
-
83 %% PRINT RESULTS
84 fprintf(' It.:%5i Obj.:%11.4f Vol.:%7.3f ch.:%7.3f\n',loop,c, ...
85 mean(xPhys(:)),change);
-
-
-
-
-
-
-
-
86 %% PLOT DENSITIES
-
87 colormap(gray); imagesc(1-xPhys); caxis([0 1]); axis equal; axis off; drawnow;
-
88 end
89 %
90 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

```

```

. %%% AN 88 LINE TOPOLOGY OPTIMIZATION CODE Nov, 2010 %%%
x %example top88_joseph(140, 50, 0.3, 3, 2, 2)
> function x=top88_ptBENDING(nelx,nely,volfrac,penal,rmin,ft)
> close all
. %% MATERIAL PROPERTIES
x E0 = 210e3;
. Emin = 1e-9;
. nu = 0.3;
> thickness=10;
> Force_amplitude=54e3;
. %% PREPARE FINITE ELEMENT ANALYSIS
. A11 = [12 3 -6 -3; 3 12 3 0; -6 3 12 -3; -3 0 -3 12];
. A12 = [-6 -3 0 3; -3 -6 -3 -6; 0 -3 -6 3; 3 -6 3 -6];
. B11 = [-4 3 -2 9; 3 -4 -9 4; -2 -9 -4 -3; 9 4 -3 -4];
. B12 = [ 2 -3 4 -9; -3 2 9 -2; 4 9 2 3; -9 -2 3 2];
x KE = thickness*1/(1-nu^2)/24*([A11 A12;A12' A11]+nu*[B11 B12;B12' B11]);
. nodenrs = reshape(1:(1+nelx)*(1+nely),1+nely,1+nelx);
. edofVec = reshape(2*nodenrs(1:end-1,1:end-1)+1,nelx*nely,1);
. edofMat = repmat(edofVec,1,8)+repmat([0 1 2*nely+[2 3 0 1] -2 -1],nelx*nely,1);
. iK = reshape(kron(edofMat,ones(8,1))',64*nelx*nely,1);
. jK = reshape(kron(edofMat,ones(1,8))',64*nelx*nely,1);
. % DEFINE LOADS AND SUPPORTS (HALF MBB-BEAM)
x F = Force_amplitude*sparse(2,1,-1,2*(nely+1)*(nelx+1),1);
. U = zeros(2*(nely+1)*(nelx+1),1);
x fixeddofs = union([1:2*(nely+1)],[(nelx+2)*(nely+1)]);
. alldofs = [1:2*(nely+1)*(nelx+1)];
. freeddofs = setdiff(alldofs,fixeddofs);
. %% PREPARE FILTER
-
. %% PRINT RESULTS
. fprintf(' It.:%5i Obj.:%11.4f Vol.:%7.3f ch.:%7.3f\n',loop,c, ...
. mean(xPhys(:)),change);
> figure(2)
> hold on
> plot(loop,c,'bo','MarkerFaceColor','b')
> plot(loop,mean(xPhys(:))*100,'ro','MarkerFaceColor','r')
> % plot(outeriter,(1+GKSL)*VM1,'ko','MarkerFaceColor','k')
> title(['Convergence volfrac = ',num2str(mean(xPhys(:))*100),' , Compliance = ',num
> grid on
> legend('Compliance','Volume Fraction %')
> xlabel('iter')
. %% PLOT DENSITIES
> figure(1)
. colormap(gray); imagesc(1-xPhys); caxis([0 1]); axis equal; axis off; drawnow;
> print(['DZ_it',num2str(loop,'%3d')],'-dpng')
. end
. %
. %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

```

[58 unmodified lines hidden]

[26 unmodified lines hidden]

OPTIMALITY CRITERIA

```
function [xnew]=OC(nelx,nely,x,volfrac,dc)
l1 = 0; l2 = 100000; move = 0.2;
while (l2-l1 > 1e-4)
    lmid = 0.5*(l2+l1);
    xnew = max(0.001,max(x-move,min(1.,min(x+move,x.*sqrt(-dc./lmid)))));
    if sum(sum(xnew)) - volfrac*nelx*nely > 0;
        l1 = lmid;
    else
        l2 = lmid;
    end
end
```

lmid: 50.0000	l1: 0.0000	l2: 50.0000
lmid: 25.0000	l1: 0.0000	l2: 25.0000
lmid: 12.5000	l1: 0.0000	l2: 12.5000
lmid: 6.2500	l1: 6.2500	l2: 12.5000
lmid: 9.3750	l1: 6.2500	l2: 9.3750
lmid: 7.8125	l1: 6.2500	l2: 7.8125
lmid: 7.0313	l1: 7.0313	l2: 7.8125
lmid: 7.4219	l1: 7.0313	l2: 7.4219
lmid: 7.2266	l1: 7.2266	l2: 7.4219
lmid: 7.3242	l1: 7.3242	l2: 7.4219
lmid: 7.3730	l1: 7.3242	l2: 7.3730
lmid: 7.3486	l1: 7.3242	l2: 7.3486
lmid: 7.3364	l1: 7.3364	l2: 7.3486
lmid: 7.3425	l1: 7.3425	l2: 7.3486
lmid: 7.3456	l1: 7.3425	l2: 7.3456
lmid: 7.3441	l1: 7.3441	l2: 7.3456
lmid: 7.3448	l1: 7.3448	l2: 7.3456
lmid: 7.3452	l1: 7.3448	l2: 7.3452
lmid: 7.3450	l1: 7.3450	l2: 7.3452
lmid: 7.3451	l1: 7.3450	l2: 7.3451

Can also use:

- fmincon
- MMA...or any other optimizer

The MMA approach, which was initially proposed by Svanberg (see Mini Project) is based on the first-order Taylor series expansion of the objective and constraint functions.

With this method, an explicit convex subproblem is generated to approximate the implicit nonlinear problem.

Matlab code command

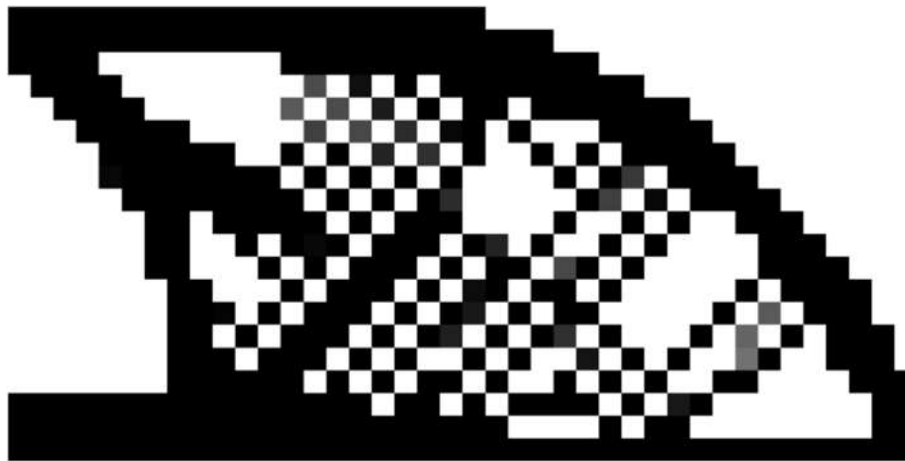
`top(nelx, nely, volfrac, penal, rmin)`

- nelx and nely: number of elements in the horizontal and vertical directions,
- volfrac: volume fraction,
- penal: penalization power,
- **rmin: filter size (divided by element size).**

Example 1

Numerical instability

- `top(40, 20, 0.5, 3, 1.0)`



effect → Checkerboard Pattern

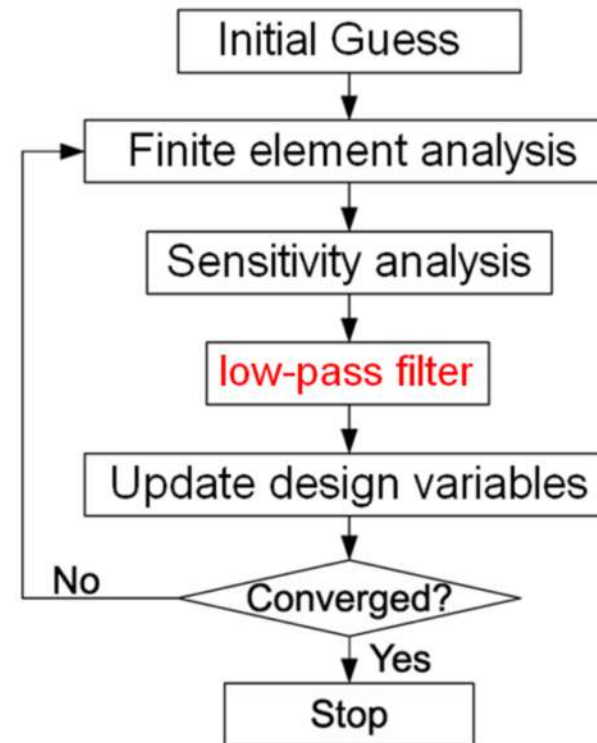
Example 1 -- Checkerboard Pattern Problem

→ Solution: LOW PASS Filter

$$\widehat{\frac{\partial c}{\partial x_e}} = \frac{1}{x_e \sum_{f=1}^N \hat{H}_f} \sum_{f=1}^N \hat{H}_f x_f \frac{\partial c}{\partial x_f}.$$

$$\hat{H}_f = r_{\min} - \text{dist}(e, f),$$

$$\{f \in N \mid \text{dist}(e, f) \leq r_{\min}\}, \quad e = 1, \dots, N$$



```
[dc] = check(nelx, nely, rmin, x, dc);
```

```
% FILTERING OF SENSITIVITIES
```

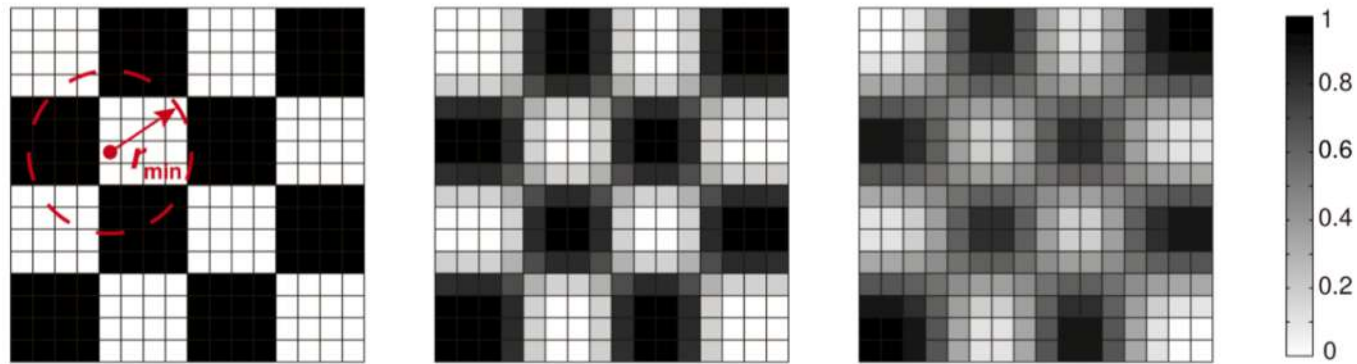
Example 2

- `top(40, 20, 0.5, 3, 1.5)`



- `top(40, 20, 0.5, 3, 3)` effect?

Filter

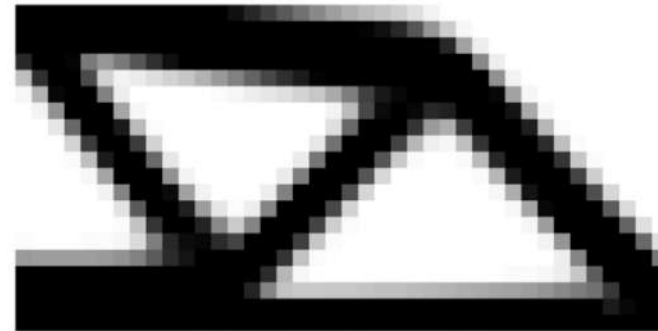


A checkerboard field and filtered fields ($r_{\min} = 1.5l_e$ and $3l_e$)

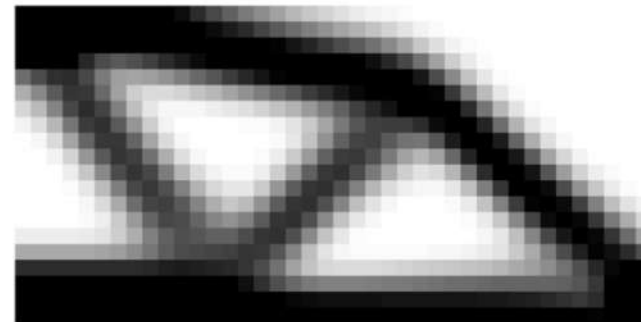
Example 3

→ Size of the filter makes it possible to obtain a more physical representation (Unblur?)

- $r_{\min}=1.5$
Obj=82.7562;



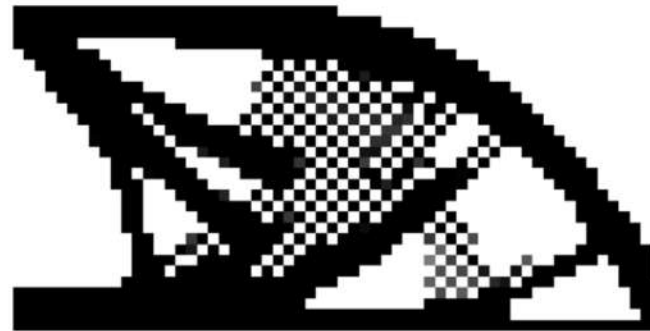
- $r_{\min}=3$
Obj=99.1929;



Example 3: Change mesh !

- `top(60, 30, 0.5, 3, 1.0)`

Obj: 83.0834



- `top(40, 20, 0.5, 3, 1.0)`

Obj: 80.4086;

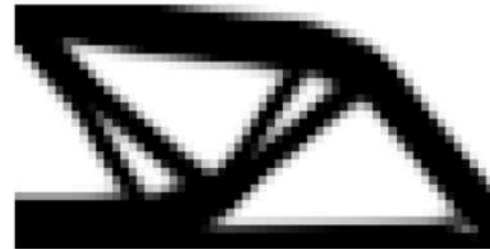


Example 4: mixing



- `top(60, 30, 0.5, 3, 1.5)`

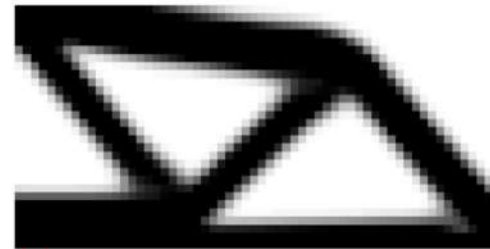
Obj: 81.3491



- `top(60, 30, 0.5, 3, 2.25)`

Obj: 83.5963

→ Filter size



→ +mesh dependency, combined eddect It's complicated

- `top(40, 20, 0.5, 3, 1.5)`

Obj=82.7562;



Common questions ?

- Grayness level ?

%% Greyness Level

```
gl = 4/nele*sum(xPhys(:).*(1-xPhys(:)));
```

- How to obtain B&W for the final design ?

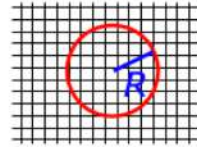
```
function xPhys=automatic_threshold(xPhys)
% %code to input xPhys and obtain the B&W design at respected volfrac
```

```
xPhys01=xPhys;
volfrac=mean(mean(xPhys))
lowlim=0;
highlim=1;
difVF=1;
while difVF>0.0001 && highlim-lowlim > 0.000001
midlim=(lowlim+highlim)/2;
xPhys01(xPhys>=midlim)=1;
xPhys01(xPhys<midlim)=0;
volfrac01=mean(mean(xPhys01));
difVF=abs(volfrac01-volfrac);
if volfrac01<volfrac
highlim=midlim;
else
lowlim=midlim;
end
end
xPhys=xPhys01;
```

Regularization by low-pass filtering

Neighborhood:

$$N_e = \{i \mid \|\mathbf{x}_i - \mathbf{x}_e\| \leq R\}$$



Checkerboards

Sensitivity filtering (Sigmund 1997, Sigmund&Maute 2012)

$$\frac{\widetilde{\partial \Phi}}{\partial \rho_e} = \frac{\sum_{i \in N_e} H(\mathbf{x}_i) \rho_i \frac{\partial \Phi}{\partial \rho_i}}{\rho_e \sum_{i \in N_e} H(\mathbf{x}_i)}$$



Density filtering (Bruns&Tortorelli/Bourdin 2001)

$$E_e(\rho) = \tilde{\rho}_e^p E_0, \quad \tilde{\rho}_e = \frac{\sum_{i \in N_e} H(\mathbf{x}_i) \rho_i}{\sum_{i \in N_e} H(\mathbf{x}_i)}$$

PDE-based filtering (Lazarov&Sigmund 2011)

$$-r^2 \Delta \tilde{\rho} + \tilde{\rho} = \rho$$

Mesh refinement



Public Codes



99 Line basic Matlab (Including FE, grad's, OC)

OS, A 99 line topology optimization code written in MATLAB, *SMO*, **2001**, 21, 120-127

88 line advanced Matlab (+advanced filters)

Andreassen, E.; Clausen, A.; Schevenels, M.; Lazarov, B. & OS, Efficient topology optimization in MATLAB using 88 lines of code, *SMO*, **2011**, 43, 1-16

On multigrid-CG for efficient topology optimization

Amir, O.; Aage, N. & Lazarov, B.S., *SMO*, **2014**, 49, 815-829

Topology optimization using PETSc:

An easy-to-use, fully parallel, open-source topology optimization framework

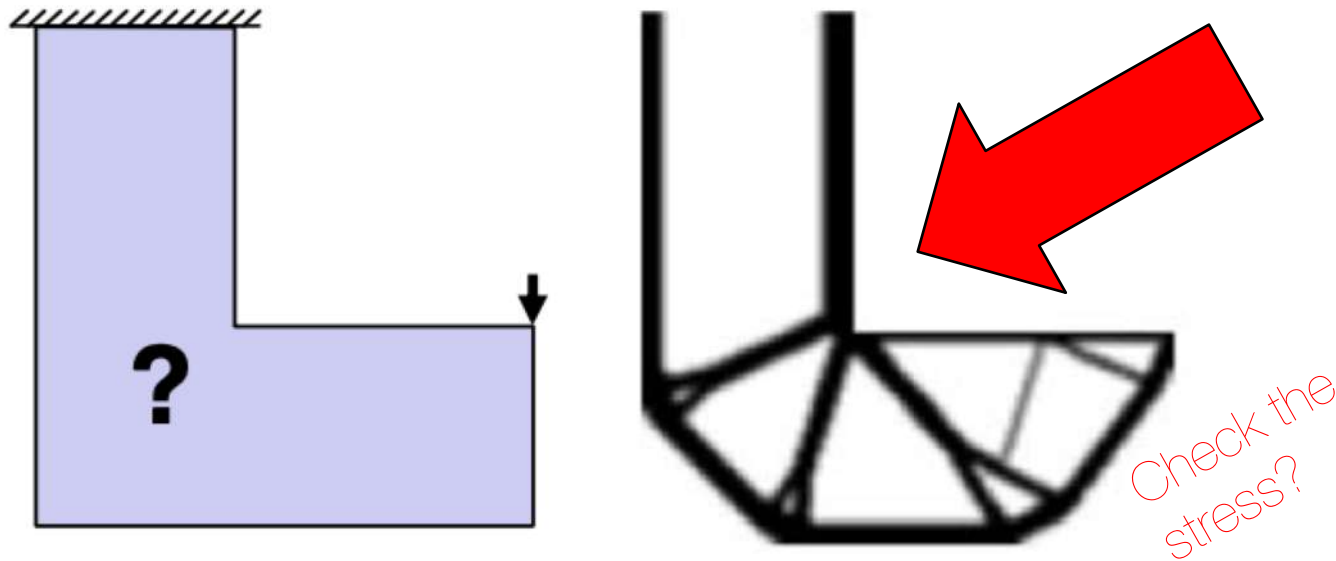
Aage, N; Andreassen, E. & Lazarov, B.S., **2015**, *SMO*, 51, 565-572



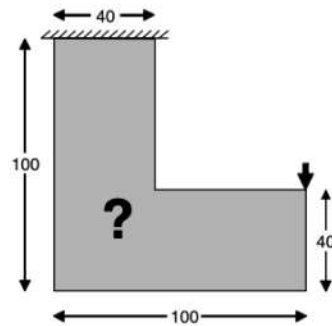
Freely downloadable from www.topopt.dtu.dk

At this time the structure is rigid... but feasible?

Le, C., Norato, J., Bruns, T., Ha, C., & Tortorelli, D. (2010). Stress-based topology optimization for continua. *Structural and Multidisciplinary Optimization*, 41(4), 605-620.



Results



Find $\mathbf{X} = [x_1, x_2, \dots, x_N]^T$

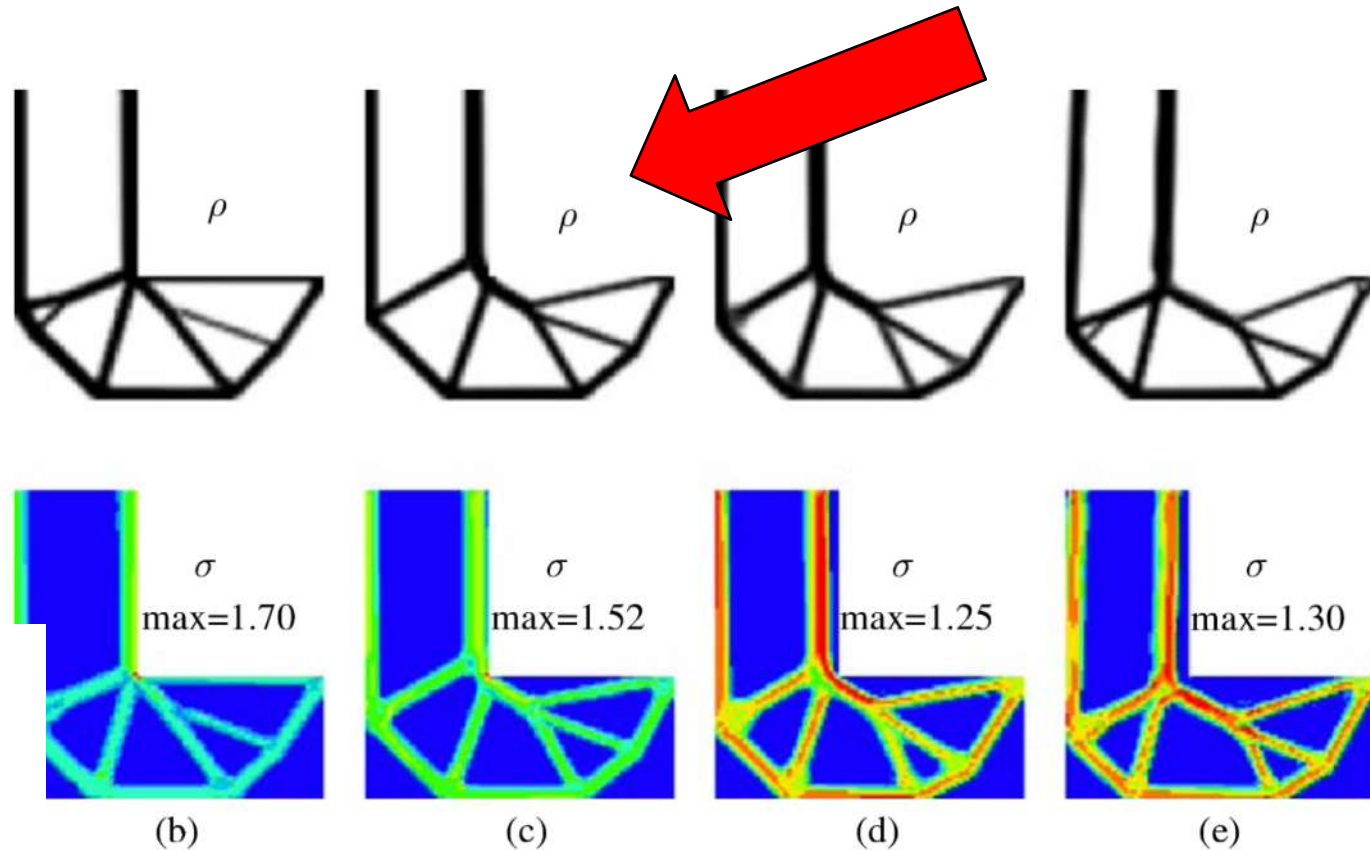
minimize $f(\mathbf{X}) = W(\mathbf{X})$

subject $\mathbf{K}(\mathbf{X})\mathbf{U}(\mathbf{X}) = \mathbf{F}$

$$\boldsymbol{\sigma}(\mathbf{X}) - [\boldsymbol{\sigma}] \leq 0$$

$$\frac{V(\mathbf{X})}{V_0} \leq f$$

$$0 < x_{\min} \leq x_e \leq x_{\max} \leq 1$$



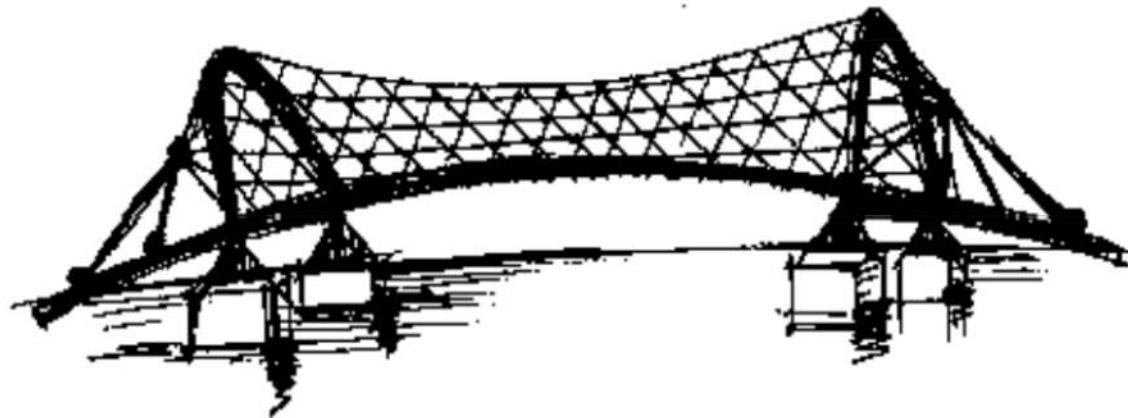
r P on the L-bracket. **a** Problem definition, **b** $P = 4$ (63 iter.), **c** $P = 6$ (71 iter.), **d** $P = 8$ (78 iter.), **e** $P =$

Is it better?

**"The art of structure is
where to put the holes"**

Robert Le Ricolais

French-American engineer and philosopher
(1894-1977)



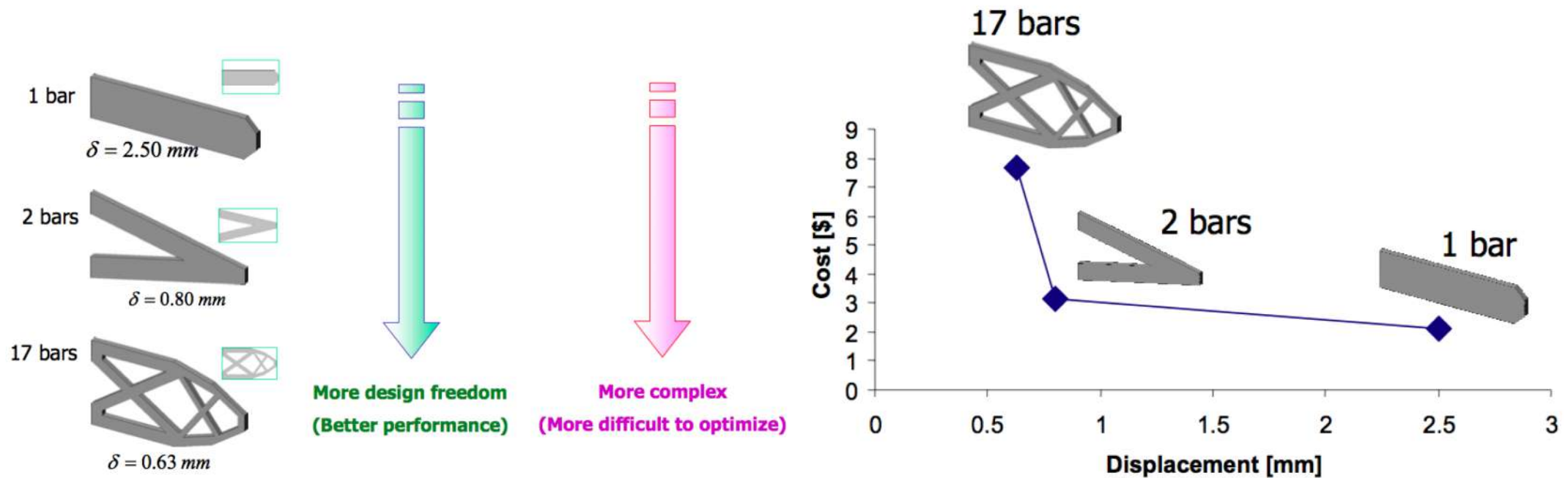
MDO_ML_21

TOPOPT vs Generative Design

- <https://www.autodesk.com/products/fusion-360/blog/topology-optimization-is-not-generative-design/>
- **Is the outcome returned CAD ready, or does it have to be rebuilt?**
 - *A mesh model that must be rebuilt as valid geometry in a CAD system. (This is topology optimization)*
OR
 - *CAD ready for any CAD System Geometry (This is generative design in Fusion 360)*
- ➔ **just wait for PART 4 and GGP framework**



WAIT A MINUTE..., this is an engineer's conclusion



The engineer or the art of compromise?

Multidisciplinary Optimization and Machine Learning for Engineering Design

19 July 2021 – 5 August 2021

<https://mdoml2021.ftmd.itb.ac.id/>

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Break before Part 3

Prof. J. Morlier