

Introduction to topology optimization

Applied and computational mathematics seminar

Mathematics Department
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Louis Komzsik
Lecturer

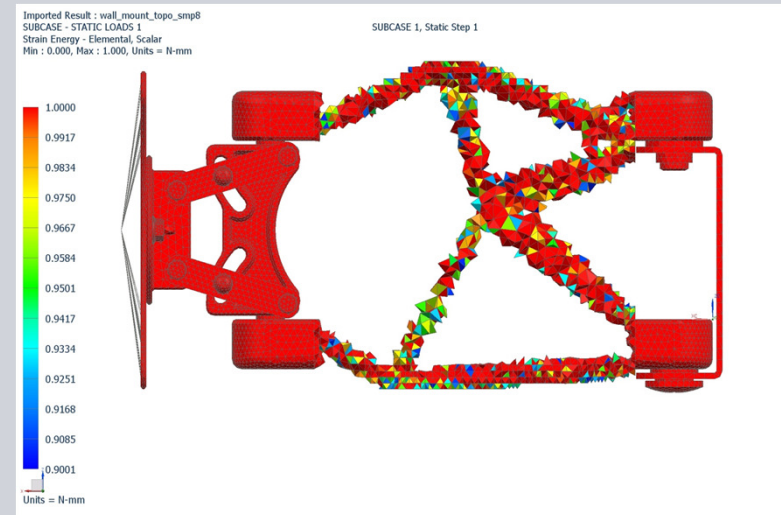
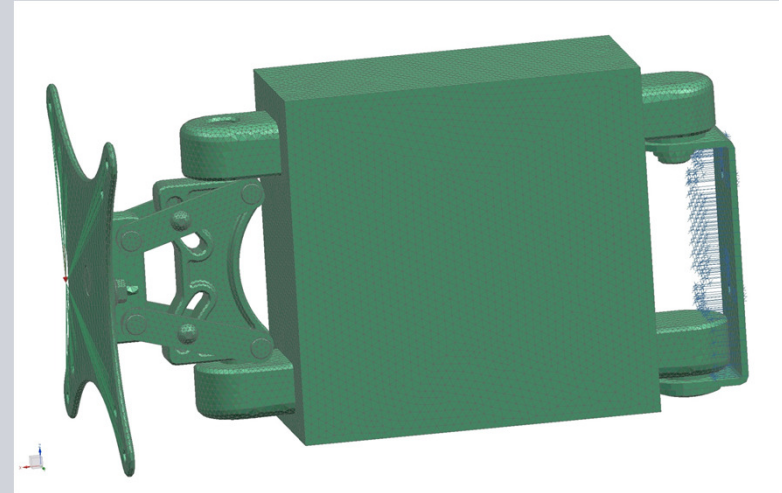
Topics of presentation

Mathematical foundation

- Concept of topology optimization
- Finite element foundation
- Objectives and constraints
- Matrix sensitivities
- Solution gradients
- Workflow of optimization

Application examples

- Demonstration example
- Cantilever structure
- 4 legged stool
- Drilling platform
- Building bridges
- Additive manufacturing



Topoplogy optimization concept

Global kinetic energy

$$E_k = \frac{1}{2} \iiint_V \rho \dot{q}^2 dV = \frac{1}{2} \iiint_V \dot{q}^T N^T \rho N \dot{q} dV$$

$$E_k = \frac{1}{2} \sum_{i=1}^N \dot{q}_i^T \iiint_{V_e} N^T \rho N dV_e \dot{q}_i = \frac{1}{2} \sum_{i=1}^N \dot{q}_i^T M_i \dot{q}_i$$

Global potential energy

$$E_s = \frac{1}{2} \iiint_V E \varepsilon^2 dV = \frac{1}{2} \iiint_V q^T B^T E B q dV$$

$$E_s = \frac{1}{2} \sum_{i=1}^N q_i^T \iiint_{V_e} B^T E B dV_e q_i = \frac{1}{2} \sum_{i=1}^N q_i^T K_i q_i$$

Element energies

$$\varepsilon_i = q_i^T K_i q_i; \kappa_i = \dot{q}_i^T M_i \dot{q}_i$$

Eliminate or decrease
density of elements when

$$\varepsilon_i \leq \varepsilon_{\min}; \kappa_i \geq \kappa_{\min}$$

Topology optimization foundation

Finite element model of structure

$$K = \sum_{i=1}^N K_i ; M = \sum_{i=1}^N M_i$$

Element density as design variables

$$d_{\min} \leq d_i \leq d_{\max} ; i = 1, \dots, N$$

Updated finite element matrices

$$K = \sum_{i=1}^N K_i^0 d_i ; K = K_i^0 d_i + \sum_{k=1, k \neq i}^N K_k^0 d_k$$

$$M = \sum_{i=1}^N M_i^0 d_i ; M = M_i^0 d_i + \sum_{k=1, k \neq i}^N M_k^0 d_k$$

Sensitivity of matrices

$$\frac{\partial K}{\partial d_i} = \frac{\partial K_i}{\partial d_i} = K_i^0 = \frac{K_i}{d_i}$$

$$\frac{\partial M}{\partial d_i} = \frac{\partial M_i}{\partial d_i} = M_i^0 = \frac{M_i}{d_i}$$

Topology optimization problem

Minimize an objective function

$$f = f(K, M)$$

subject to one or more constraints

$$g_j; j = 1, 2, \dots, m$$

Main objective = minimize weight of structure

$$W < W_0$$

Limit static deformation at a given location

$$u_j \leq u_{\max}$$

Increase structural compliance in static solution

$$C \geq C_{\min}$$

Natural frequency magnitude above excitations

$$\lambda_i \geq \lambda_{\min}$$

Weight objective sensitivity

Global gravity vector $a_G(i) = [\cdot a_G(i_1) \cdot a_G(i_2) \cdot a_G(i_3) \cdot a_G(i_4) \cdot]^T$

Weight of the structure $W = \sum_{i=1}^N W_i$

Updated weight $W(d_i) = d_i W_i^0 + \sum_{k=1, k \neq i}^N W_k d_k$

Mass basis $W(d_i, M) = (M_i^0 d_i + \sum_{k=1, k \neq i}^N M_k^0 d_k) a_G$

Sensitivity of weight $\frac{\partial W}{\partial d_i} = M_i^0 a_G(i) = \frac{M_i}{d_i} a_G(i)$

Gravity load $F_g = W; \frac{\partial F_g}{\partial d_i} \neq 0$

Static deformation sensitivity

Linear static solution

$$Ku = F \quad u^T K = (K^T u)^T = (Ku)^T = F^T$$

Solution sensitivity

$$\frac{\partial K}{\partial d_i} u + K \frac{\partial u}{\partial d_i} = \frac{\partial F}{\partial d_i}$$

Constant loading force

$$\frac{\partial F}{\partial d_i} = 0 \quad u^T K \frac{\partial u}{\partial d_i} = -u^T \frac{\partial K}{\partial d_i} u$$

Substitute and premultiply

$$F^T \frac{\partial u}{\partial d_i} = -u^T \frac{\partial K}{\partial d_i} u = -u^T \frac{K_i}{d_i} u$$

Element strain energy

$$\varepsilon_i = u^T K_i u$$

Deformation sensitivity at node j

$$\frac{\partial u(j)}{\partial d_i} = -\frac{1}{F(j)d_i} \varepsilon_i$$

Structural compliance sensitivity

Structural compliance definition

$$C = F^T u$$

Sensitivity of compliance

$$\frac{\partial C}{\partial d_i} = \frac{\partial F^T}{\partial d_i} u + F^T \frac{\partial u}{\partial d_i}$$

Static solution sensitivity

$$\frac{\partial K}{\partial d_i} u + K \frac{\partial u}{\partial d_i} = \frac{\partial F}{\partial d_i} \quad \frac{\partial u}{\partial d_i} = K^{-1} \left(\frac{\partial F}{\partial d_i} - \frac{\partial K}{\partial d_i} u \right)$$

Compliance in general

$$\frac{\partial C}{\partial d_i} = \frac{\partial F^T}{\partial d_i} u + F^T K^{-1} \left(\frac{\partial F}{\partial d_i} - \frac{\partial K}{\partial d_i} u \right)$$

$$\frac{\partial C}{\partial d_i} = \frac{\partial F^T}{\partial d_i} u + u^T \frac{\partial F}{\partial d_i} - u^T \frac{\partial K}{\partial d_i} u$$

Constant force specific

$$\frac{\partial C}{\partial d_i} = -u^T \frac{\partial K}{\partial d_i} u = -\frac{1}{d_i} \varepsilon_i$$

Natural frequency sensitivity

Eigenvalue analysis solution

$$(K - \lambda M) \varphi = 0$$

Eigenvalue solution sensitivity

$$\frac{\partial K}{\partial d_i} \varphi + K \frac{\partial \varphi}{\partial d_i} = \lambda \frac{\partial M}{\partial d_i} \varphi + \frac{\partial \lambda}{\partial d_i} M \varphi + \lambda M \frac{\partial \varphi}{\partial d_i}$$

Assumptions

$$\varphi^T M \varphi = I \quad \frac{\partial \varphi}{\partial d_i} = 0$$

Eigenvalue sensitivity

$$\frac{\partial \lambda}{\partial d_i} = \varphi^T \left(\frac{\partial K}{\partial d_i} - \lambda \frac{\partial M}{\partial d_i} \right) \varphi$$

With matrix sensitivities

$$\frac{\partial \lambda}{\partial d_i} = \varphi^T \frac{K_i}{d_i} \varphi - \lambda \varphi^T \frac{M_i}{d_i} \varphi = \frac{1}{d_i} (\varepsilon_i - \kappa_i)$$

Element kinetic energy

$$\kappa_i = \dot{\varphi}^T M_i \dot{\varphi}; \dot{\varphi} = \lambda \varphi$$

Optimization problem definition

Seeking for the objective

$$f(\bar{d}^*) = \text{extremum}$$

Subject to constraints

$$g_j(\bar{d}^*) \leq g_j^{\text{max}}; j = 1, \dots, m$$

Design variables

$$\bar{d}^T = [d_1 \quad \dots \quad d_N]$$

Objective sensitivities

$$\bar{f}^T = \left[\begin{array}{ccc} \frac{\partial f}{\partial d_1} & \dots & \frac{\partial f}{\partial d_N} \end{array} \right]$$

Constraints sensitivities

$$G^T = \left[\begin{array}{ccc} \dots & \dots & \dots \\ \frac{\partial g_j}{\partial d_1} & \dots & \frac{\partial g_j}{\partial d_N} \\ \dots & \dots & \dots \end{array} \right]$$

Topology optimization flow

Start analysis job

- Set current design variables
- Modify element matrices
- Update assembled model
- Solve analysis problem
- Collect and compute
- Call optimizer to receive
- Continue if warranted

$$\bar{d}^k = \left[d_1^k \quad \dots \quad d_N^k \right]^T ; k = 1, 2, \dots$$

$$K_i^k = K_i^0 d_i^k ; M_i^k = M_i^0 d_i^k$$

$$K^k = \sum_{i=1}^N K_i^k ; M^k = \sum_{i=1}^N M_i^k$$

$$K^k u = F \quad (K^k + \lambda M^k) \phi = 0$$

$$\mathcal{E}_i^k, \kappa_i^k ; f(\bar{d}^k), g_j(\bar{d}^k) ; \bar{f}(\bar{d}^k), G(\bar{d}^k)$$

$$\bar{d}^{k+1}, \bar{d}^*$$

$$k = k + 1 ; \text{if } k \leq k_{\max} \rightarrow 2.$$

Complete analysis job

Mathematical optimization process

Moving asymptotes

$$L_i^{k < 3} = d_i^k - (d_{\max} - d_{\min}) / 2 \quad U_i^{k < 3} = d_i^k + (d_{\max} - d_{\min}) / 2$$

$$L_i^{k \geq 3} = d_i^k - c_k (d_i^{k-1} - L_i^{k-1}) \quad U_i^{k \geq 3} = d_i^k + c_k (U_i^{k-1} - d_i^{k-1})$$

Convexity factors
(simplified)

$$p_i^k = (U_i^k - d_i^k)^2 \frac{\partial f}{\partial d_i^k}; \text{ if } \frac{\partial f}{\partial d_i^k} \geq 0;$$

$$q_i^k = (d_i^k - L_i^k)^2 \frac{\partial f}{\partial d_i^k}; \text{ if } \frac{\partial f}{\partial d_i^k} \leq 0;$$

Mathematical problem

$$\hat{f}(\bar{d}^k) = f(\bar{d}^k) + \sum_{i=1}^N \left(\frac{p_i^k}{U_i^k - d_i^0} + \frac{q_i^k}{d_i^0 - L_i^k} \right)$$

$$\hat{g}_j(\bar{d}^k) = g_j(\bar{d}^k) + \sum_{i=1}^N \left(\frac{r_i^k}{U_i^k - d_i^0} + \frac{s_i^k}{d_i^0 - L_i^k} \right)$$

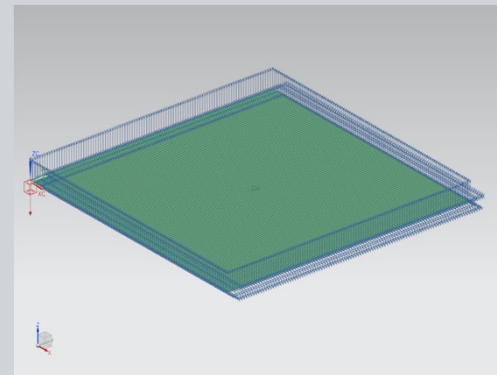
Find and return optimum

$$\hat{f}(\bar{d}^k) = \min; \hat{g}_j \leq g_j^{\max} \rightarrow \bar{d}^{k+1}, \bar{d}^*$$

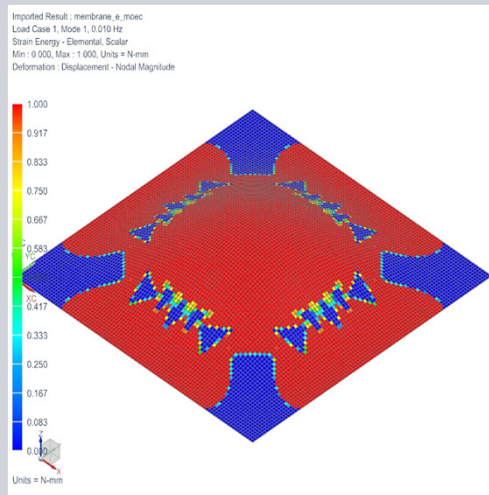
The effect of solution and constraint type

Model statistics

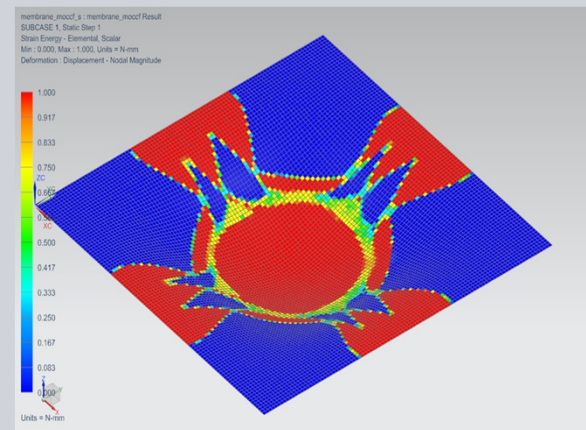
- 10,000 elements (cquad4)
- 10,000 design variables



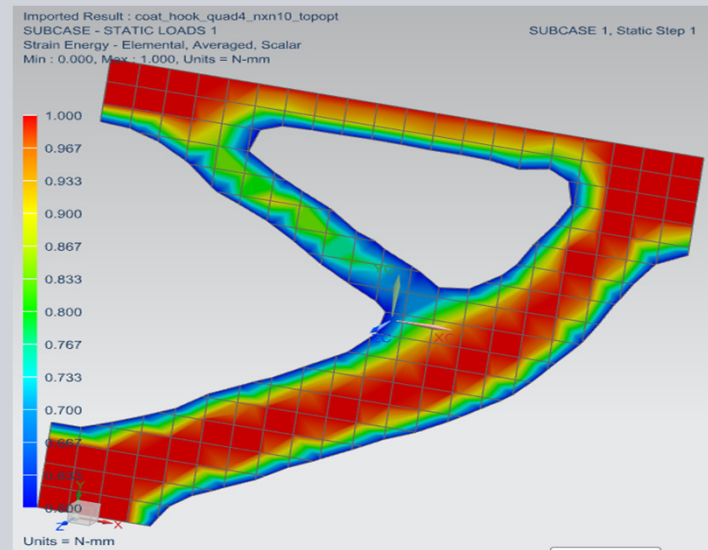
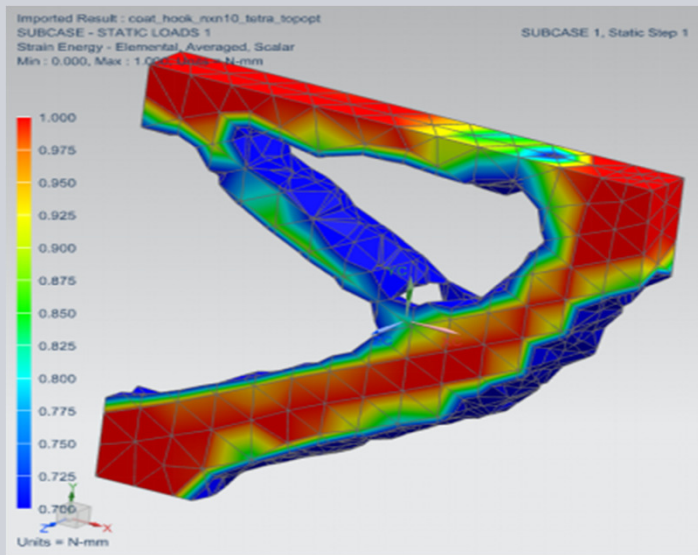
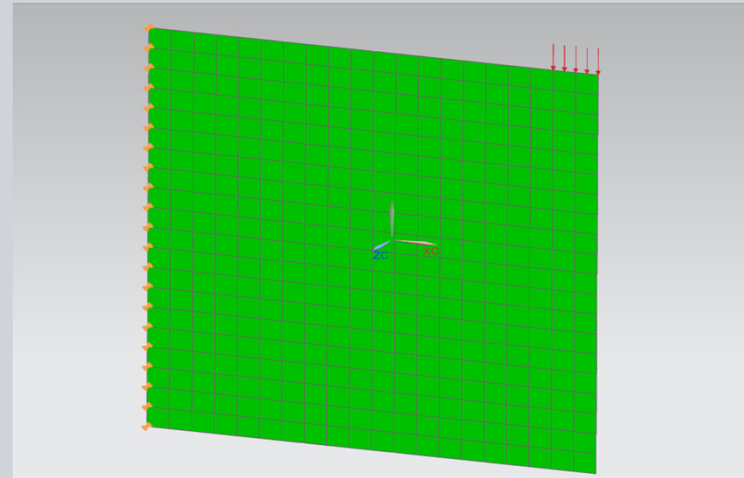
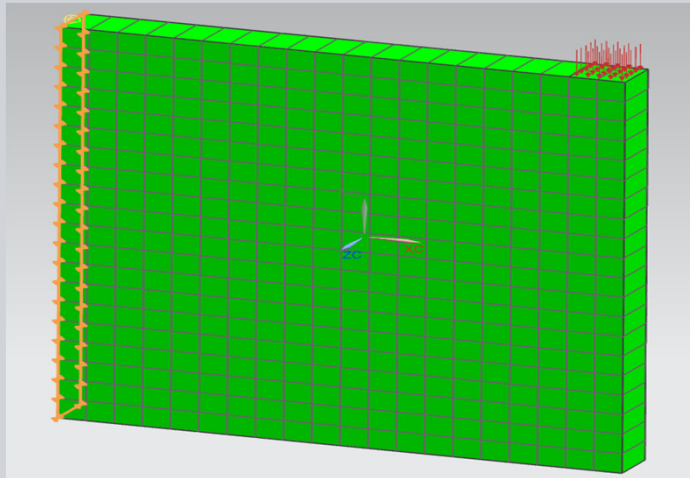
Objective : minimize mass
Constraint : first eigenvalue



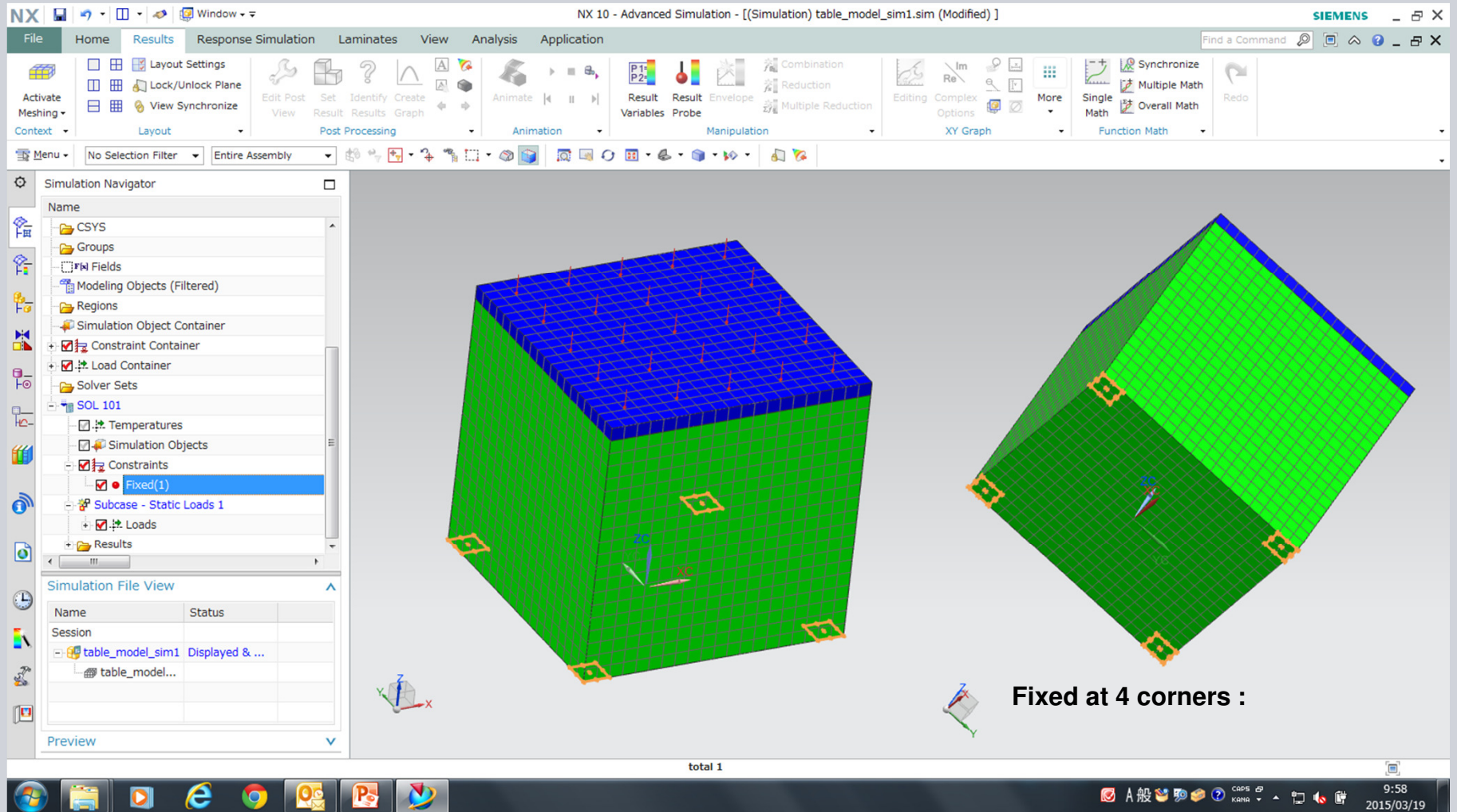
Objective : minimize mass
Constraint : compliance



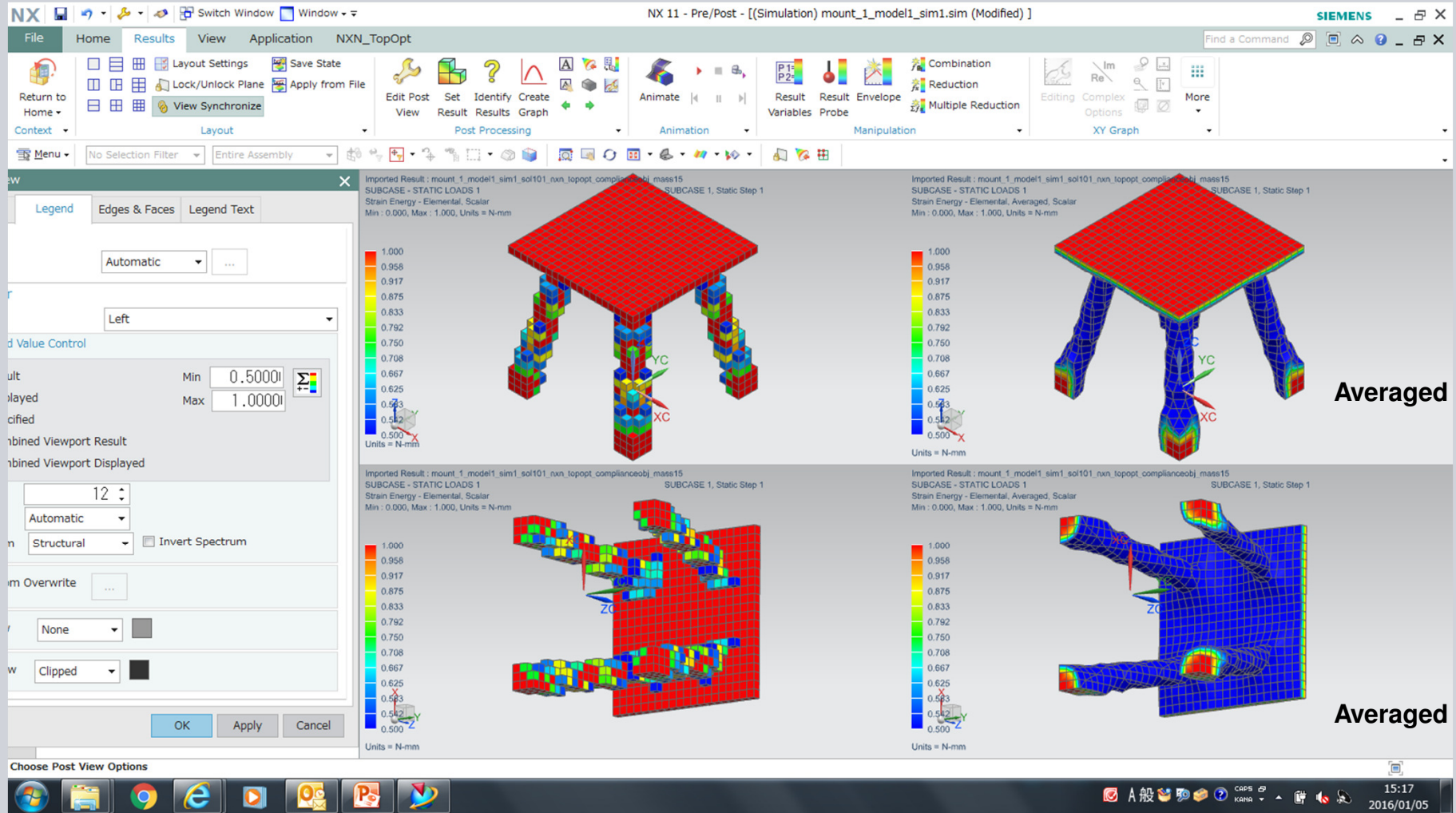
The influence of modeling



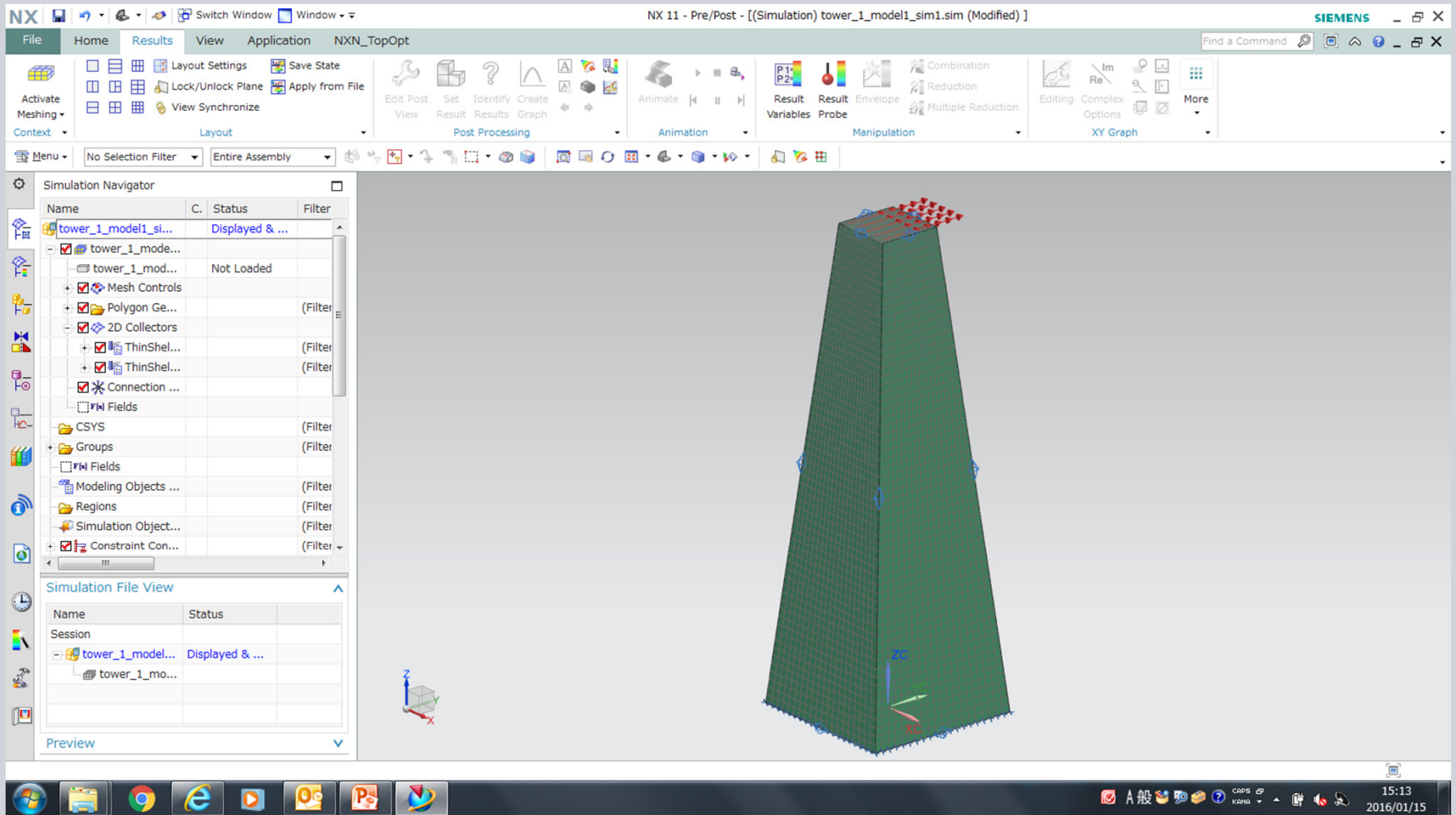
Design a stool



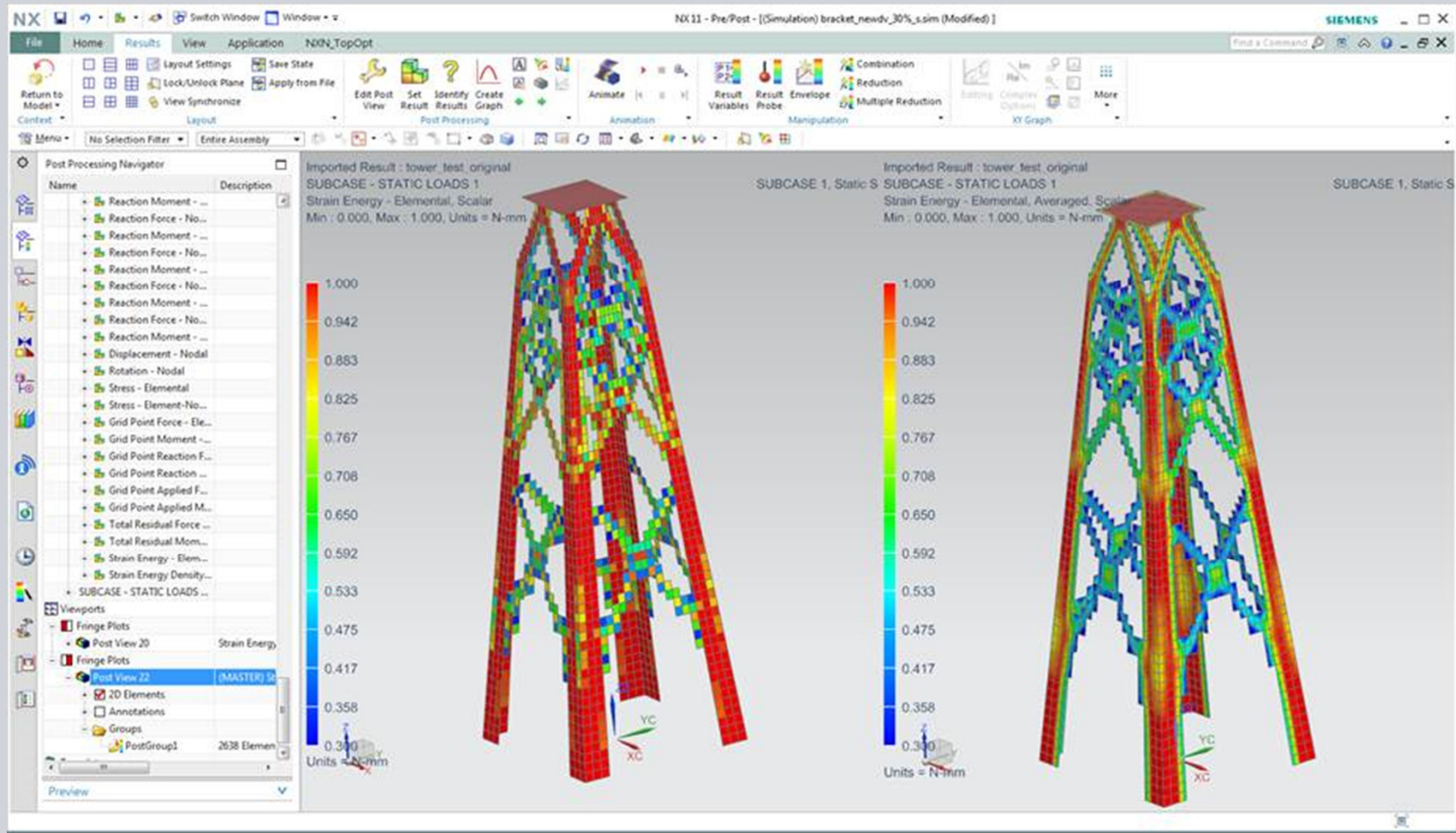
Minimize mass, compliance constraint



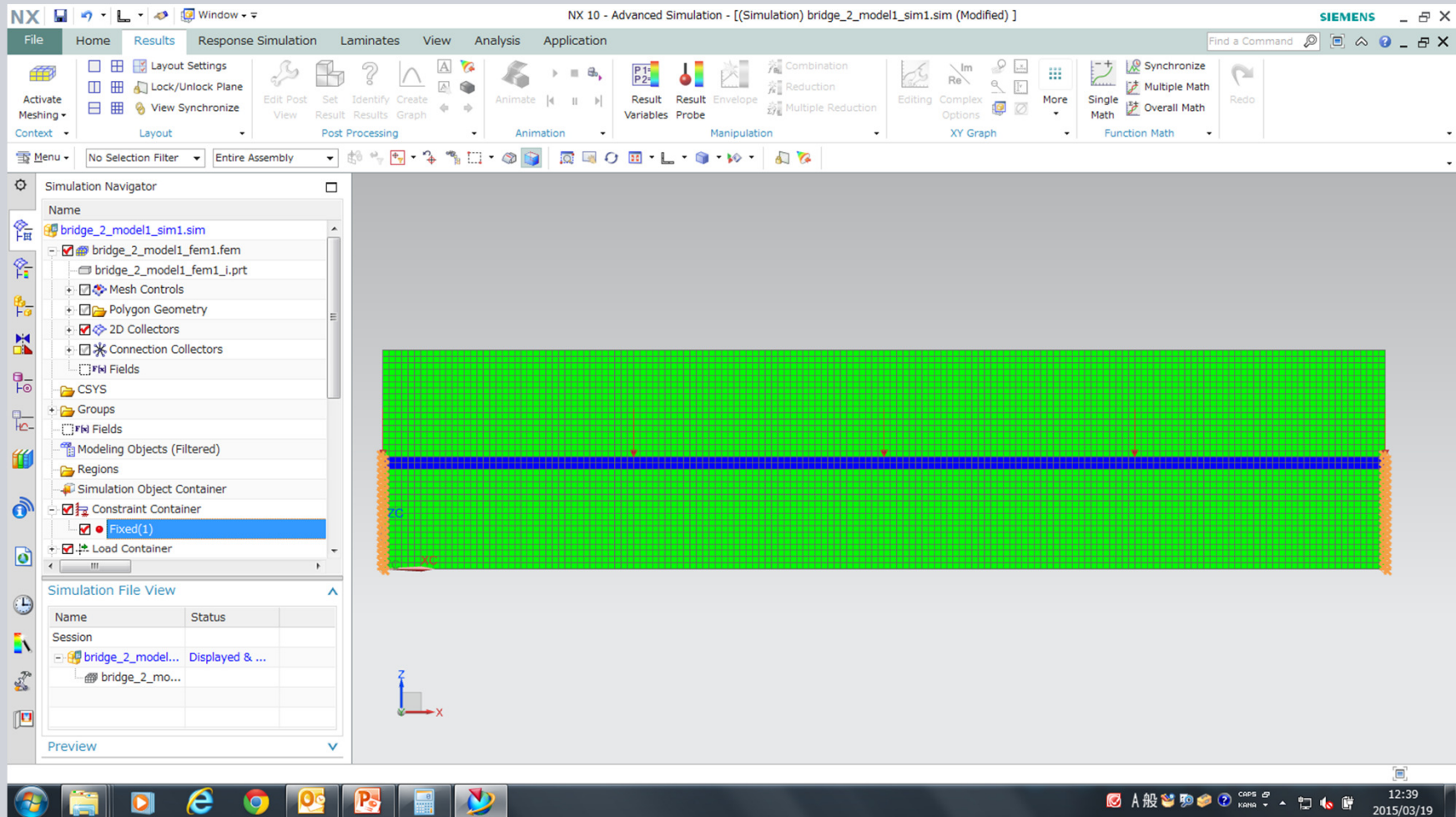
Design a drilling platform



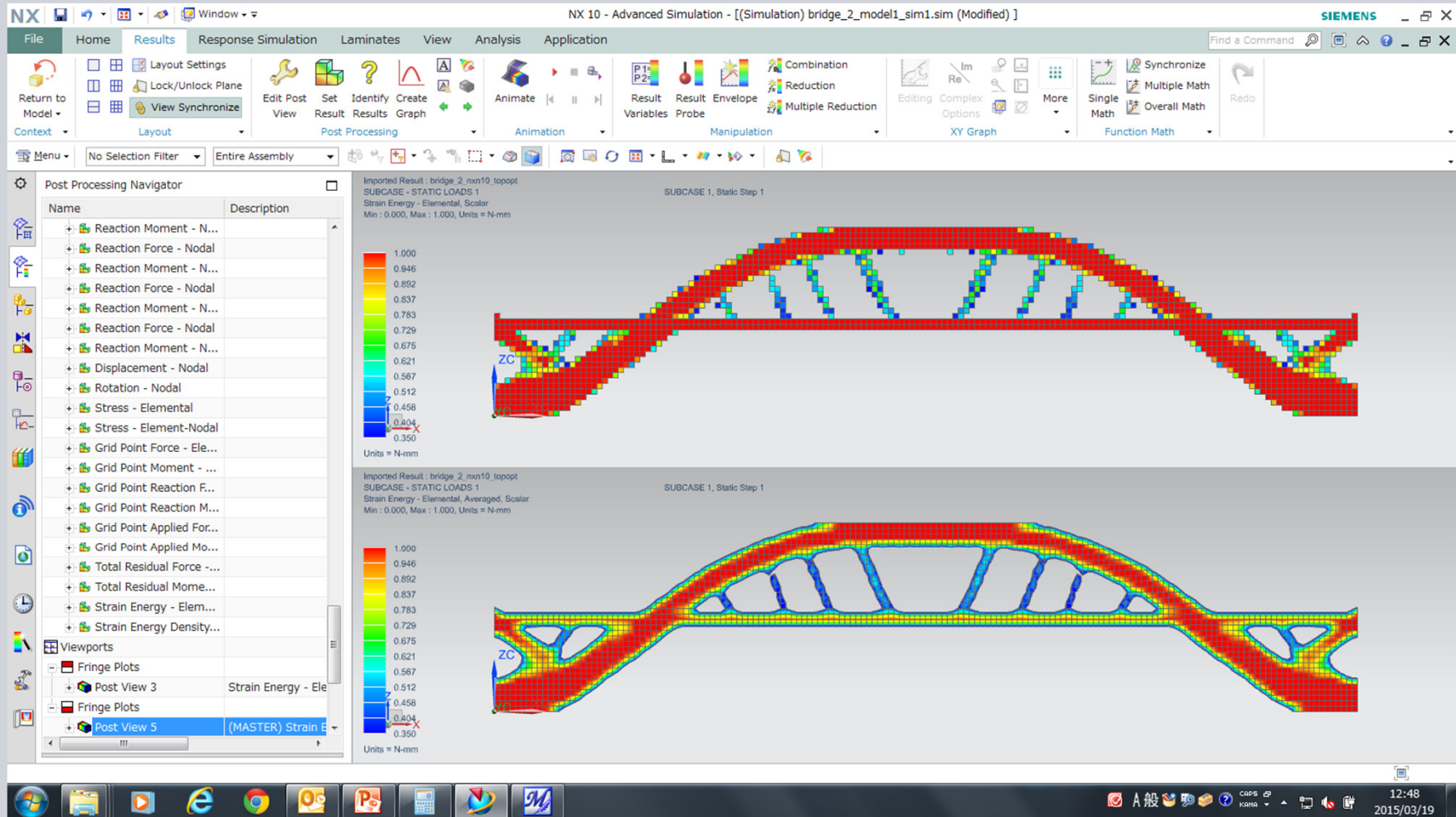
Minimize mass, complaine constraint



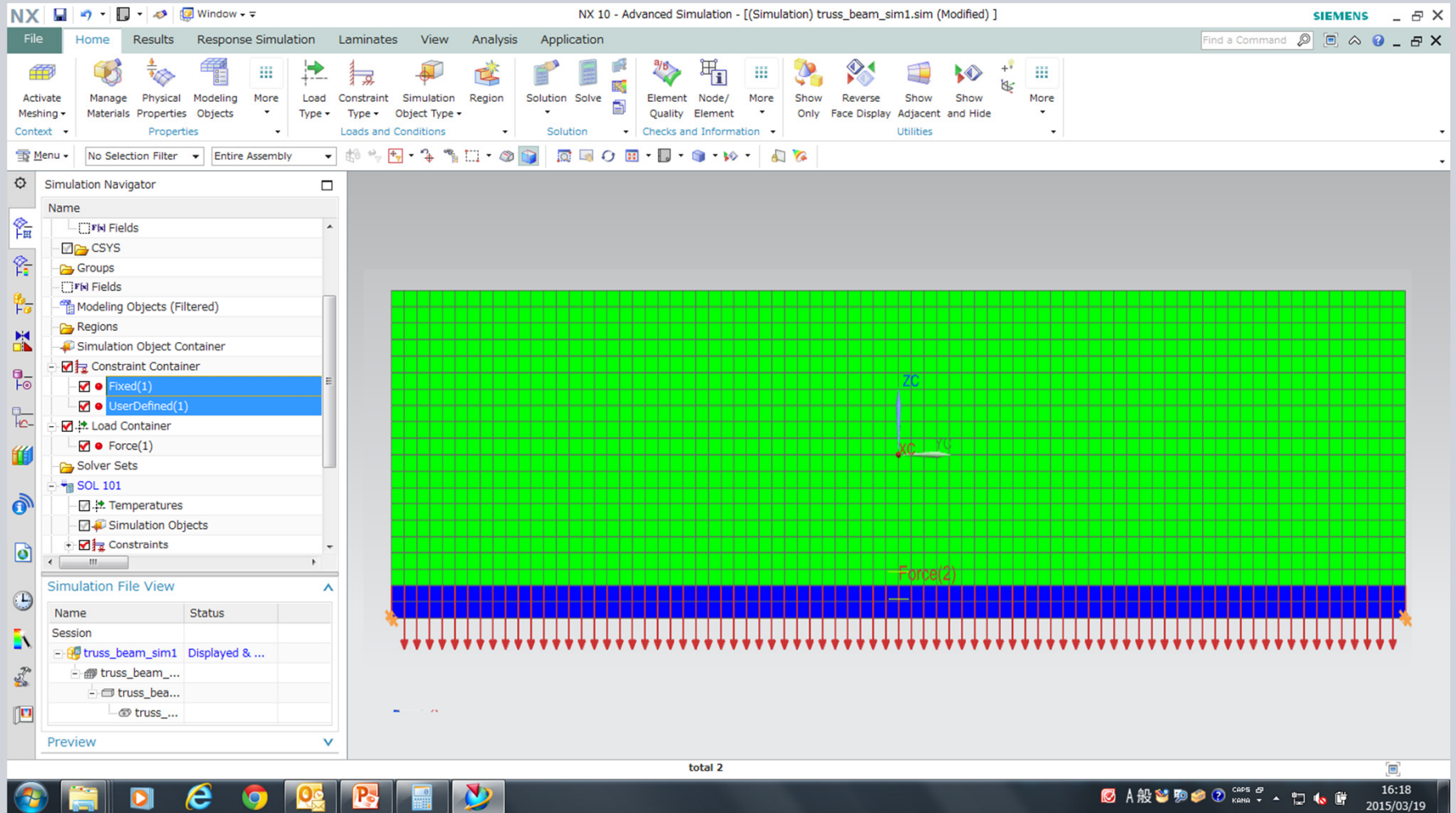
Mid-loaded bridge model design space



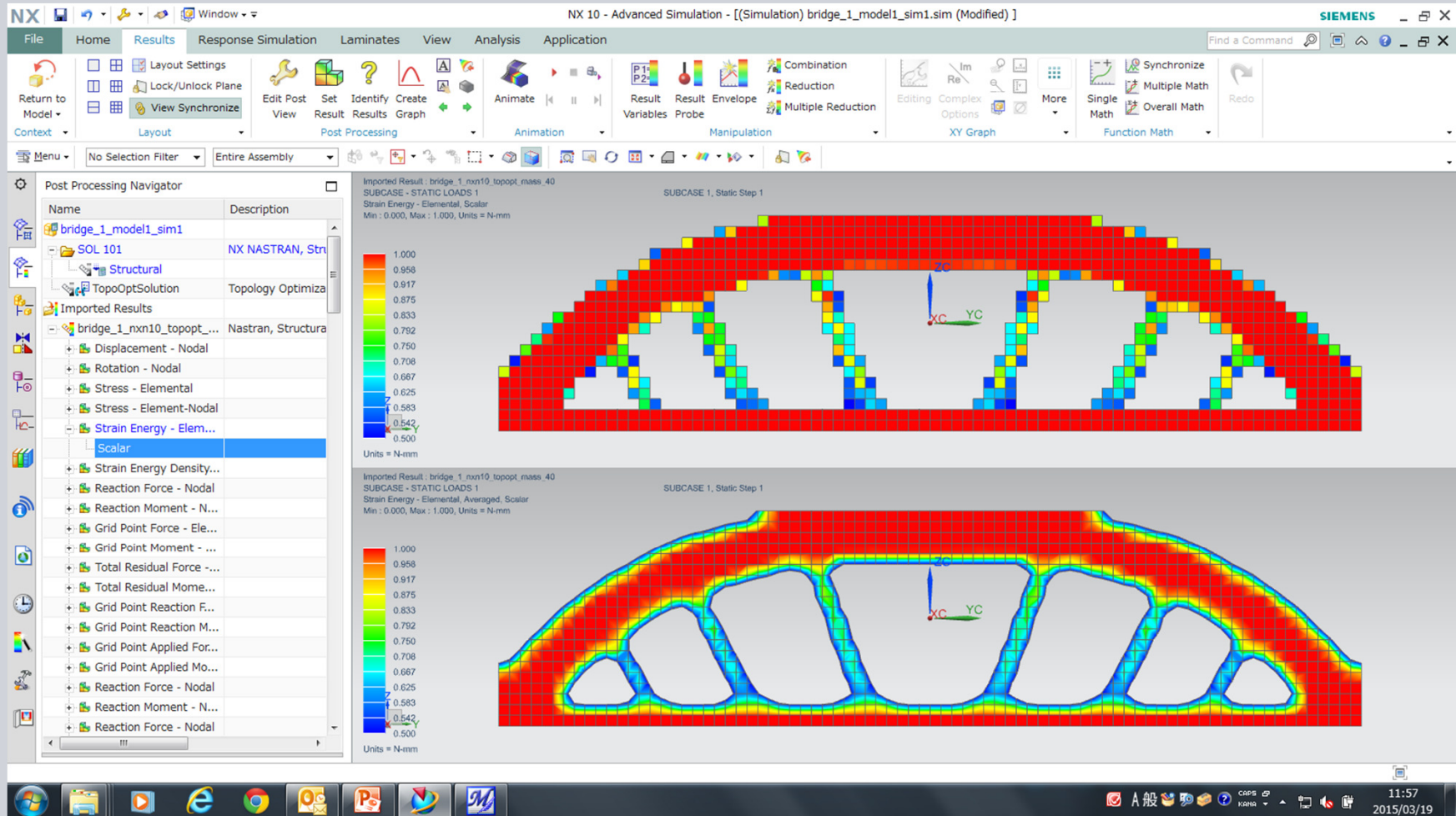
Minimize mass, displacement constraint



Bottom loaded bridge model design space



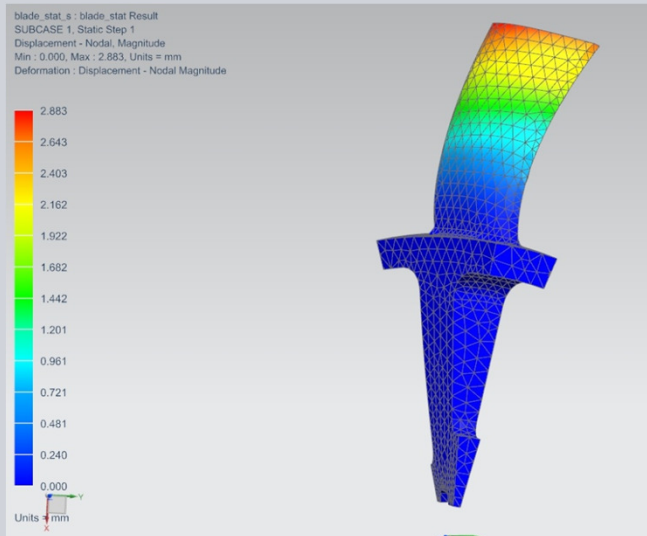
Minimize mass, displacement constraint



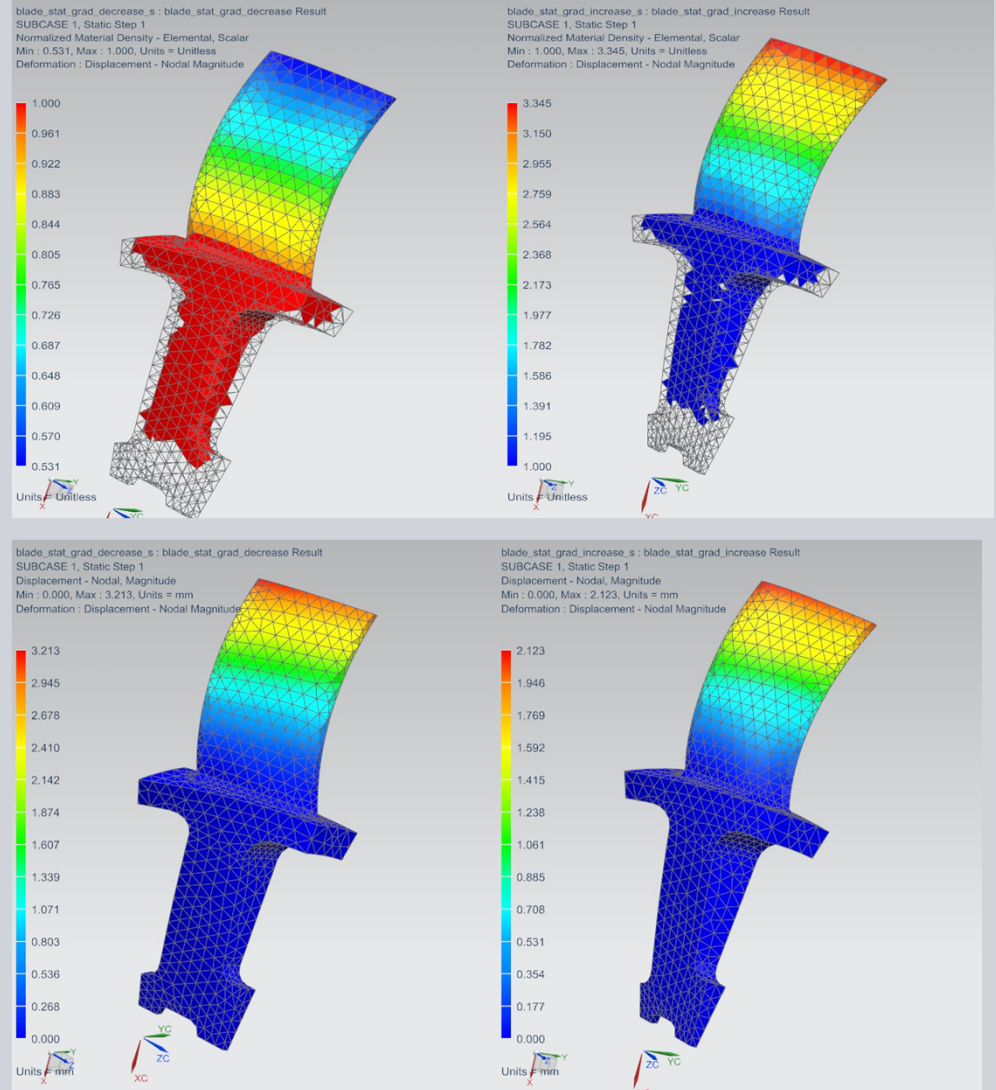
Effect of gradated density to static deformation

Material gradation

Nominal density deformation

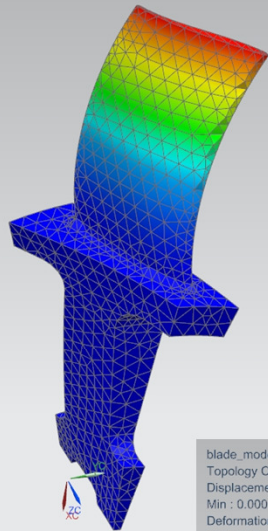


Modified deformations



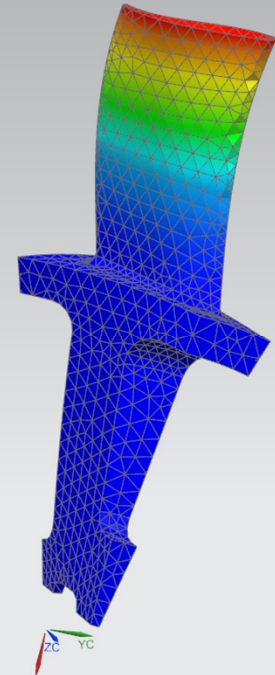
Effect of gradated density to natural frequency

blade_mode.s : blade_mode Result
Load Case 1, Mode 1, 645.825 Hz
Displacement - Nodal, Magnitude
Min : 0.000, Max : 3.019, Units = mm
Deformation : Displacement - Nodal Magnitude



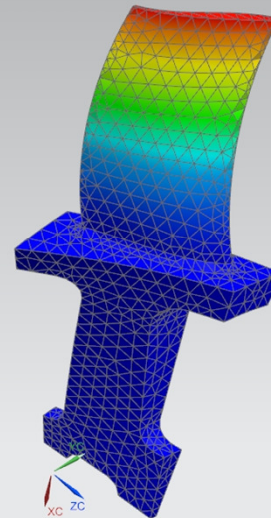
Nominal mode shape

blade_mode_grad_decrease.s : blade_mode_grad_decrease Result
Topology Optimization, Mode 1, 779.834 Hz
Displacement - Nodal, Magnitude
Min : 0.000, Max : 3.870, Units = mm
Deformation : Displacement - Nodal Magnitude



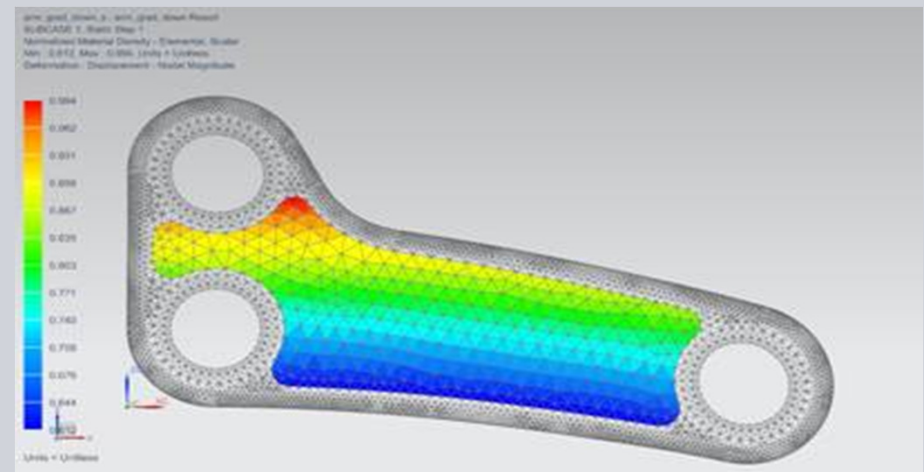
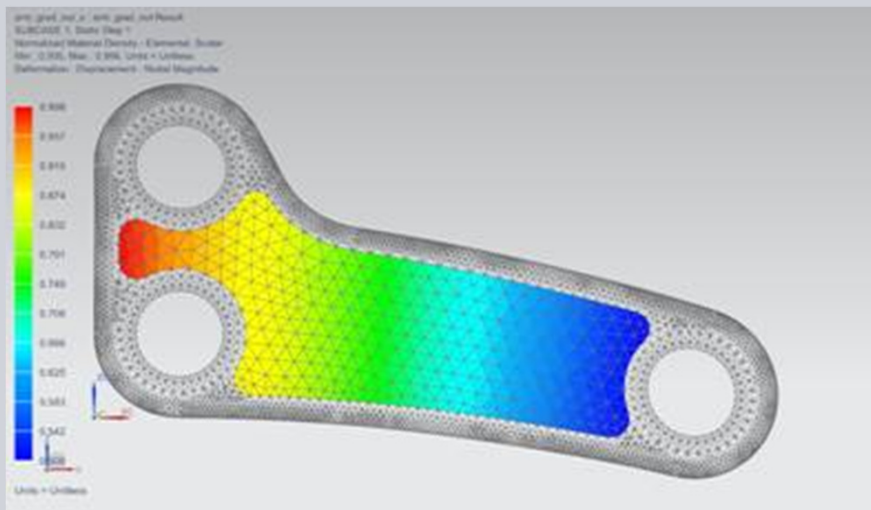
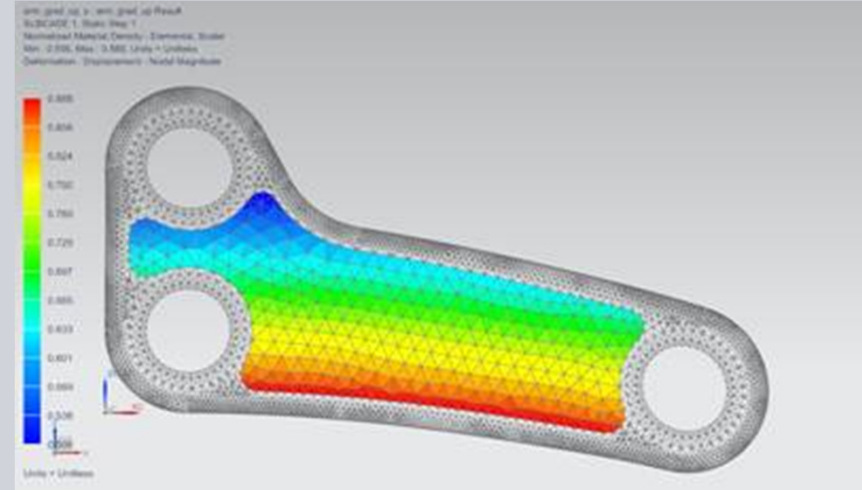
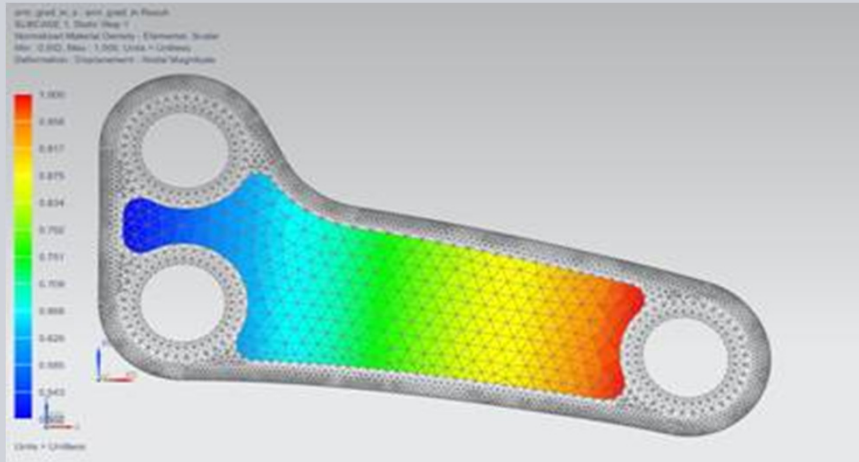
Increasing density inward

blade_mode_grad_increase.s : blade_mode_grad_increase Result
Topology Optimization, Mode 1, 440.877 Hz
Displacement - Nodal, Magnitude
Min : 0.000, Max : 1.736, Units = mm
Deformation : Displacement - Nodal Magnitude



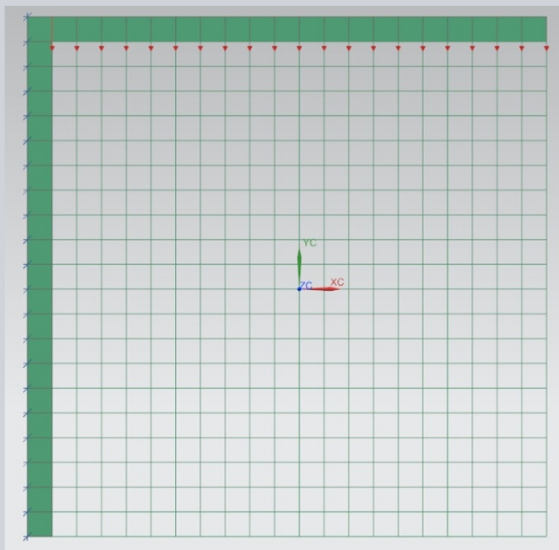
Increasing density outward

Density distribution as design tool



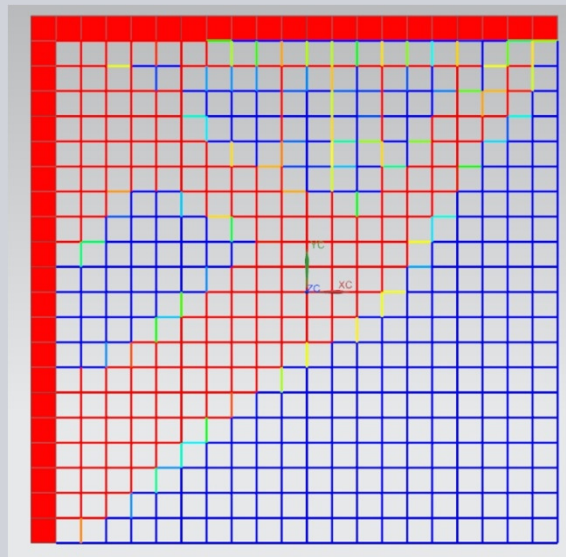
Topology optimization of lattice supported structures

Structure with lattice

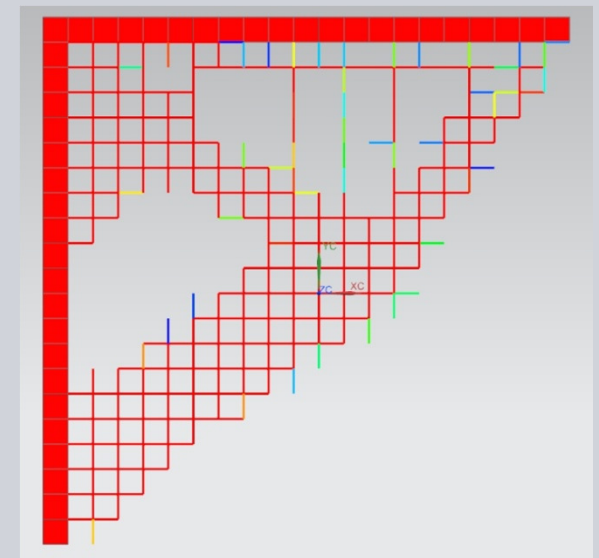


Restricted design space

Optimization result

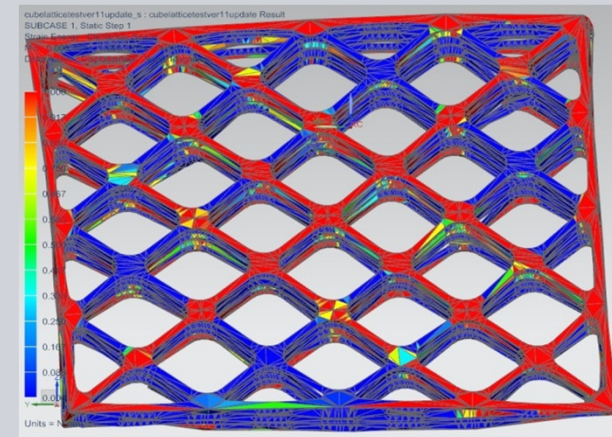
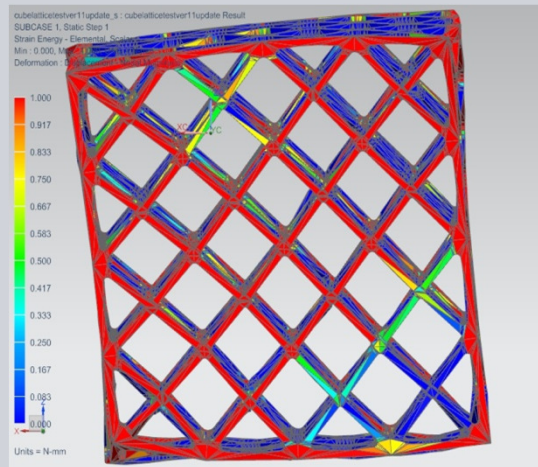
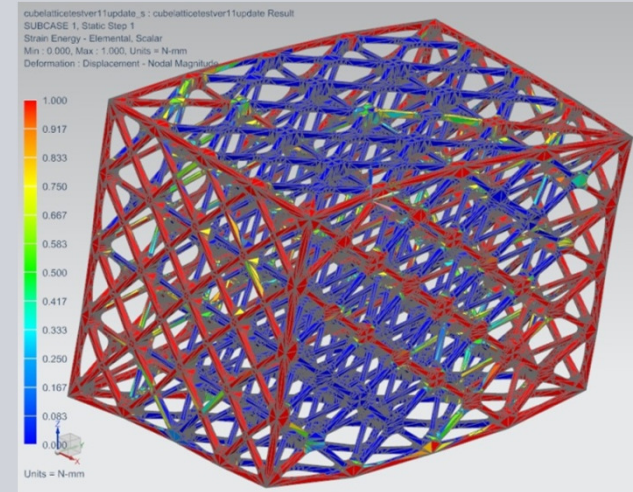
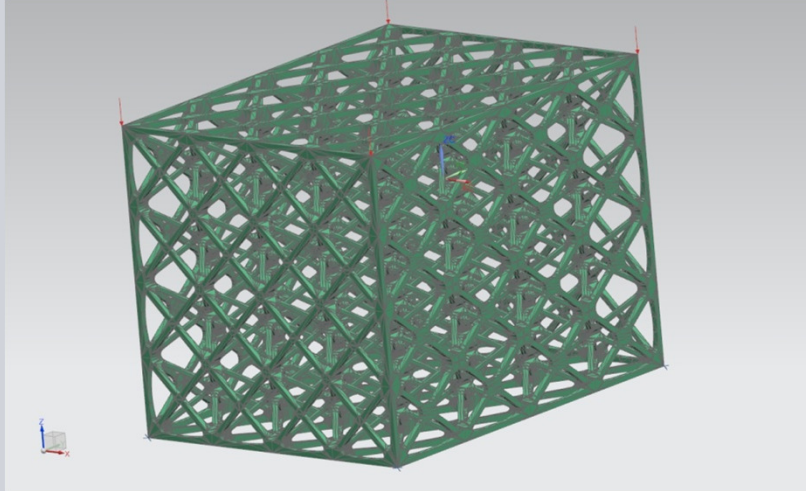


Necessary lattice



Minimal lattice support

Optimization of additive manufacturing lattice cells



Conclusions

Summary

Topology optimization became an everyday tool in engineering by lowering the need of engineering intuition and finite element method enables its implementation

Topology optimization produces empty spaces inside or a graded density distribution but additive manufacturing is able to build cavities directionally vary the density

References

Bendsoe, M. P. and Sigmund, O.: Topology optimization, Theory, Methods and Applications, Springer Verlag, Berlin, 2004

Svanberg, K.: A class of globally convergent optimization methods based on conservative convex separable approximations, SIAM J. of Opt., 2002, 12, 555-573

