

Topology Optimization with Design-dependent Loading: An Adaptive Sensitivity-separation Design Variable Update Scheme

Yang Jiang^a,
Adeildo S. Ramos Jr.^b
Glaucio H. Paulino^a

^aGeorgia Institute of Technology, USA

^bFederal University of Alagoas, Brazil

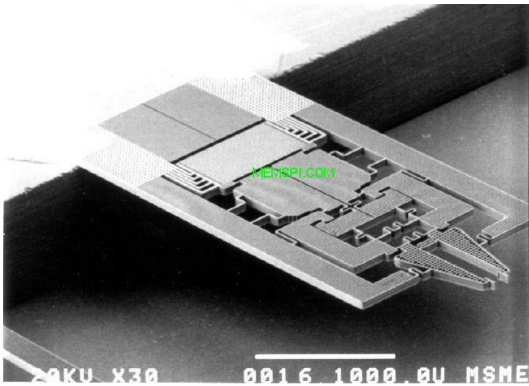
Outline

- **Motivation**
- **Formulations of TopOpt with design-dependent loading**
- **Sensitivity-separation update scheme**
- **Numerical examples**
- **Concluding remarks**

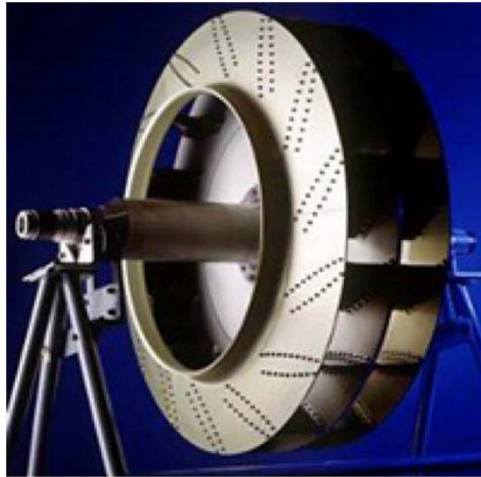
Y. Jiang, A. S. Ramos Jr. and G. H. Paulino (2019) Topology optimization with design-dependent loading: An adaptive sensitivity-separation design variable update scheme (submitted).

Motivation: Design-dependent loading

- Design-dependent loads, including electromagnetic forces, heat load, centrifugal force and gravitational forces, can significantly influence the optimized design.



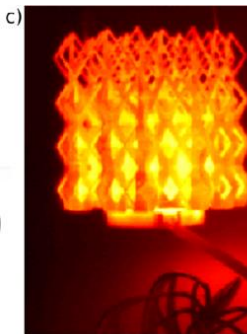
Micro-electro-mechanical system
(www.memspi.com)



Centrifugal fan
(Lee, et al)



Gateway Arch
(www.interestingamerica.com)



Heat sink
(Lazarov, et al)

Topology optimization with design-dependent loading: Solid formulation

- Compliance minimization with a material volume constraint.

$$\min_{\mathbf{z}} \quad C = (\mathbf{f} + \mathbf{g}(\mathbf{z}))^T \mathbf{u}$$

$$\text{s.t.} \quad G = \mathbf{v}^T \mathbf{x} - \bar{V} \leq 0$$

$$z_{min,i} \leq z_i \leq z_{max,i}, \quad i = 1, \dots, n_e$$

$$\text{with} \quad \mathbf{K}(\mathbf{z}) \mathbf{u} = \mathbf{f} + \mathbf{g}(\mathbf{z})$$

$$\mathbf{P}\mathbf{z} = \mathbf{x}$$

- The objective is non-monotonous due to the design-dependent loading.

$$\frac{\partial C}{\partial \mathbf{z}} = -\mathbf{u}^T \frac{\partial \mathbf{K}}{\partial \mathbf{z}} \mathbf{u} + \boxed{2\mathbf{u}^T \frac{\partial \mathbf{g}}{\partial \mathbf{z}}}$$

Topology optimization with design-dependent loading: Void formulation

- Void formulation applies an upper bound on the void region, i.e., a lower bound on material volume.

$$\min_{\mathbf{z}} \quad C = (\mathbf{f} + \mathbf{g}(\mathbf{z}))^T \mathbf{u}$$

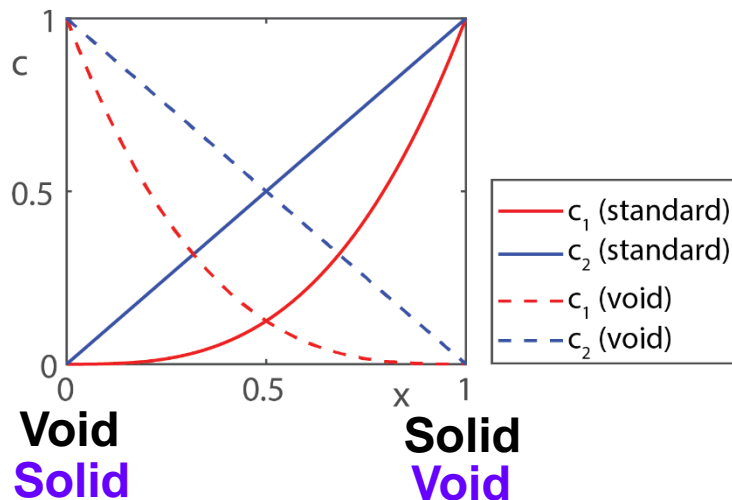
$$\text{s.t.} \quad G_{void} = \mathbf{v}^T \mathbf{x} - \bar{V}_{void} \leq 0$$

$$z_{min,i} \leq z_i \leq z_{max,i}, \quad i = 1, \dots, n_e$$

$$\text{with} \quad \mathbf{K}(\mathbf{z}) \mathbf{u} = \mathbf{f} + \mathbf{g}(\mathbf{z})$$

$$\mathbf{P}\mathbf{z} = \mathbf{x}$$

- Void formulation uses a reflected material property interpolation function



**Solid
formulation**

**Void
formulation**

$$E_k = x_k^p E_0, p \geq 1$$

$$g_{k,V} = -\frac{1}{n_{node,k}} \gamma v_k x_k.$$

$$E_k = (1 - x_k)^p E_0, p \geq 1$$

$$g_{k,V} = -\frac{1}{n_{node,k}} \gamma v_k (1 - x_k).$$

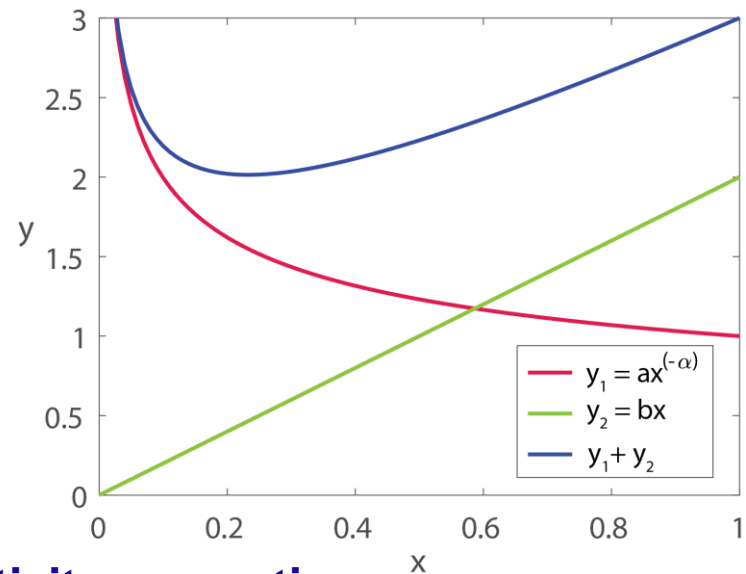
Sensitivity-separation update scheme: Constructing sub-problems

- The objective function is approximated by the sum of non-monotonous separable functions.

$$C(\mathbf{z}_c) \approx C_{app}(\mathbf{z}_c)$$

$$= \sum_{i=1}^{n_e} \left(a_i z_{c,i}^{-\alpha} + b_i z_{c,i} \right) + c$$

where $a_i \geq 0$, $b_i \geq 0$, $\alpha > 0$.



- The construction is enabled by sensitivity separation.

$$\frac{\partial C_{app}}{\partial z_i} = \left[-\alpha a_i (z_{c,i})^{-\alpha-1} \right] + \left[b_i \right]$$

$$\frac{\partial C}{\partial z_i} = \left[\left(\frac{\partial C}{\partial z_i} \right)_n \right] + \left[\left(\frac{\partial C}{\partial z_i} \right)_p \right]$$

$$a_i = - \left(\frac{\partial C}{\partial z_i} \right)_n (z_{c,i})^{\alpha+1} / \alpha$$

$$b_i = \left(\frac{\partial C}{\partial z_i} \right)_p$$

Sensitivity-separation update scheme: Solving the approximate sub-problem

- The update is obtained by solving a Lagrangian dual problem.

$$\begin{aligned} \min_{\mathbf{z}} \quad & \sum_{i=1}^{n_e} (a_i z_i^{-\alpha} + b_i z_i) \\ \text{s.t.} \quad & G = \mathbf{v}^T \mathbf{x} - \bar{V} \leq 0 \\ & z_{\min,i} \leq z_i \leq z_{\max,i}, \quad i = 1, \dots, n_e \end{aligned}$$

with $\mathbf{P}\mathbf{z} = \mathbf{x}$

$$a_i = - \left(\frac{\partial C}{\partial z_i} \right)_n (z_{c,i})^{\alpha+1} / \alpha$$

$$b_i = \left(\frac{\partial C}{\partial z_i} \right)_p$$

$$L(\mathbf{z}, \lambda) = \sum_{i=1}^{n_e} (a_i z_i^{-\alpha} + b_i z_i) + \lambda G$$



$$\frac{\partial L}{\partial z_i} = -\alpha a_i (z_i)^{-\alpha-1} + b_i + \lambda \frac{\partial G}{\partial z_i} = 0$$

$$\Downarrow \quad \eta = 1/(\alpha+1)$$

$$z_{opt}(\lambda) = \left(\frac{a_i}{b_i + \lambda \frac{\partial G}{\partial z_i}} \right)^\eta = \left(\frac{- \left(\frac{\partial C}{\partial z_i} \right)_n}{\left(\frac{\partial C}{\partial z_i} \right)_p + \lambda \frac{\partial G}{\partial z_i}} \right)^\eta z_{i,c}$$



$$z_i(\lambda) = \begin{cases} \underline{z}_i, & \text{if } z_{opt} < \underline{z}_i \\ z_{opt}, & \text{if } \underline{z}_i \leq z_{opt} \leq \bar{z}_i \\ \bar{z}_i, & \text{if } z_{opt} > \bar{z}_i \end{cases}$$

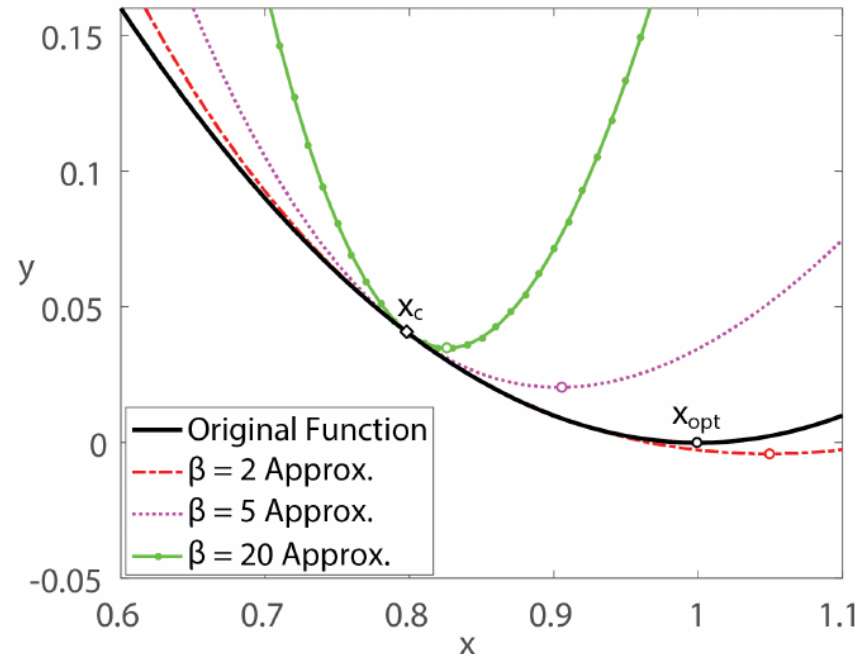
- The single variable λ can be efficiently obtained by bisection method.

Sensitivity-separation update scheme: Adaptive sensitivity separation by β

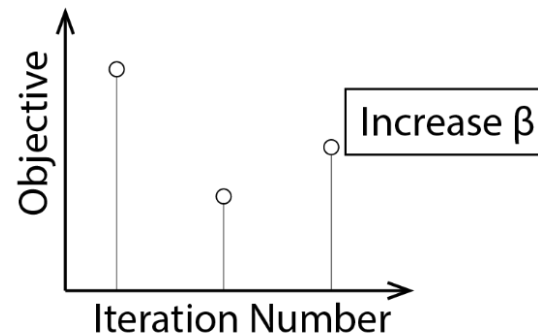
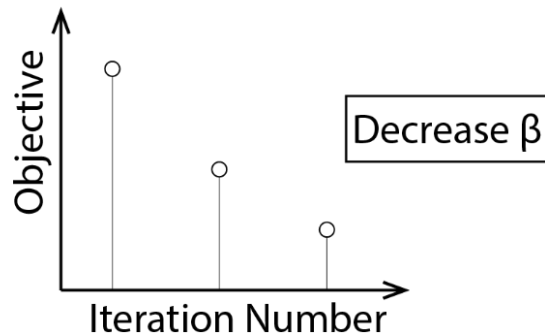
- The larger β is, the more conservative the update is.

$$\left(\frac{\partial C}{\partial z_i}\right)_n = \begin{cases} (1 + \beta) \frac{\partial C}{\partial z_i}, & \text{if } \left(\frac{\partial C}{\partial z_i}\right) \leq 0 \\ (-\beta) \frac{\partial C}{\partial z_i}, & \text{if } \left(\frac{\partial C}{\partial z_i}\right) > 0 \end{cases}$$

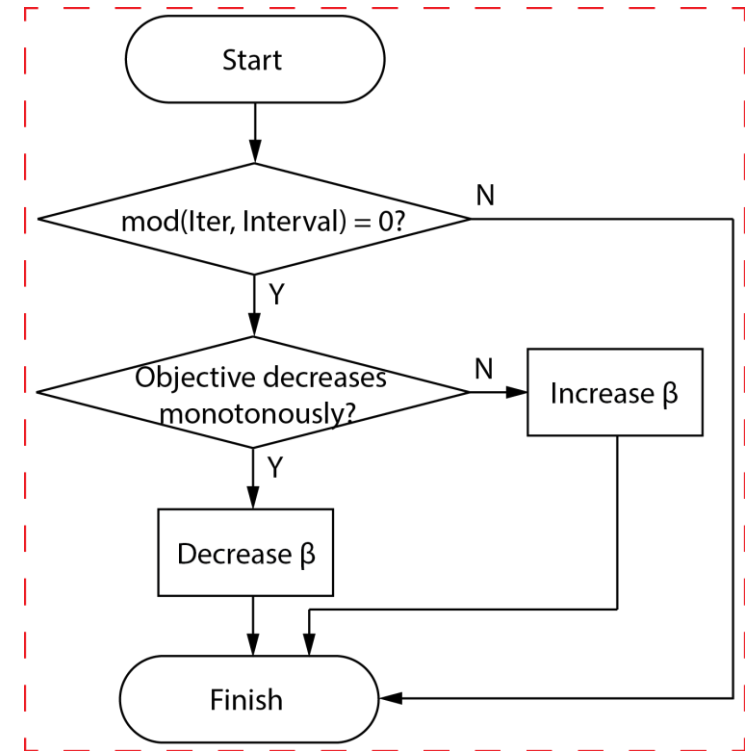
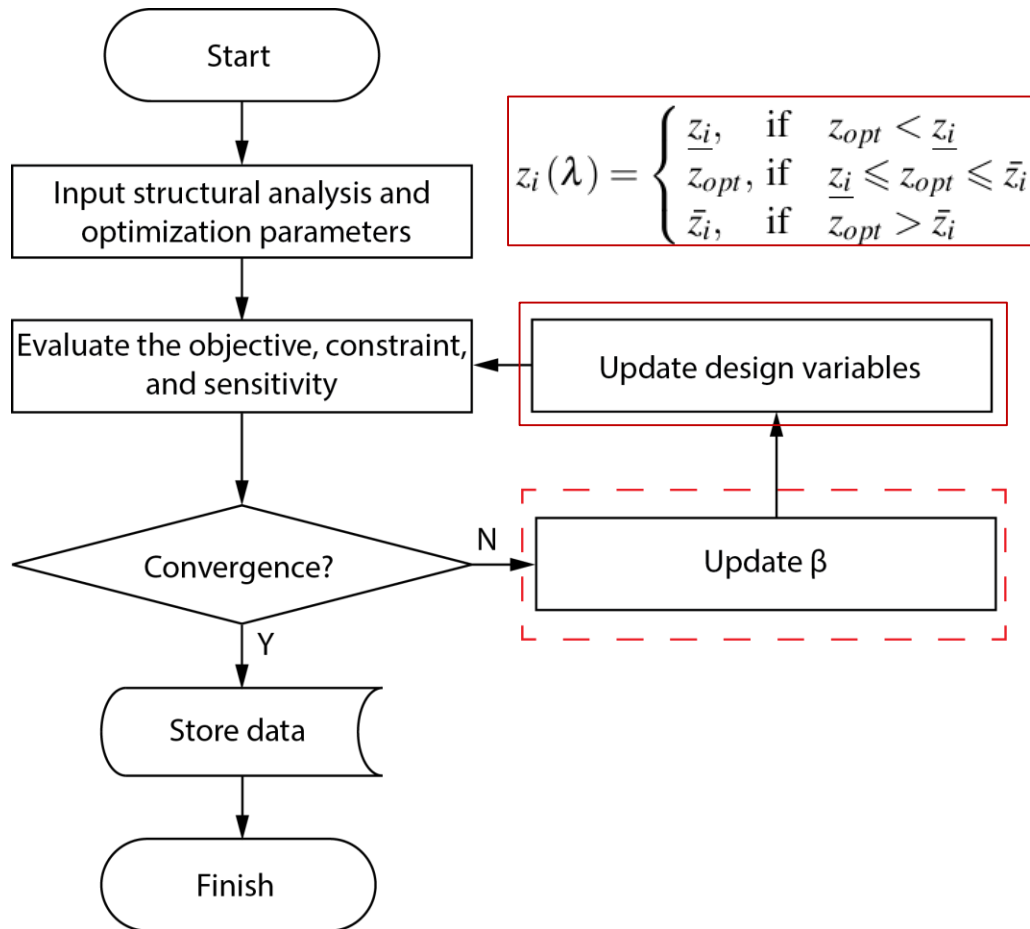
$$\left(\frac{\partial C}{\partial z_i}\right)_p = \begin{cases} (-\beta) \frac{\partial C}{\partial z_i}, & \text{if } \left(\frac{\partial C}{\partial z_i}\right) \leq 0 \\ (1 + \beta) \frac{\partial C}{\partial z_i}, & \text{if } \left(\frac{\partial C}{\partial z_i}\right) > 0 \end{cases}$$



- The sensitivity separation is adaptive as we update β adaptively.

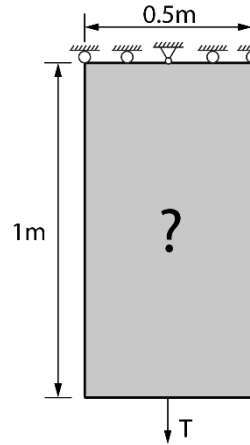


Algorithm: TopOpt with adaptive sensitivity separation



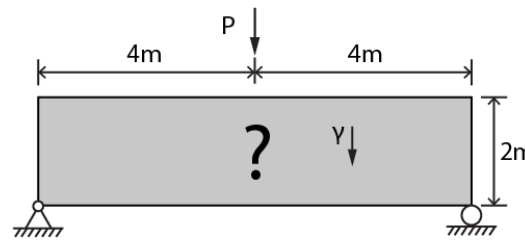
Numerical examples

Hanging bar



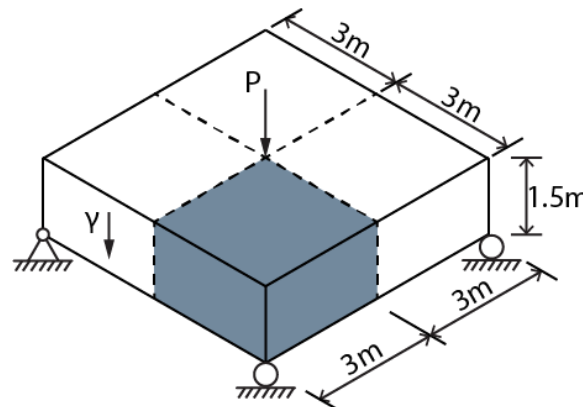
Verification problem

2D MBB beam



2D design under various loading scenarios

3D Dome

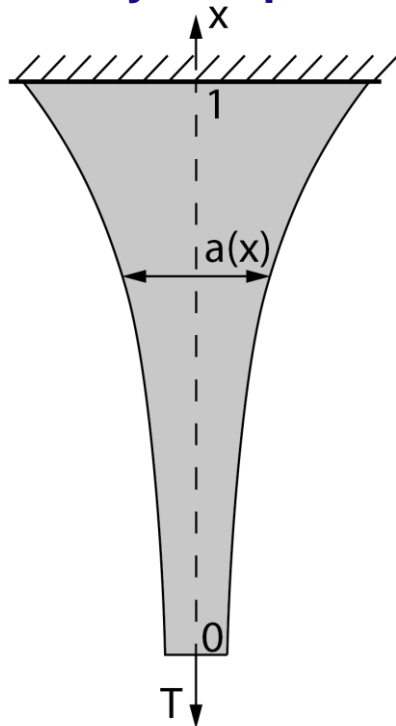


3D design extension

Hanging bar

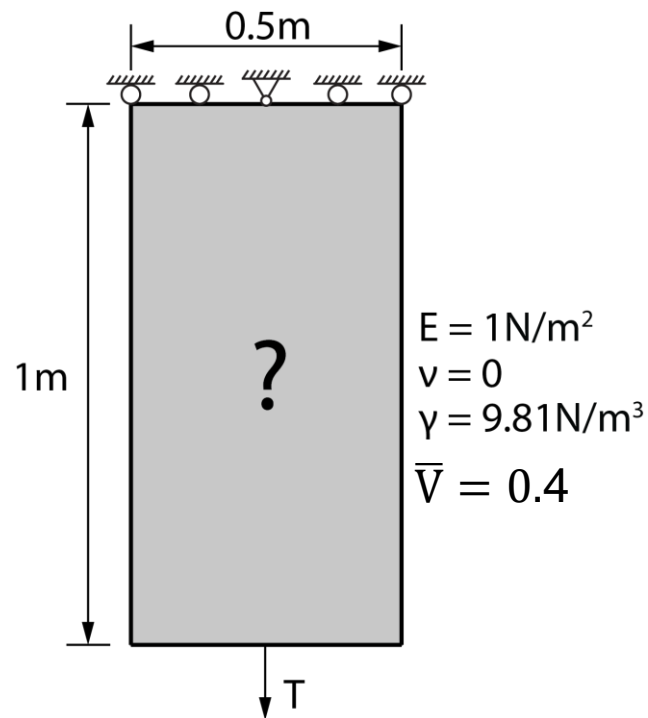
- Cross-sectional area distribution for minimum tip displacement

1D analytical problem



$$a(x) = \left(\frac{T}{\gamma}\right) \ln\left(1 + \bar{V} \frac{\gamma}{T}\right) e^{x \ln\left(1 + \bar{V} \frac{\gamma}{T}\right)}$$

2D numerical problem



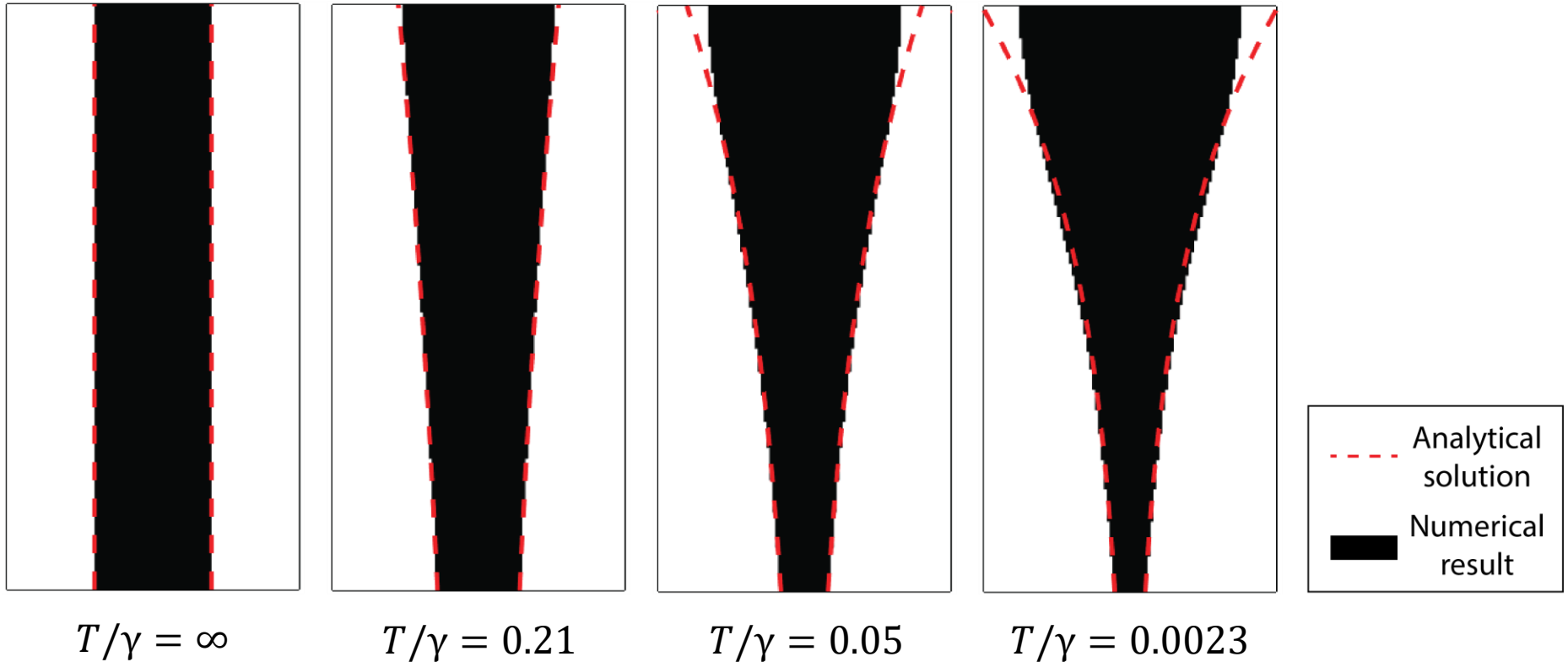
Objective: $C = \mathbf{f}^T \mathbf{u}$

B.L. Karihaloo, W.S. Hemp, Maximum strength/stiffness design of structural members in presence of self-weight, Proc. R. Soc. Lond. A 389, 119-132 (1983)

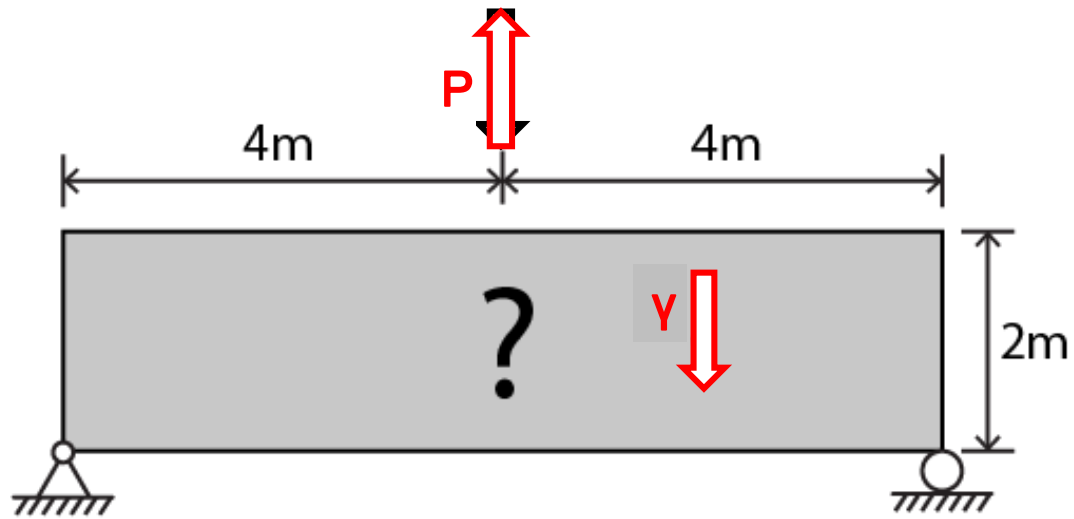
B. Zheng, C. Chang and H. C. Gea, Topology optimization considering body forces, Int. J. Simul. Multidisci. Des. Optim. 3, 316-320 (2009)

Hanging bar

- The numerical results match the analytical solution.



MBB beam



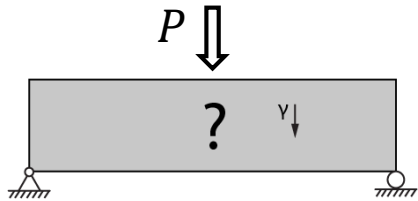
$$E = 30000 \text{ kN/m}^2$$

$$\nu = 0.3$$

$$\gamma = 23.5 \text{ kN/m}^3$$

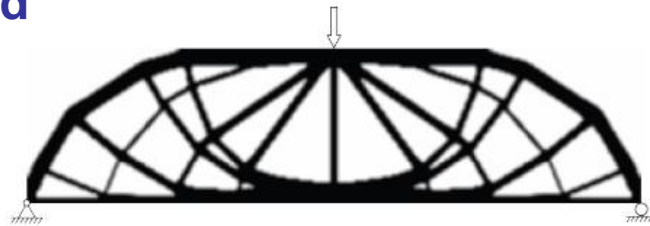
MBB beam under different loading scenarios

Downward external load

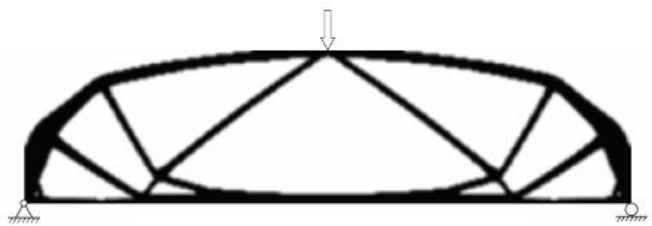


$$\bar{V} = 0.38$$

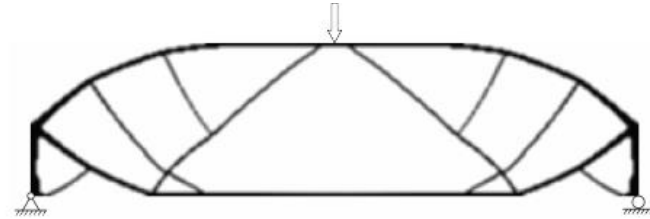
$$W = \gamma * V_{\text{total}}$$



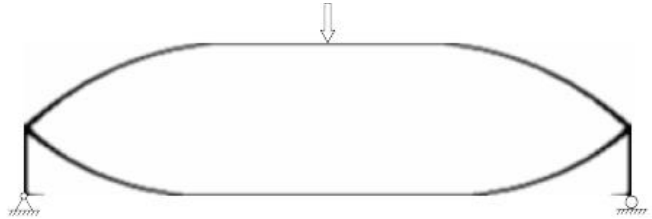
$$P/W = \infty, V = 0.38$$



$$P/W = 1, V = 0.25$$

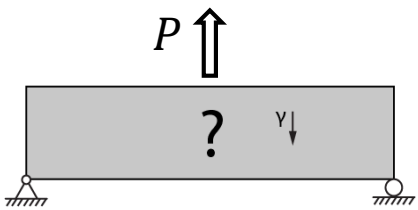


$$P/W = 0.005, V = 0.12$$



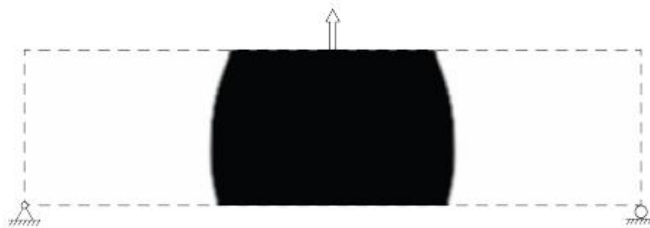
$$P/W = 0.0015, V = 0.049$$

Upward external load

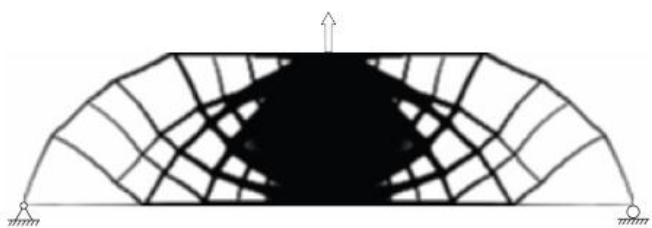


$$\bar{V} = 0.38$$

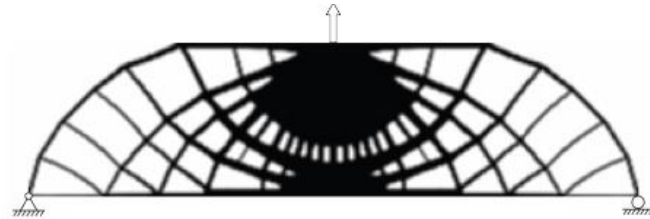
$$W = \gamma * V_{\text{total}}$$



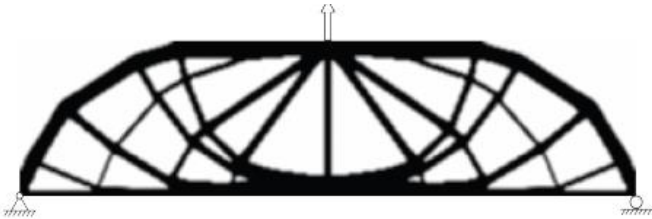
$$P/W = -0.38, V = 0.38$$



$$P/W = -0.383, V = 0.38$$



$$P/W = -0.4, V = 0.38$$

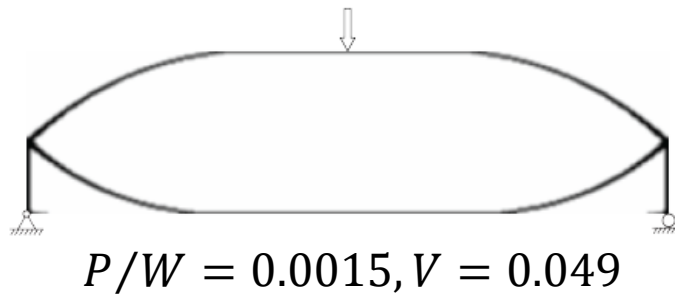


$$P/W = -\infty, V = 0.38$$

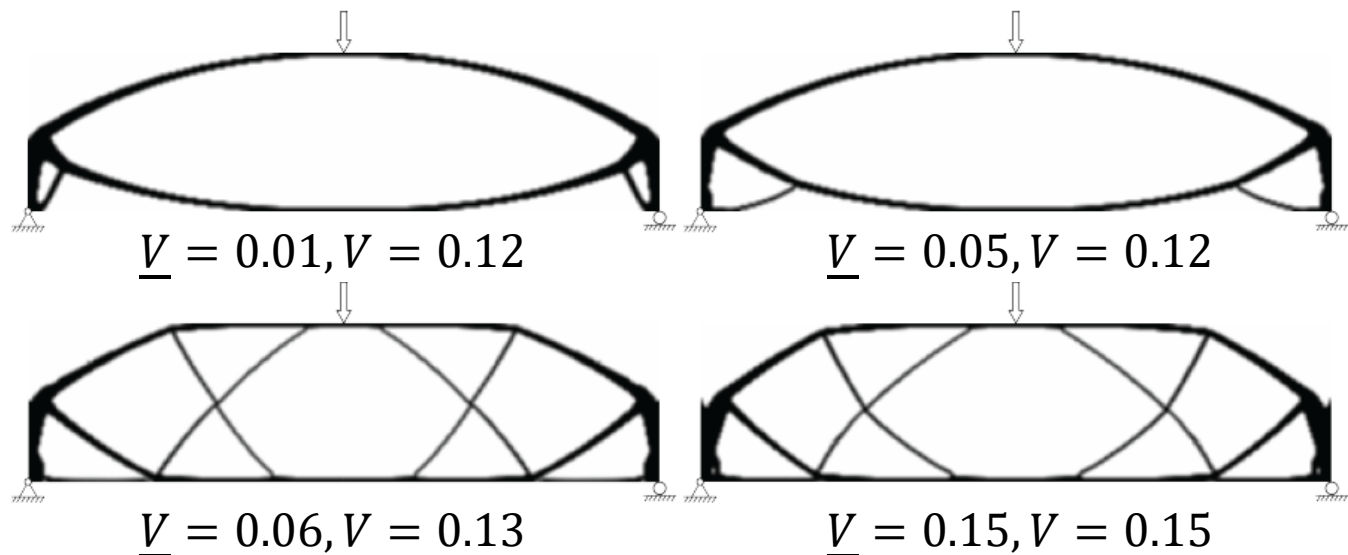
MBB beam design under large design-dependent loading

- In cases dominated by design-dependent loading, void formulation may provide feasible designs.

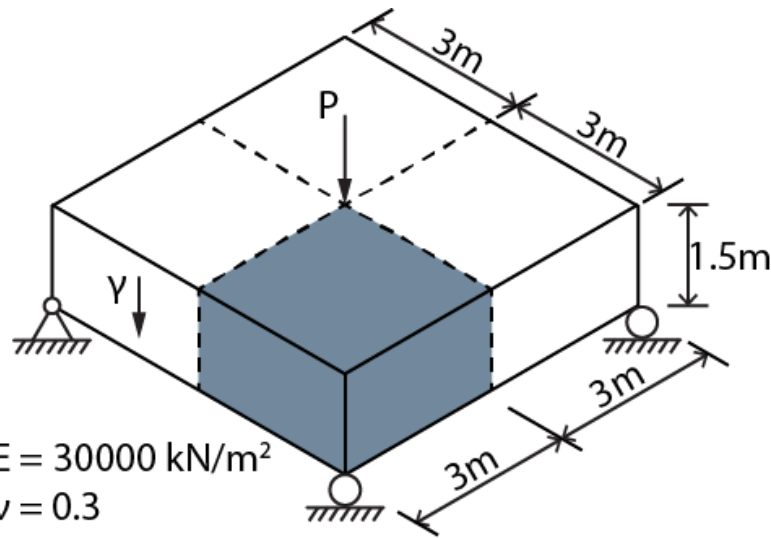
From solid formulation



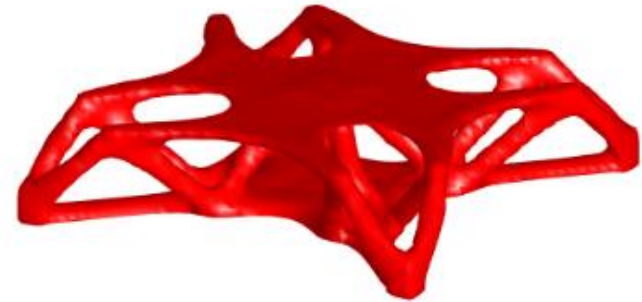
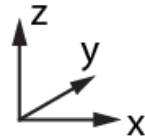
From void formulation



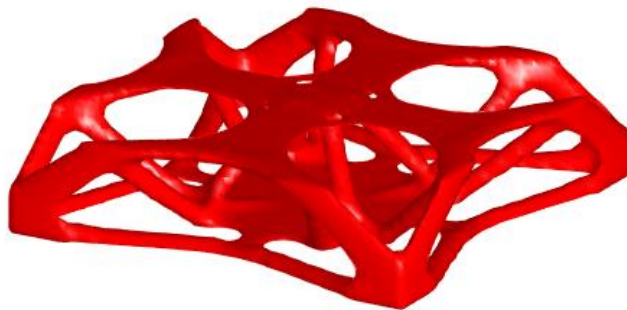
3D Dome



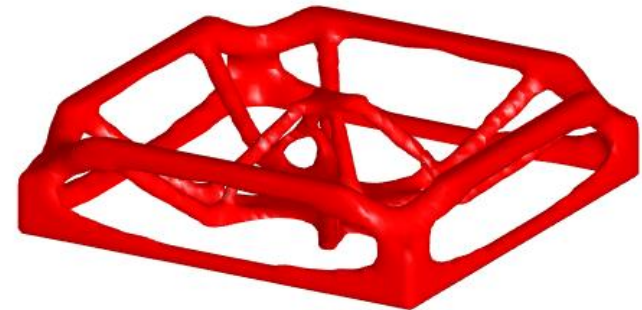
$E = 30000 \text{ kN/m}^2$
 $\nu = 0.3$
 $\gamma = 23.5 \text{ kN/m}^3$
 $\bar{V} = 0.15$
 $W = \gamma V_{\text{total}}$



$P/W = 1, V = 0.15$

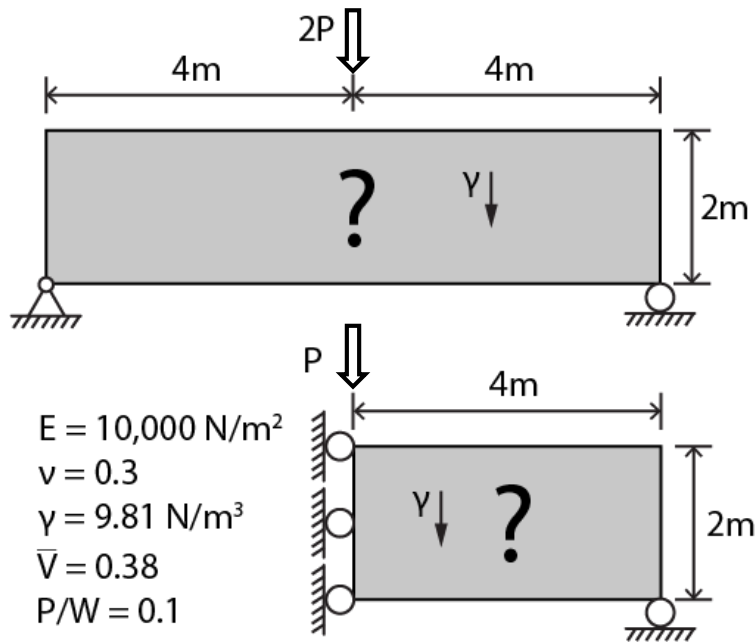


$P/W = 0.1, V = 0.15$



$P/W = 0.025, V = 0.11$

Computational efficiency comparison



		40X20 mesh	80X40 mesh	160X80 mesh
GCMMA	Objective value	7.161	6.049	5.475
	Number of iterations	65	258	500
	Update time (s)	5	75	1329
	Total time (s)	7	109	1895
Sensitivity Separation	Objective value	7.179	6.253	5.705
	Number of iterations	114	203	259
	Update time (s)	0.2	0.5	1
	Total time (s)	4	26	100

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

M. Bruyneel and P. Duysinx. "Note on topology optimization of continuum structures including self-weight". Structural and Multidisciplinary Optimization 29.4 (2005), pp. 245–256.

Concluding remarks

- ❑ **To solve topology optimization problems with design-dependent loadings, we propose a novel design variable update scheme, which is based on sensitivity separation.**
- ❑ **This scheme solves problems with a material volume upper bound (solid formulation) or lower bound (void formulation).**
- ❑ **Numerical examples in 2D and 3D demonstrate that the update scheme is efficient and is able to handle different loading scenarios.**

Y. Jiang, A. S. Ramos Jr. and G. H. Paulino (2019) Topology optimization with design-dependent loading: An adaptive sensitivity-separation design variable update scheme (submitted).

Topology Optimization with Design-dependent Loading: An Adaptive Sensitivity-separation Design Variable Update Scheme

Thank you!