A sensitivity interpolation algorithm for the concurrent optimization of bodies sharing a common design space

Giorgio Previati¹ · Massimiliano Gobbi¹ · Federico Ballo¹

Received: 29 January 2020 / Revised: 4 November 2020 / Accepted: 10 March 2021 /
Published online: 26 April 2021
© The Author(s) 2021

Abstract

In this paper the problem of the concurrent topological optimization of two different bodies sharing a region of the design space is dealt with. This design problem focuses on the simultaneous optimization of two bodies (components) where not only the material distribution of each body has to be optimized but also the design space has to be divided among the two bodies. This novel optimization formulation represents a design problem in which more than one component have to be located inside a limited allowable room. Each component has its own function and load carrying requirements. In the paper a novel development solution algorithm is presented. With respect to previously published papers, the new algorithm comprises an interpolation of the density fields which allows a complete independence of the meshes of the two bodies. As the bodies can be meshed with any arbitrary mesh, this new algorithm can be applied to any real geometry. The developed algorithm is used to design a complex three dimensional system, namely a multi-component arm for a tube bending machine.

Keywords Structural optimization · Topology optimization · Multi-component system optimization · SIMP

1 Introduction

In the literature many different engineering problems referring to the topological optimization of structural components can be found [see for instance Rozvany (2009), Guo and Cheng (2010), Eschenauer and Olhoff (2001), Bendsøe and

✉ Giorgio Previati
giorgio.previati@polimi.it

Massimiliano Gobbi
massimiliano.gobbi@polimi.it

Federico Ballo
federicomaria.ballo@polimi.it

¹ Department of Mechanical Engineering, Politecnico di Milano, 20158 Milan, Italy

Sigmund (2004), Zhu et al. (2016) and Zhang et al. (2016b)]. In most cases, only one component is considered. Recently, however, problems involving multi-domain optimization, such as the design of multi-phase or multi-material structures (Tavakoli and Mohseni 2014), have been considered.

Referring to the particular problem of the optimization of systems composed by two bodies sharing the same design domain, some applications can be found. In Lawry and Maute (2018) and Strömberg (2018), the level set method is used for the optimization of the interface between two, or more, phases (components) in order to obtain some prescribed interaction force. In Zhang et al. (2016a), a completely different approach, the Moving Morphable Components method, which seems to be adaptable to solve such problems, is discussed.

In Prevati et al. (2018) and Ballo et al. (2019), the authors have presented a simple algorithm based on the SIMP approach (Rozvany 2009; Bendsoe and Sigmund 2004). The method is able to optimize the material distribution of two different bodies while, at the same time, allocate in an efficient way the space shared between the bodies. The algorithm has also proved to be easily implementable and can be used along with available optimization algorithms. The applicability of the algorithm is limited to bodies presenting the same mesh in the shared portion of the domain. Such limitation prevents the utilization of the algorithm on arbitrary shaped optimization domains.

This paper presents a newly developed algorithm. By considering a density field interpolation on the shared part of the domain, the limitation regarding the correspondence of the elements of the two meshes can be removed. By removing this limitation, the applicability of the method is extended to any design domain and any kind of finite element, as explained in Sects. 2 and 3 of the paper. In Sect. 4, the new algorithm will be tested by considering a two-dimensional example. Then, in Sect. 5, the application of the presented algorithm to the optimization of a multi-component arm for a tube bending machine will be shown.

2 Problem formulation and numerical solution

The formulation of the concurrent structural optimization of two bodies sharing the same design space has been given and discussed in Prevati et al. (2018). In this section, it is briefly summarized.

Figure 1 depicts the two bodies sharing (a portion of) the design space. The optimization process consists in finding the optimal material distribution for each body along with the optimal division of the shared part of the domain space among the two bodies. The two optimization problems are solved simultaneously and the allocation of the shared part of the domain changes accordingly to the actual structural configuration of the two bodies. The following nomenclature is considered.

- Ω_1 and Ω_2 : bodies to be optimized.
- Ω_{1-2} : shared portion of the design space that can be occupied by any of the two bodies (or left void). Any given point of this region can be occupied by only one of the two bodies.

Fig. 1 Generalized geometries of the design domains (Ω_1 and Ω_2) of two bodies sharing a portion of the design space. Each body has its own system of applied forces ($\mathbf{f}_1$ and $\mathbf{f}_2$) and boundary constraints (Γ_1 and Γ_2). Ω_{1-2} represents the shared portion of the design space. Left: initial definition of the domains. Right: division of the domains with assignment of the shared portion of the design space to each body

- Ω_1^* and Ω_2^* : the two unshared parts of the domains Ω_1 and Ω_2 respectively (s.t. $\Omega_1 = \Omega_1^* \cup \Omega_{1-2}$ and $\Omega_2 = \Omega_2^* \cup \Omega_{1-2}$).
- $\Omega_{1-2}^{(1)}$ and $\Omega_{1-2}^{(2)}$: portion of the shared design space assigned to Ω_1 or Ω_2 respectively (s.t. $\Omega_{1-2}^{(1)} \cup \Omega_{1-2}^{(2)} = \Omega_{1-2}$ and $\Omega_{1-2}^{(1)} \cap \Omega_{1-2}^{(2)} = \emptyset$).
- $\overline{\Omega}_1$ and $\overline{\Omega}_2$: actual design space of the two bodies, given by $\overline{\Omega}_1 = \Omega_1^* \cup \Omega_{1-2}^{(1)}$ and $\overline{\Omega}_2 = \Omega_2^* \cup \Omega_{1-2}^{(2)}$.
- $\mathbf{f}_1$ and $\mathbf{f}_2$: applied load on Ω_1 and Ω_2 respectively.
- Γ_1 and Γ_2 : boundary conditions on Ω_1 and Ω_2 respectively.

Obviously, for a physical meaningful solution of the problem, $\overline{\Omega}_1$ and $\overline{\Omega}_2$ must be connected.

It must be emphasized that the portion of the shared domain Ω_{1-2} to be allocated to each body (i.e. $\Omega_{1-2}^{(1)}$ and $\Omega_{1-2}^{(2)}$) is not given a priori, but must be allocated during the optimization process and can change according to the evolution of the shapes of the two bodies themselves.

Under the hypotheses of

- linear elastic bodies with small deformations;
- the two bodies do not interact in the shared portion of the domain (i.e. no contact is considered between the two bodies in the shared portion of the domain, interactions between the two bodies outside this region is possible);
- each body is made by only one isotropic material;
- both materials have the same reference density, but they can have different elastic moduli;
- loads and boundary conditions can be applied at any location of the domains, even in the shared portion;
- the structural problem is formulated by the finite element theory;

the problem can be stated in the framework of finite elements as

$$\begin{aligned}
& \text{Find} \quad \min_{\mathbf{u}_1, \mathbf{u}_2, \rho} (\mathbf{f}_1^T \mathbf{u}_1 + \mathbf{f}_2^T \mathbf{u}_2) \\
& \text{s.t. :} \quad \mathbf{K}(E_e^1) \mathbf{u}_1 = \mathbf{f}_1 \text{ and } \mathbf{K}(E_e^2) \mathbf{u}_2 = \mathbf{f}_2 \\
& \quad E_e^1 = \rho(x_e)^p E_e^{*1}, \quad x_e \in \overline{\Omega_1} \\
& \quad E_e^2 = \rho(x_e)^p E_e^{*2}, \quad x_e \in \overline{\Omega_2} \\
& \quad \Omega_{1-2}^{(1)} \cup \Omega_{1-2}^{(2)} = \Omega_{1-2} \text{ and } \Omega_{1-2}^{(1)} \cap \Omega_{1-2}^{(2)} = \emptyset \\
& \quad \sum_{e=1}^{N_1} \rho(x_e) + \sum_{e=1}^{N_2} \rho(x_e) \leq V \quad 0 < \rho_{\min} \leq \rho \leq 1 \\
& \text{where :} \quad \overline{\Omega_1} = \sum_{e=1}^{N_1} \mathbf{K}_e(E_e^1) \text{ and } \mathbf{K}(E_e^2) = \sum_{e=1}^{N_2} \mathbf{K}_e(E_e^2) \\
& \quad \overline{\Omega_1} \text{ and } \overline{\Omega_2} \text{ connected}
\end{aligned} \tag{1}$$

where $\mathbf{u}_1$ and $\mathbf{u}_2$ are the displacement fields of the two bodies, x_e are the coordinates of the centre of the considered element, $\mathbf{K}$ is the stiffness matrix of each body with $\mathbf{K}_e$ the element stiffness matrix, E_e^1 and E_e^2 are the elastic moduli of each element of the two bodies, with E_e^{*1} and E_e^{*2} reference elastic moduli, N_1 and N_2 are the number of elements of each body and ρ is the pseudodensity, with p the penalty term.

In Prevati et al. (2018); Ballo et al. (2019), the authors have presented a solution algorithm able to solve the problem stated in Eq. 1 under the condition that the two bodies present the same mesh in the shared part of the domain, i.e. in this region there is a one-to-one correspondence of each element of the two bodies. This condition introduces strong limitations to the applicability of the method to actual engineering problems, as the designer is forced to choose the same mesh for the two bodies in the shared portion of the design domain, regardless of the actual shape of the bodies themselves. In the following, this condition is removed, thus leaving the designer total freedom in selecting the finite element mesh that best copes with the geometry of each single body.

The new solution algorithm is described in the diagram reported in Fig. 2. The algorithm is basically divided in two parts, namely sub-algorithm (a) and sub-algorithm (b). The two parts are highlighted by the two dashed rectangles of Fig. 2. Sub-algorithm (a) is based on the standard structural optimization algorithm derived in Sigmund (2001). The innovative part of the algorithm is mostly in the sub-algorithm (b) and is related to the subdivision of the shared part of the domain.

Referring to the diagram of Fig. 2, the algorithm starts from *Block 1* with the solution of the FEM model on the two domains Ω_1 and Ω_2 . The solution of the FEM model can be done in a single analysis, if interaction is present between the two bodies, or in two separate analyses, if no interaction is present. Nonlinear FE analysis can be employed.

After the FEM models have been solved, in *Block A1*, the sensitivities of the total compliance of each element with respect to the pseudo-density are computed as

$$\frac{\partial c}{\partial x_e} = -p \rho(x_e)^{p-1} \mathbf{u}_e^T \mathbf{K}_e^* \mathbf{u}_e \tag{2}$$

Fig. 2 Diagram of the algorithm for the concurrent topological optimization. Symbols refer to Fig. 1

It must be emphasized that sensitivities are computed for all the elements of Ω_1 and Ω_2 . In particular, the shared part of the domain Ω_{1-2} is described by two sensitivity

fields, one with respect to the elements of the first body on Ω_{1-2} and one with respect to the elements of the second body on Ω_{1-2} .

The two sensitivity fields computed on the shared part of the domain Ω_{1-2} have now to be compared in order to divide the domain between the two bodies in the most effective way. In the original version of the algorithm (Prevati et al. 2018), this comparison was done elementwise. The common part of the domain was meshed in the same way with respect to the two bodies and each element was assigned to the body that exhibits the highest sensitivity. This, of course, was a quite strong limitation to the actual applicability of the method to real world problems with complex geometry. In this new version, the comparison is done pointwise and independently of the meshes of the two bodies. The procedure is done in *Block B1* and *Block B2*. Firstly, in *Block B1*, the two sensitivity fields on Ω_{1-2} are interpolated on a given grid of points, which is common for the two bodies. Then, in *Block B2*, the sensitivities are compared pointwise. On the basis of this comparison, each point is attributed either to $\Omega_{1-2}^{(1)}$ or to $\Omega_{1-2}^{(2)}$.

Connectivity among the assigned points is then enforced in *Block B3*.

The obtained division of the domain is finally mapped back on the elements of the two bodies in *Block B4*.

For clarity, the algorithm for the interpolation of the sensitivities and for the assignment of the shared part of the domain (i.e. from *Block B1* to *Block B4*) is described in Sect. 3.

The sensitivities on Ω_1 and Ω_2 are filtered separately on the two domains in *Block A2* by the filter proposed in Sigmund (2001)

$$\widehat{\frac{\partial c}{\partial \rho_e}} = \frac{1}{\rho_e \sum_{f=1}^N \hat{H}_f} \sum_{f=1}^N \hat{H}_f \rho_f \frac{\partial c}{\partial \rho_f} \quad (3)$$

The weighting factor $\hat{H}_f$ is written as

$$\hat{H}_f = r_{\min} - \text{dist}(e, f), \quad \text{with } \{f \in N | \text{dist}(e, f) \leq r_{\min}\}, \quad e = 1, \dots, N \quad (4)$$

where the distance is computed between the centre of the elements e and f .

Once the elements describing Ω_1 and Ω_2 have been identified, the new pseudo-densities are computed in *Block A3*. The new density field is computed as (Sigmund 2001; Bendsøe and Sigmund 2004)

$$\rho_e^{\text{new}} = \begin{cases} \max(\rho_{\min}, \rho_e - m), & \text{if } \rho_e B_e^\eta \leq \max(\rho_{\min}, \rho_e - m) \\ \rho_e B_e^\eta, & \text{if } \max(\rho_{\min}, \rho_e - m) < \rho_e B_e^\eta < \min(1, \rho_e + m) \\ \min(1, \rho_e + m) & \text{if } \min(1, \rho_e + m) \leq \rho_e B_e^\eta \end{cases} \quad (5)$$

where m is a positive move limit and B_e represents the optimality condition

$$B_e = \frac{-\frac{\partial c}{\partial \rho_e}}{\lambda \frac{\partial V}{\partial \rho_e}}, \quad \frac{\partial V}{\partial \rho_e} = v_e \quad (6)$$

The Lagrange multiplier λ is computed by a Trust-Region Dogleg algorithm. The computation of the new pseudo-density field is performed on the whole domain $\Omega_1 \cup \Omega_2$. In this way, the total mass is not assigned a priori, but is divided among the two bodies according to the computed sensitivities. To guarantee that the total mass on the two physical domains Ω_1 and Ω_2 matches the target on the design volume fraction, the Lagrange multiplier λ is computed over $\Omega_1 \cup \Omega_2$ and not $\Omega_1 \cup \Omega_2$.

The pseudo-density of the non-active elements of Ω_1 and Ω_2 is set equal to ρ_{min} in Block B5.

3 Sensitivity interpolation and assignment of the shared part of the domain

In this section, the algorithm for the interpolation of the sensitivities and the assignment of the shared part of the domain is described. The aim of such algorithm is to divide the shared part of the domain when there is no correspondence between the mesh of the two bodies. To illustrate this situation, let us consider the example of Fig. 3. The example is based on the well-known problem of a cantilever loaded at the free end [see Bendsøe and Sigmund (2004) p. 49]. In the proposed example, two cantilevers are considered and are positioned to have an overlap in the respective design domains.

Figure 4a shows the two different meshes for the two bodies. The meshes are very different, on Ω_1 a four node bilinear mesh is considered, while on Ω_2 a three node linear mesh. The two meshes are coarse, with a mean size of 2 mm, for representation reasons.

Firstly, a grid of uniformly spaced points is set in the shared domain. This grid, denoted as interpolation grid in the following, is the same for the two bodies. The sensitivities will be interpolated at each point of the interpolation grid and used to assign the

Fig. 3 Two beams under bending numerical example definition

Fig. 4 Two beams under bending. **a** Mesh of the two domains. **b** Detail of the grid of interpolating points on the shared part of the design domains

domain to either body 1 or body 2. In Fig. 4b a detail of the shared part of the domain is reported, where the black dots represent the points of the interpolation grid.

The interpolation scheme is shown in Fig. 5. For each of the two body meshes, the centre of the elements (i.e. the centroids) are used as nodes to create an auxiliary linear triangular mesh. The centres of the elements are where the sensitivities have been computed, so are a natural choice as nodes of the auxiliary mesh used for the interpolation. The element containing each of the points of the interpolation grid is identified. The sensitivity at each point of the interpolating grid is then computed by applying the usual finite elements interpolation as

$$\left(\frac{\partial c}{\partial \rho}\right)_p = \sum_{i=1}^3 N_i \left(\frac{\partial c}{\partial \rho_e}\right)_i \quad (7)$$

where $\left(\frac{\partial c}{\partial \rho}\right)_p$ is the sensitivity at each point p of the grid and $\left(\frac{\partial c}{\partial \rho_e}\right)_i$ is the sensitivity at the i -th node of the element containing p , which has been computed at the centre of the corresponding element of the mesh of the body (Eq. 2). N_i are the three

Fig. 5 Construction of the triangular mesh for the interpolation of the grid points. Left: body 1. Right: body 2

standard shape functions of the linear triangle element. It has to be observed that, being the sensitivity computed on elements of different size, each value computed by Eq. 2 has to be divided by the area of the corresponding element.

Once the sensitivity has been interpolated on the shared part of the domain for the two bodies, the interpolated fields are compared at each point of the interpolation grid. Each point is then attributed to the first body ($\Omega_{1-2}^{(1)}$) if the sensitivity interpolated from the solution on Ω_1 is greater than the sensitivity interpolated from Ω_2 , or to the second body ($\Omega_{1-2}^{(2)}$) otherwise.

The connectivity of the two sets $\Omega_{1-2}^{(1)}$ and $\Omega_{1-2}^{(2)}$ is enforced as described in the following. The points assigned to the two sets $\Omega_{1-2}^{(1)}$ and $\Omega_{1-2}^{(2)}$ are clustered according to their connectivity to the points of the same set. If a point (or a group of points) is insulated with respect to the main part of its set, it is assigned to the other set. This analysis is performed simultaneously on the two sets, in this way neither set has the priority on the other.

Finally, the regions $\Omega_{1-2}^{(1)}$ and $\Omega_{1-2}^{(2)}$ have to be mapped back from the interpolation grid to the two meshes of the bodies. In other words, the elements of Ω_1 that belong also to $\overline{\Omega}_1$ and the elements of Ω_2 that belong also to $\overline{\Omega}_2$ are considered active, while the other elements are considered not active. In this operation, three situations can arise. An element can be entirely inside the active domain, entirely outside the active domain or partially inside and partially outside the active domain. In the latter case, the element has to be marked as partially active. To take into consideration these cases, the activity of each element is evaluated as shown in Fig. 6. Each point of the interpolation grid has been marked as active (red circle, value 1) or inactive (black circle, value 0). To compute the activation level of the elements of the body mesh, for each element a certain number of control points is used. As control points, a very simple and convenient choice are the Gauss points of the element (Liu et al. 2005; Li et al. 2018; Kang and Wang 2011; Wang et al. 2013), represented as squares in Fig. 6. There is no restriction on the number of Gauss points to be used for the interpolation. In the following examples, the same number of Gauss points used for the element integration is adopted to maintain the same formulation of the finite element solver. The activation value of each Gauss point is computed by

Fig. 6 Mapping of the activity field from the interpolation grid points to the centre of each element

considering a linear triangular grid on the interpolation points as done for the interpolation of the sensitivity field. Once the activation level has been interpolated at each Gauss point, the activation level of the entire element is computed by applying the quadrature rule on the values at the Gauss points and associated to the centre of the element itself as

$$a_e = \sum_{i=1}^{n_{GP}} w_i a_{GP,i} \quad (8)$$

where a_e and $a_{GP,i}$ are the activation level of the element and of each Gauss point respectively, n_{GP} is the number of Gauss points and w_i are the Gauss point weights. In Fig. 6, the activation level of each Gauss point is in black and the resulting activation level of the element is in blue.

In Fig. 6, the bilinear quadrilateral mesh of Ω_1 has been considered. The same approach can be applied to the linear triangles of the mesh of Ω_2 , or to any other mesh. For this example, the same number of Gauss points of the base elements has been considered. However, this is not required and any number of Gauss points can be used.

The activation level a_e of each element is then considered in the computation of its new pseudo-density by modifying Eq. 5 as

$$\rho_e^{new} = \begin{cases} \max(\rho_{min}, \rho_e - m), & \text{if } a_e \rho_e B_e^\eta \leq \max(\rho_{min}, \rho_e - m) \\ a_e \rho_e B_e^\eta, & \text{if } \max(\rho_{min}, \rho_e - m) < a_e \rho_e B_e^\eta < \min(1, \rho_e + m) \\ \min(1, \rho_e + m) & \text{if } \min(1, \rho_e + m) \leq a_e \rho_e B_e^\eta \end{cases} \quad (9)$$

The proposed interpolation approach has been presented by considering a two-dimensional example. The extension to the three-dimensional case is straightforward. A three-dimensional interpolation grid is created. The values of sensitivity are computed at each point of the grid by interpolating the sensitivities computed at the centres of the elements by considering an auxiliary mesh of linear tetrahedrons with vertices in the centres of the elements. The back mapping of the sensitivities is realized by considering the Gauss points of the elements and an auxiliary mesh of linear tetrahedrons with vertices in the interpolation points.

4 Two dimensional numerical example

In this section a simple two dimensional problem is reported. The problem is the one depicted in Fig. 3 and consists of two cantilevers with an end load sharing part of the domain. Each of the two cantilevers corresponds to the well-known problem of the cantilever with load end (Bendsøe and Sigmund 2004). The two cantilevers have the same geometry, material, load and boundary conditions. However, the two domains are meshed with very different meshes.

The proposed example is solved for different meshes combination. Four mesh types are considered, namely linear triangular (T3), quadratic triangular (T6),

bilinear quadrilateral (Q4) and quadratic quadrilateral (Q8). The properties of the finite element meshes are defined in Table 1. The mean element size is 0.5 mm for both triangular and quadrilateral elements, thus quadrilateral elements have approximately a double area with respect to the triangular elements. The volume fraction is set to 0.55 and the interpolation grid has a dimension of 0.5 mm. The obtained compliances and mass fractions of the two bodies are analysed and compared for the different mesh combinations.

The results of the concurrent topological optimization of the two cantilevers are reported in In Fig. 7. At first, the results of the original algorithm presented in Prevati et al. (2018) are shown for comparison. In this case, the two bodies are meshed with the same mesh of Q4 elements with corresponding positions of the nodes. Then, in Fig. 7b, c, the results for two mesh combinations obtained by using the algorithm proposed in this paper are shown. In particular, in Fig. 7b, the combination T3-Q8 is reported, while in Fig. 7c the combination T6-Q4. The subdivision of the common part of the design space and the two optimized structures are shown.

Referring to the two structures of Fig. 7b, c, the two optimized topological layouts are very similar, as one may expect as the two sub-problems are the same and the found topology is also similar to the theoretical solution of the problem of the cantilever with loaded end (Bendsøe and Sigmund 2004, p. 49). Also, the solution is very similar for the two different mesh combinations. When compared with the output of the original algorithm (Fig. 7a), it emerges that the results of the algorithm presented in this paper are very similar, showing that the interpolation process does not affect the final solution.

In Table 2, the resulting compliance and mass of the optimized cantilever are reported for all the mesh combinations. As reference, the compliance of the structures obtained by the original algorithm presented in Prevati et al. (2018) is also reported. In all the considered mesh combinations, the compliance of the two bodies is very close and the mean compliance is similar for the different cases. Also, for all mesh combinations, the compliance is similar to the compliance of the two cantilevers obtained by using the original algorithm. A slightly higher compliance is obtained for the quadratic elements, which can be related to the less stiff mathematical formulation of these elements. In fact, at the first step, when all the domains are equal and the density is uniform, the compliances for the different elements read 45.52 J for T3, 47.14 J for T6, 46.28 J for Q4 and 47.24 J for Q8. Referring to the

Table 1 Meshes used for the solution of the example of Fig. 3

Mesh combination	T3	T6	Q4	Q8
Mesh type	Triangular	Triangular	Quadrilateral	Quadrilateral
Number of nodes	3	6	4	8
Order	Linear	Quadratic	Bilinear	Quadratic
Number of Gauss points	1	3	4	9
Number of Gauss points for interpolation (Sect. 3)	3	7	4	9
Characteristic dimension [mm]	0.5	0.5	0.5	0.5

Fig. 7 Solution of the two dimensional example of Fig. 3 with two different mesh types. **a** Original algorithm presented in Previati et al. (2018), linear quadrilateral elements. **b** Linear triangular elements and quadratic quadrilateral elements. **c** Quadratic triangular elements and linear quadrilateral elements

masses, for all the mesh combinations a mass division very close to 50% has been obtained for the two bodies. Slightly more mass is attributed to the body with more compliant mesh, in order to compensate for the different stiffness of the meshes.

5 Concurrent topological optimization of a tool support swing arm

The described algorithm has been employed for the optimization of a tool support swing arm for a tube bending machine. This research activity has been completed in collaboration with BLM Group (BLM 2019). This example has been originally presented in Ballo et al. (2019, 2020). Due to the very high production rate of the machine, the swing arm is subjected to high accelerations and, as a consequence, an inertial torque arises. The optimization of the system is thus important to reduce energy consumption and increase the production rate.

Figure 8 shows a 3D model of the tool support swing arm, where all its main parts are highlighted. The figure shows only half of the model due to the symmetry in the geometry of the system. The two parts that are optimised are the support arm and the sledge, shown respectively in blue and red colour in Fig. 8. The support arm rotates around the vertical rotation axis of Fig. 8, while the sledge is allowed to slide on a guide rail on the support arm in the y direction of Fig. 8. The sledge sustains the tube clamp die and is set in motion through an electric motor and a ball screw drive.

In Fig. 9, the finite element model of the tool support swing arm is depicted. In the model, symmetry has been exploited by considering only half of the system. The tool load due to the contact between the clamp die and the tube surface is applied to the sledge by a multi-point constraint. A contact interaction is imposed between the sledge and the support arm in correspondence of the guide rail. It is worth noting that, by including this contact condition, a non linear finite element analysis has to be run for each optimization step. The ball screw drive is modelled by a rigid connector between the sledge and the support arm as depicted in Fig. 9.

The system is subject to an angular acceleration around the rotation axis. Also, three different loads are considered. The loads are applied one at the time and for different positions of the sledge along the rail. Forces F_2 and F_3 act only in the y direction, while force F_1 has x and y components. Due to the angular acceleration and force F_1 , the system undergoes a non symmetric deformation. To maintain the geometrical symmetry of the components, a symmetry constraint is applied in the optimization algorithm. The symmetry constraint is enforced as explained in Kosaka and Swan (1999). The components are manufactured by additive manufacturing, thus no manufacturability constraint is considered.

The optimization domains of the support arm and of the sledge are also reported in Fig. 9. In the figure, a relatively small region is highlighted in green. This region is quite critical for the stiffness of both components and cannot be used by both as they would interfere when the sledge is in the most inner position. This region is considered as shared domain that can be assigned to either component during the optimization process. With respect to the algorithm of Fig. 2, in the optimization of the tool support swing arm, not only the connectness of the two design spaces is enforced,

Table 2 Results of the two cantilevers example (Fig. 3). Mesh properties in Table 1

Mesh combination	Orig. Alg.				
	Q4-Q4	Prevati et al. (2018)	Q4-Q8	T3-Q4	T6-Q4
Compliance body 1 [J]	12.14		12.18	12.00	12.36
Compliance body 2 [J]	12.14		12.43	12.13	12.18
Mean compliance [J]	12.14		12.30	12.06	12.27
% Mass body 1	50.00		49.80	49.96	50.18
% Mass body 2	50.00		50.20	50.04	49.82
Mesh combination	T3-Q8		T6-Q8	T3-T6	
Compliance body 1 [J]	12.05		12.40	12.05	
Compliance body 2 [J]	12.39		12.44	12.34	
Mean compliance [J]	12.22		12.42	12.20	
% Mass body 1	49.77		50.00	49.84	
% Mass body 2	50.23		50.00	50.16

Fig. 8 Half view of the 3D model of the tool support swing arm of a tube bending machine

Fig. 9 Tool support swing arm model (BLM 2019)

but an additional condition that the shape of the domains must allow the sliding of the sledge is considered. The two components to be optimized are meshed by four-nodes tetrahedrons with mean size of 1.5 mm. A total of about 1.65×10^6 elements is present in the model. The total compliance of the system is minimised by enforcing an upper bound on the overall volume fraction.

The optimization problem is solved by using AbaqusTM2016 for the solution of the non linear finite element model and a handwritten code to read the results, implement the optimization algorithm and write the new input file. The handwritten

code is realized by using Python and Matlab™. At the beginning of the analysis, the same pseudodensity has been assigned to all the elements.

The results of the optimization process, with an overall target volume fraction of 0.55, are shown in Fig. 10. In Fig. 10a an overall view of the optimized system is reported. In Fig. 10b, c, details of the two optimized components can be seen. The shared domain, marked with two arrows in Fig. 10b, c, has been divided in a quite complex way, allowing an internal reinforcement for the sledge and a lower reinforcement for the support arm. Finally, Fig. 11 highlights a detail of the division of the shared domain between the two optimised structures.

6 Conclusion

In the paper, a novel algorithm has been presented for the concurrent topological optimization of two bodies sharing part of the design domain. The new algorithm allows for the utilization of any arbitrary mesh on the bodies. Also, the algorithm has been used with a commercial finite element software by considering a SIMP approach and a symmetry constraint in the solution, proving the applicability of the method with an existing optimization algorithm. In this way, the algorithm can be

Fig. 10 Optimization results—surfaces with pseudodensity greater than 0.3. **a** Complete system. **b** Detail of the optimized sledge. **c** Detail of the optimized support arm. The two arrows indicate the shared region

Fig. 11 Optimization results—detail of the division of the common part of the domain. Elements with pseudodensity less than 0.3 have been removed

applied to real world optimization problems, enabling the designer to select topology and size of the mesh that best adapt to the shape of each single body.

The new algorithm has been tested on a simple two dimensional problem and then applied to the optimization of the arm of a tube bending machine designed in collaboration with BLM Group. The application has shown the ability of the algorithm to solve real problems and find non trivial efficient solutions for the assignment of the shared domain. Further developments of the method that consider the possibility to include contact interactions in the shared part of the domain will be investigated.

Funding Open access funding provided by Politecnico di Milano within the CRUI-CARE Agreement.

Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit <http://creativecommons.org/licenses/by/4.0/>.

References

- BLM Group (2019). <http://www.blmgroup.com>. Accessed 24 Jan 2019
- Ballo F, Gobbi M, Prevati G (2019) Concurrent topological optimisation: optimisation of two components sharing the design space. In: Rodrigues HC, Herskovits J, Mota Soares CM, Araújo AL, Guedes JM, Folgado JO, Moleiro F, Madeira JFA (eds) EngOpt 2018 proceedings of the 6th international conference on engineering optimization, Springer International Publishing, Cham, pp 725–738. https://doi.org/10.1007/978-3-319-97773-7_64
- Ballo F, Gobbi M, Prevati G (2020) Concurrent topological optimization of a multi-component arm for a tube bending machine. In: Le Thi H, Le H, Pham Dinh T (eds) Optimization of complex systems:

- theory, models, algorithms and applications. WCGO 2019. Springer, Cham, pp 68–77. https://doi.org/10.1007/978-3-030-21803-4_7
- Bendsøe MP, Sigmund O (2004) Topology optimization. In: Theory, methods, and applications, 2nd edn, Springer, Berlin. <https://doi.org/10.1007/978-3-662-05086-6>
- Eschenauer HA, Olhoff N (2001) Topology optimization of continuum structures: a review. *Appl Mech Rev* 54(4):331. <https://doi.org/10.1115/1.1388075>
- Guo X, Cheng GD (2010) Recent development in structural design and optimization. *Acta Mech Sin* 26(6):807–823. <https://doi.org/10.1007/s10409-010-0395-7> [arXiv:1010.1724](https://arxiv.org/abs/1010.1724)
- Kang Z, Wang Y (2011) Structural topology optimization based on non-local Shepard interpolation of density field. *Comput Methods Appl Mech Eng* 200(49–52):3515–3525. <https://doi.org/10.1016/j.cma.2011.09.001>
- Kosaka I, Swan CC (1999) A symmetry reduction method for continuum structural topology optimization. *Comput Struct* 70(1):47–61. [https://doi.org/10.1016/S0045-7949\(98\)00158-8](https://doi.org/10.1016/S0045-7949(98)00158-8)
- Lawry M, Maute K (2018) Level set shape and topology optimization of finite strain bilateral contact problems. *Int J Numer Methods Eng* 113(8):1340–1369. <https://doi.org/10.1002/nme.5582>. [arXiv:1701.06092](https://arxiv.org/abs/1701.06092)
- Li Y, Liu GR, Yue JH (2018) A novel node-based smoothed radial point interpolation method for 2D and 3D solid mechanics problems. *Comput Struct* 196:157–172. <https://doi.org/10.1016/j.compstruc.2017.11.010>
- Liu GR, Zhang GY, Gu YT, Wang YY (2005) A meshfree radial point interpolation method (RPIM) for three-dimensional solids. *Comput Mech* 36(6):421–430. <https://doi.org/10.1007/s00466-005-0657-6>
- Prevati G, Ballo F, Gobbi M (2018) Concurrent topological optimization of two bodies sharing design space: problem formulation and numerical solution. *Struct Multidiscip Optim*. <https://doi.org/10.1007/s00158-018-2097-x>
- Rozvany GIN (2009) A critical review of established methods of structural topology optimization. *Struct Multidiscip Optim* 37(3):217–237. <https://doi.org/10.1007/s00158-007-0217-0>
- Sigmund O (2001) A 99 line topology optimization code written in matlab. *Struct Multidiscip Optim* 21(2):120–127. <https://doi.org/10.1007/s001580050176>
- Strömberg N (2018) Topology optimization of orthotropic elastic design domains with mortar contact conditions. In: Schumacher A, Vietor T, Fiebig S, Bletzinger K-U MK (ed) *Advances in structural and multidisciplinary optimization: proceedings of the 12th world congress of structural and multidisciplinary optimization*, Springer, Braunschweig, Germany, pp 1427–1438. <https://doi.org/10.1007/978-3-319-67988-4>
- Tavakoli R, Mohseni SM (2014) Alternating active-phase algorithm for multimaterial topology optimization problems: a 115-line MATLAB implementation. *Struct Multidiscip Optim* 49(4):621–642. <https://doi.org/10.1007/s00158-013-0999-1>
- Wang Y, Kang Z, He Q (2013) An adaptive refinement approach for topology optimization based on separated density field description. *Comput Struct* 117:10–22. <https://doi.org/10.1016/j.compstruc.2012.11.004>
- Zhang W, Yuan J, Zhang J, Guo X (2016a) A new topology optimization approach based on moving morphable components (MMC) and the ersatz material model. *Struct Multidiscip Optim* 53(6):1243–1260. <https://doi.org/10.1007/s00158-015-1372-3>
- Zhang W, Zhu J, Gao T (2016b) *Topology optimization in engineering structure design*. Elsevier, Oxford
- Zhu JH, Zhang WH, Xia L (2016) Topology optimization in aircraft and aerospace structures design. *Arch Comput Methods Eng* 23(4):595–622. <https://doi.org/10.1007/s11831-015-9151-2>

Publisher's Note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.