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Article

Sensitivity Analysis and Filtering of Machinable Parts Using Density-Based Topology Optimization

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Abstract: Topology optimization has become a very popular design tool when an optimum design with a concrete objective and subjected to several restrictions is searched. However, the resultant shape of the optimization process isn't usually easy to manufacture with a typical manufacturing method as machining, being necessary an interpretation and validation of the result; besides, some parameters can be chosen, and the final shape is directly dependent on them. In this paper, a sensitivity analysis of the main parameters involved in 3D topology optimization, the penalization and the filter radius, is carried out for the concrete case of the density-based method, analyzing how manufacturable the results are for 3 axis machining. Also, a new filter and method is proposed for obtaining highly manufacturable and low interpretable parts. The main aim of the study is to help designers to choose the more appropriate parameters and showing what they must have into account for obtaining the desired optimized shapes. Note that the density-based topology optimization method has been chosen due to its popularity in commercial packages, so the conclusions may have direct applications in designers' work. Finally, the results of the study are verified through different cases for checking the validity of the conclusions.

Keywords: topology optimization; manufacture filter; penalization; filter radius; machining

1. Introduction

Topology optimization (TO) is a design tool for obtaining optimal shapes with known conditions, based on the optimization problem definition; In other words, it aims to achieve the best possible result for a given function while applying specific constraints. Topology optimization has been increasingly used during the last years, as well as the research on the topic and the development of commercial software for obtaining topology parts, as mentioned by Shiye and Jiejiang [1]. Nevertheless, the foundations of TO theory were established over a century ago [2], with major applications to the actual theory with the guidelines developed by Dorn [3]. Years later, the computerization of topology optimization arrived [4] and later further applications in the digital era [5].

In this epigraph, a brief explanation of the density-based TO is presented in order to show the used methods, solving techniques and filters; as well as the application of the studied parameters. The values of these and MatLab® codes are provided for replicating the results of the study, which can be applied to the main topology optimization codes discussed in the literature [6,7]. This chapter is divided in two sections: one related to the used topology optimization method, and other related to the filters oriented towards subtractive manufacturing.

1.1. Density-Based Topology Optimization

The density-based approach starts from a design domain Ω that is discretized using nodes and elements. This design domain is the maximum space (line, surface or volume) where the final design

could exist, and a density value is assigned to each element of the discretized design domain, hence it is parametrized by this density. The density value is not binary, so it can vary from 0 (void) to 1 (solid), and this value is applied by using an interpolation function [8,9]. The use of only black and white elements easily leads to an ill-posed result [10]. Many solutions have been proposed [11–13], but none are totally satisfactory. An interesting proposal was to include not only binary values for the density of each element [9,14], those values being chosen automatically following a power rule (Zhou & Rozvany, 1991): the interpolation scheme, that will be explained later in this manuscript.

The used density-based approach is extensively described in several studies [6,16], and follows the simplified flowchart showed in **Error! Reference source not found.**, where every step is numbered for future references.

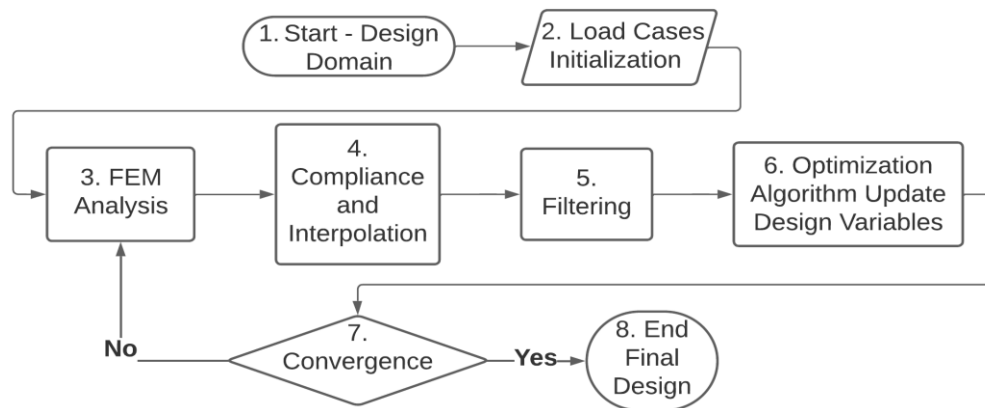


Figure 1. Simplified flowchart of the density-based topology optimization process.

In this study, special attention is given to levels 4 and 5 of the flowchart in **Error! Reference source not found.**, as these processes involve the parameters under study. Specifically, for the interpolation in level 4, Solid Isotropic Material with Penalization method (SIMP) is utilized; additionally, a density filter is employed as the filtering technique in level 5.

1.2. Interpolation Scheme, the SIMP

The SIMP is based on the relation of the stress state of every element with its relative density. The formulation is included in Equation (1):

$$E_i = E(x_i) = E_{\min} + x_i^p (E_0 - E_{\min}) \quad (1)$$

Being E_i the relative Young Modulus of each element (calculated using FEM and representing the stress state of the element; note that it is also dependant on its degrees of freedom), E_{\min} as the almost zero elastic modulus of void material for avoiding mathematical incompatibility issues, E_0 as the elastic modulus of the original material, x_i as the relative density of the element i , and p as the penalization factor. This penalization factor is one of the parameters to use in the sensitivity analysis of this manuscript. According to the Equation (1), it is evident that a higher p will result in a more binary outcome (Jiang, 2017), corresponding to the graph of a power rule function presented in **Error! Reference source not found.**

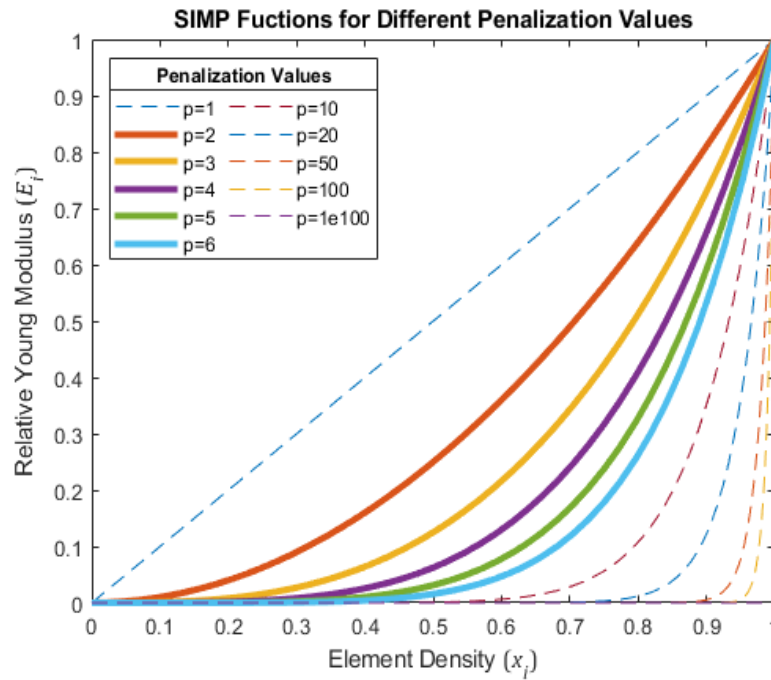


Figure 2. Graphical representation of the SIMP interpolation power rule for penalization values between 1 and 100.

SIMP represents a robust, highly configurable, and computationally efficient scheme. It does not require homogenization and is adaptive to various design conditions and constraints. Due to these advantages, SIMP is chosen as the initial interpolation scheme for the research conducted in this work. However, using SIMP comes with challenges such as dealing with intermediate densities, mesh dependency, and potential non-convergence [17].

1.3. Filtering, the Density Filter

The density-based method employs finite element analysis (step 3 in **Error! Reference source not found.**) and the SIMP method to assign each element a relative density value ranging from 0 to 1. The classical approach is finding a black-and-white layout (zeros and ones) that minimizes the compliance of the structure, subjected to a maximum volume fraction in relation to the original volume of the design domain [6].

The use of purely black and white elements often leads to an ill-posed result [10]. Many solutions have been proposed [11–13], but none have proven entirely satisfactory. An interesting proposal was to include non-binary density values for each element [9,14], automatically determined through a power rule [15], the SIMP.

Nonetheless, even with SIMP, mesh dependency, checkerboard, ill-posed or local minima results can still arise [18]. To address these issues the density filter [19] is employed in this study. A basic formulation of this filter is presented in Equation (2):

$$\tilde{x}_i = \frac{\sum_{j \in N_i} H_{ij} v_j x_j}{\sum_{j \in N_i} H_{ij} v_j} \quad (2)$$

where \tilde{x}_i is the filtered density of an element, N_i represents the neighborhood of a specific element, v_j is the volume of the element j , and H_{ij} is a weight factor. The neighbourhood of an element is defined as all the elements whose centre is within a certain distance R from the centre of the element in analysis. This can be defined as Equation (3):

$$N_i = \{j: \text{dist}(i, j) \leq R\} \quad (3)$$

And the weight factor H_{ij} can be defined as Equation (4):

$$H_{ij} = R - \text{dist}(i, j) \quad (4)$$

Being j always inside the neighbourhood N_i and representing the geometric center of each element. The constant R is defined as the filter radius, and is the second parameter used in the sensitivity analysis throughout this paper. Now, \tilde{x}_i from Equation (2), known as the filtered density of the element, is substituted in the SIMP (Equation (1)) as Equation (5):

$$E_i = E(\tilde{x}_i) = E_{\min} + \tilde{x}_i^p (E_0 - E_{\min}) \quad (5)$$

1.4. Study Description and Novelty

As density-based TO is the most popular method in the commercial packages, such as Altair Hyperworks suite [20], Ansys suite [21], Dassault Systèmes suite [22] or PTC suite [23]; this manuscript aims to analyze the two main parameters involved in the optimization process: penalty and filter radius. The objective of this study is to determine the best parameters for obtaining manufacturable results with minimal interpretation for 3 axis machining. This research aims to assist designers in making informed decisions and save time in preprocessing and simulation phases.

A wide range of values is considered for the mentioned parameters, and calculation times, discreteness and machinability of the results are presented. It is important to note that other studies [1,24,25] make remarkable contributions in this area, but none of them are specifically oriented to a concrete manufacturing method and are often limited to a short set of values for the studied parameters, they don't take into account the mesh dependency, or study only the 2D case. Other works [26,27] concentrate on achieving these manufacturable parts by adding filters and additional constraints to the TO process, but a more direct application is searched in the present paper.

The given text describes the development of a novel analysis tool that can assess the machinability of a part for a 3-axis machine. This tool is later applied as a filter to automatically modify the non extra-constrained part, making it more machinable and less interpretable. A MatLab® will be provided for the use of the scientific and designers community, and for the replication of the research results.

The structure of the paper is organized as follows:

- The initial section discusses the chosen parameters values criterion is, along with the load case and material parameters of the main study case for the sensitivity analysis.
- Following that, the parameters used for analyzing the results are exposed and explained, with particular emphasis in the measure of the discreteness and the evaluation of the machinability,
- Subsequently, results of the 92 simulations are exposed and analyzed, being able to obtain the conclusions related to the aim of the study.
- The machining filter is presented and verified through different verification tests.
- Finally, the conclusions drawn from the study and potential avenues for future research are presented.

2. The Case Study

2.1. Parameters' Values Selection Criterion

The text highlights that the filter radius (R) and penalization (p) are the two main parameters being analyzed. These parameters should not be chosen arbitrarily but require careful consideration to achieve a satisfactory outcome.

For penalization p , previous studies [1,24] indicate that a minimum value of 3 should be used for isotropic materials with a Poisson Ratio $\nu \sim 1/3$, within the defined topology optimization method. The relative density of each element, as shown in **Error! Reference source not found.**, follows a power rule, and it is observed that higher penalization values lead to sharper trends towards either 0 or 1 density. However, as the penalization increases, the differences in the results become smaller [25]. To analyze the parameter's impact, values ranging from three (3) to six (6), inclusive, with increments of one (1) are chosen based on previous studies [1,24] indicate. By using these values, it becomes possible

to observe the transitional changes in the resultant shapes when larger penalization values are employed.

For filter radius R , the values have been taken from different scopes, and they are summarized in Table 1. These sources include:

- Values used in other mentioned studies (OS).
- Values derived from the Euclidean distance between the element under analysis and the neighboring layer of elements (ED).
- Large estimated values intended to observe the behavior of the method with extreme values (LV).

Table 1. Chosen study filter radius, the reason of the choice of these radii and the percentage of the elements in maximum direction that represents the radii. OS=Other Studies. ED=Euclidean Distance. LV=Large Values.

R	1.5	1.8	2	2.45	2.5	2.9	3	3.4	3.47	3.5	3.8	4
Reason	OS	ED	ED	ED	ED	ED	OS	ED	ED	ED	ED	ED
R	4.2	4.4	4.5	4.7	5	5.2	6	8	10	15	20	
Reason	ED	ED	ED	ED	ED	ED	OS	LV	LV	LV	LV	

The values of the filter radii in Table 1 have been rounded to reduce computation time. This approach was chosen because the filter radius is a highly recurrent value throughout the process. Rounding the number to a value that does not overlap with the next filter radius value does not impact the result.

2.2. Case Study Description and Material

One aim of the study is to analyze parameters in a 3D structure, so the initial design space needs reliable dimensions in the x , y , and z directions. The typical load case involves minimizing compliance with a volume fraction constraint in a cantilever beam with a vertically applied distributed force. In this study, a point force (tip load) is used to emphasize the 3D nature of the case.

Considering that the element size is 1 in all directions, the initial design space consists of 80 elements in the x direction, 30 elements in the y direction, and 20 elements in the z direction. This choice helps generate fewer flat shapes in the optimization result. **Error! Reference source not found.** illustrates the case study with the applied load and the number of elements.

For this study, a volume fraction of 0.2 is applied, and a material with a Poisson Ratio (ν) of 0.3 is used. The value of E_{min} in Equation 1 is set to 10^{-9} which is close enough to zero to avoid affecting calculations and prevent mathematical singularities; and E_0 is set to 1, which is a typical used value in mentioned previous studies for researching and reducing calculation time. These properties are chosen for comparing the results to other studies [1,24] and having a direct application in the know and accesible MatLab® codes [6,7].

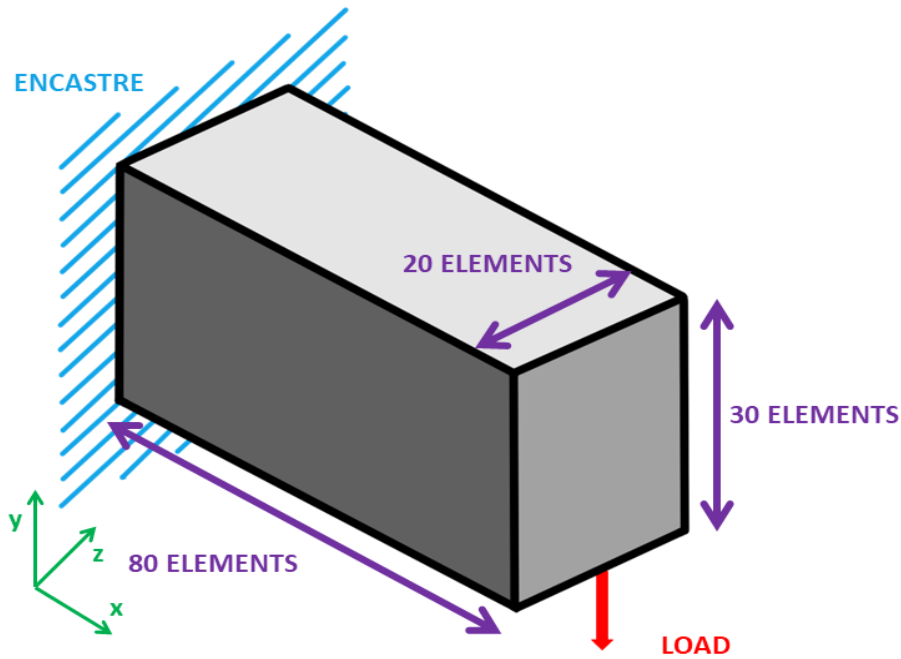


Figure 3. Study Load Case and Design Domain.

The convergence is achieved when the maximum change in the density of any element between two consecutive iterations is less than 0.01, which is the recommended value based on the experience of previous studies [6].

2.3. Parameters for the Results Analysis

The sensitivity analysis aims to evaluate the machinability by exploring different filter radii and penalties to assist designers in their projects. The parameters extracted from the simulation results are intended to objectively demonstrate this characteristic. These parameters include the number of iterations, time per iteration, measure of non-discreteness, and machinability.

The number of iterations and time per iteration are directly obtained from MatLab® and will be presented in the subsequent section. However, further explanations are needed for the measure of non-discreteness and machinability, which will be provided below.

2.3.1. Measure of Non-Discreteness

Discreteness refers to a global characteristic of the final design, indicating how binary the result is with distinct void ($\tilde{x}_i=0$) and solid ($\tilde{x}_i=1$) elements. This parameter is important because the representation of the result includes only elements with density \tilde{x}_i above a threshold (in this case, 0.5). Consequently, there are elements that do not appear in the final design but still contribute to compliance, and their significance increases as the discreteness of the result grows. Since the designer needs to interpret the shape displayed in the topology optimization (TO) result, and one objective of the study is to obtain easily interpretable components, it is essential to control and minimize the discreteness.

To analyze the characteristic of discreteness, Sigmund introduced a tool called the measure of non-discreteness (M_{nd}), which is widely utilized in the investigation of this matter [28]. The measure of non-discreteness is represented by Equation 6:

$$M_{nd} = \frac{\sum_{e=1}^N 4\tilde{x}_e(1-\tilde{x}_e)}{N} \times 100\% \quad (6)$$

In the provided context, N refers to the ID of each element in the design domain. The parameter M_{nd} represents the measure of non-discreteness. It should be noted that the lower the value of M_{nd} , the more binary the result of the analysis is, indicating a higher level of discreteness.

2.3.2. Machinability

Here, the machinability of the optimized shape is analyzed by assessing the accessibility of a hypothetical tool to each frontier element. A frontier element refers to the last element with a density higher than the chosen threshold. This analysis involves an element-by-element evaluation.

To initiate the analysis, the set of elements exceeding the selected density threshold is divided into distinct groups:

- The external layer (*EL*): Comprising all the elements located on the boundaries of the design domain. These elements are always accessible because they do not have any solid interface with the exterior.
- The core layers (*CL*): Representing the internal solid region, consisting of all the elements forming the shape of the part excluding the frontier elements.
- The frontier layer (*FL*): Consisting of the elements above the density threshold and located between the solid and void phases.

The frontier layer (*FL*) can be determined by considering that it includes all the elements exceeding the threshold, but not belonging to the *EL* or *CL* groups, as stated in Equation (7):

$$\sum_{n=1}^n N = \sum_{e=1}^e EL + \sum_{c=1}^c CL + \sum_{f=1}^f FL \quad (7)$$

The *FL* is divided into three distinct groups in the subsequent step. These groups are based on the accessibility of the elements for a tool and their machinability. The groups are as follows:

1. Elements directly accessible for a tool: These are the elements within the *FL* that can be readily machined using a tool.
2. Elements not directly accessible but machinable by machining a neighboring element: These elements are not directly accessible to a tool but can still be manufactured through the machining of a neighboring element.
3. Non-machinable elements: This group consists of elements within the *FL* that are not machinable.

This division allows for a more detailed understanding of the machinability characteristics of the elements within the *FL*, providing insights into their accessibility for machining operations. For the *FL* elements in groups 1 and 2, the process for identifying them is as follows:

For an element to be considered directly accessible for a tool, it must satisfy a specific condition: at its free faces, there must be a row of elements with densities below the selected threshold that extends to the boundaries of the design domain. To illustrate this process, let's consider the simplified 2D example in **Error! Reference source not found.**, where the identification process is depicted. The blue elements represent those with densities exceeding the threshold, while the green element $e_{(x,y)}$ is the element currently being analyzed.

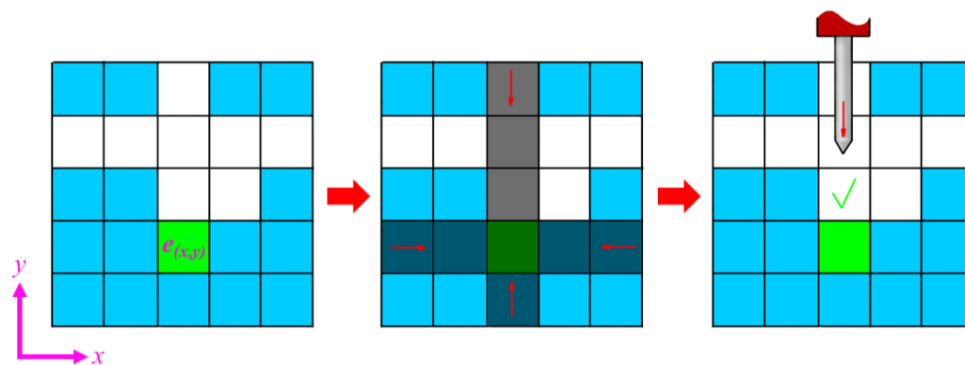


Figure 4. Element directly accessible for a tool detection process. In blue, solid phase; in green, the element under analysis in coordinates (x,y) .

To determine if an element satisfies the condition of being directly accessible for a tool in the y axis, the row and column leading to that element are analysed from the outside towards the inside in both directions (shaded elements in the central draw of **Error! Reference source not found.**). If at any point during the analysis an element is encountered with a density exceeding the threshold, the condition isn't fulfilled for that specific element.

To be considered machinable through its neighbor, an element must meet a specific condition: a free row must extend to its free face. To determine this, the neighboring rows are analyzed from the outside towards the inside until reaching the free face of the element. If, during this analysis, none of the studied lines contains an element with a density above the threshold, then the condition is fulfilled, indicating that the element can be machined through its neighboring elements. **Error! Reference source not found.** illustrates this process in a similar manner to the explained process in **Error! Reference source not found.** but in the x axis direction.

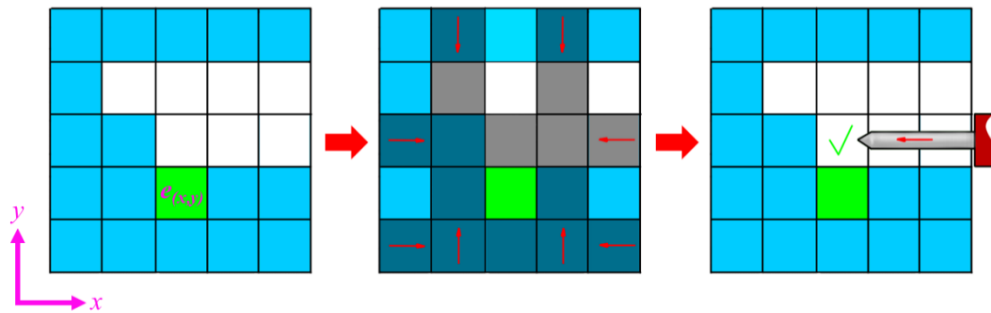


Figure 5. Element accessible for a tool through a neighbor detection process. In blue, solid phase; in green, the element under analysis in coordinates (x,y) .

The output parameter is a percentage of the machinable elements with respect to all the frontier elements. The MATLAB ® code of this process and the rest of the machining filter (that is explained in subsequent epigraphs) are shown in Appendix A.

1. Sensitivity Analysis

The results of the TOs are presented in Tables 1 to 4, which can be found in the Appendices B.1, B.2, B.3 and B.4. Note that the relative density x_i of each element is represented with a greyscale factor in Tables 2 to 4; where a darker element represents a value closer to 1. Only elements with a relative density $x_i > 0.5$ are shown. Finally, only full parts are considered, meaning that partial shapes with poor density concentration areas are discarded as satisfactory outcomes.

- Numerical results for the analysis are shown in Table 1, being:
- Sim. ID, the identifier of the specific TO case.
- Penalization, p ; and Filter Radius R .
- Iterations, the number of iterations required to achieve the convergence condition. Note that a maximum of 5000 iterations was set, so if the TO reaches to this value, convergence is not reached.
- It. Time, the average time the TO took in every iteration.
- Objective, the compliance of the structure, calculated as in Equation (8):

$$c(x) = U^T K U = \sum_{e=1}^N E_e(x_e) u_e^T k_0 u_e \quad (8)$$

where K , U and F are the global stiffness matrix and the global displacement and forces vectors, respectively; N is the total number of elements, u_e is the e element displacement vector, k_0 is the element stiffness matrix and x_e is the e element density. It is calculated in the step 3 of **Error! Reference source not found.**, the FEM analysis. It is used as a reference value for comparing different cases.

- Non-Discreteness, the described parameter in subsection 2.3.1.
- Machinability, the described parameter in subsection 2.3.2.

Next, in **Error! Reference source not found.** and **Error! Reference source not found.**, some remarkable results are displayed along with a corrected graph of the discrete results data. The curve fitting strategy employed is a two-term exponential curve for Figure 6a, 6b and 7a; and a three-term polynomial curve for Figure 7a.

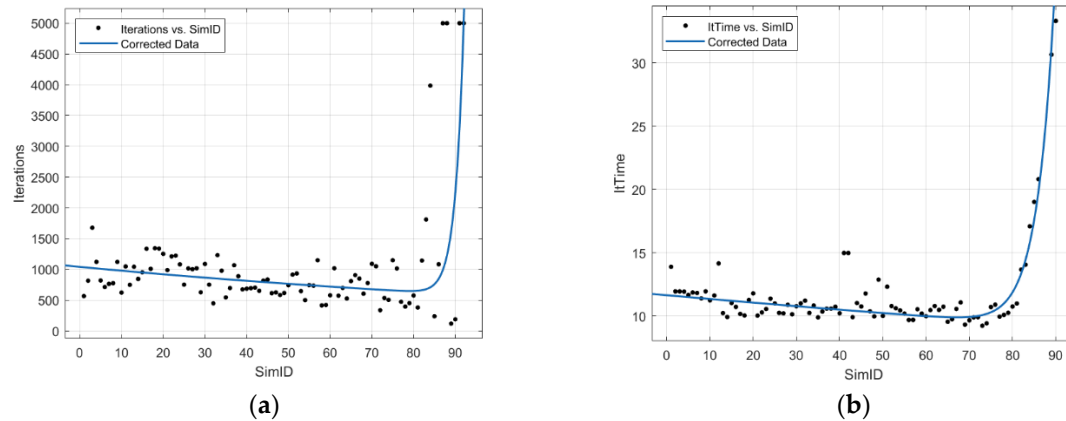


Figure 6. Comparison of the iterations and time per iteration in every TO. **(a)** Sim. ID vs Iterations graph: dots are the discrete data and blue is the corrected data. **(b)** Sim. ID vs Mean Iteration Time: dots are the discrete data and blue is the corrected data.

As the filter radius increases, it is expected that higher filter radius values lead to simpler shapes, as the density filter in Equation (2) affects a greater number of elements. However, when the value of the radius becomes very large, the results start to exhibit discontinuities, and the computational cost increases.

Nevertheless, the corrected data depicted in **Error! Reference source not found.** reveals an unexpected trend. It shows that as the filter radius grows, the topology optimization (TO) process requires fewer iterations and less time per iteration for computation. This finding contradicts the initial expectation, as a larger filter radius entails analyzing more elements in each iteration.

Error! Reference source not found. provides an analysis of the manufacturability of the TOs. In Figure 7a, it is evident that higher penalization values lead to slightly lower M_{nd} . This is an expected outcome since higher penalization corresponds to a more aggressive interpolation of the SIMP function defined by Equation (1) and **Error! Reference source not found.**.

Regarding non-discreteness, the results exhibit regular and predictable behavior for filter radii ranging from 4.25% to 6.5% of the EMD (Sim. ID 19-72), and the values increase with the corresponding radii. Similarly, machinability also increases with the filter radius since, as previously mentioned, a higher radius corresponds to simpler shapes that are more easily machinable.

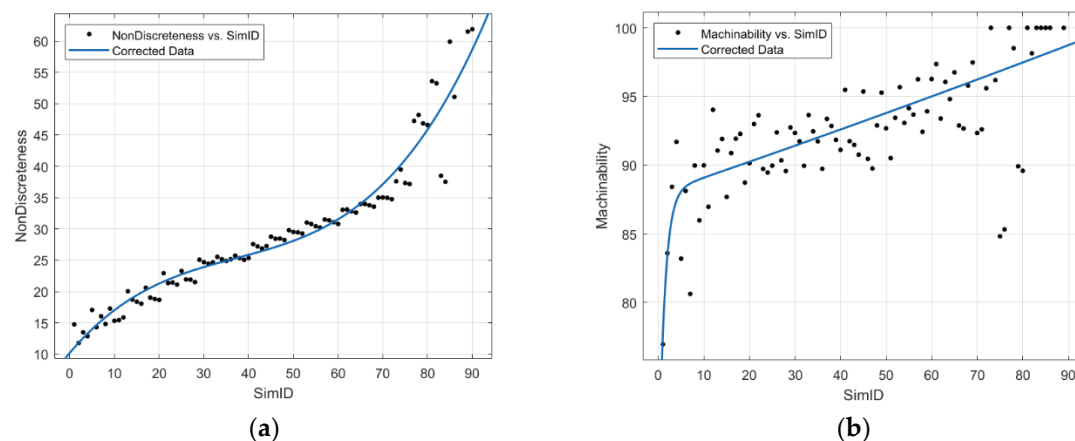


Figure 7. Comparison of the measure of non-discreteness and machinability in every TO. (a) Non-Discreteness vs Sim. ID: dots are the discrete data and blue is the corrected data. (b) Machinability vs Sim. ID: dots are the discrete data and blue is the corrected data.

In Figure 7, it is observed that results become more inconsistent and exhibit greater dispersion and poor predictability when large filter radii are used. Notably, in the default density-based TO approach, it is intriguing to note that achieving the best values for discreteness and machinability simultaneously is not possible, as depicted in Figure 8. The corrected data in Figure 8 is represented using a two-term exponential curve.

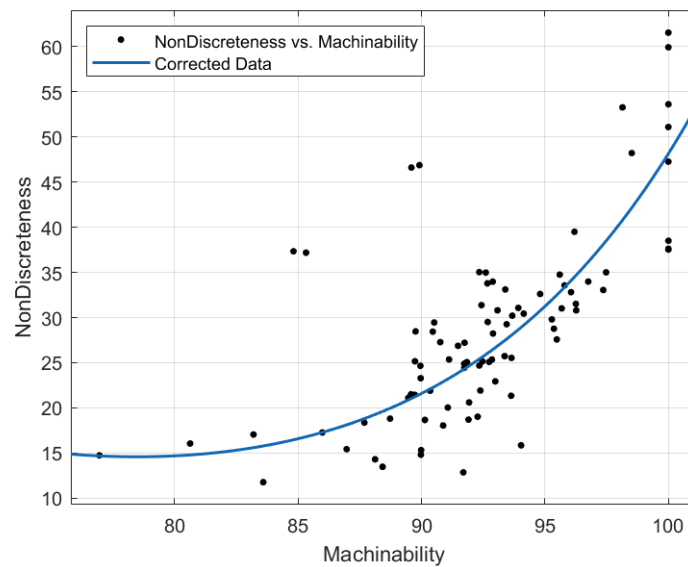


Figure 8. Comparison of Machinability and Measure of Non-Discreteness. Dots are the discrete data, and blue is the corrected data.

Machinability filter in combination with a Heaviside step filter [16,29–31] may result in an improvement of the M_{nd} -Machinability combination, as will be exposed in the next epigraph.

Lower filter radius and lower penalization values lead to better compliance outcomes. Additionally, results show lower dispersion with the smallest radii. These findings are expected because more complex shapes result in higher moment of inertia, leading to increased stiffness. This can clearly be seen in Figure 9, where the corrected data is represented using a two-term exponential line.

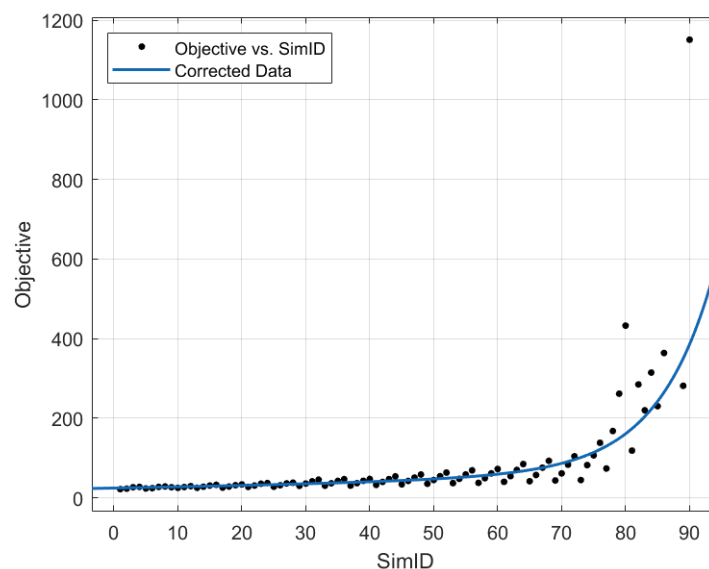


Figure 9. Comparison of Sim. ID and objective function results. Dots are the discrete data and blue is the corrected data.

2. Additions to Filtering

The filter is positioned inside the optimization algorithm, that in this case is the Optimality Criteria (OC) method described by Liu and Tovar in their work [6], using the parameters described in previous studies [32–35]. The flowchart depicted in **Error! Reference source not found.** illustrates the position of the filters (machining filter and Heaviside step filter) using the numbering notation from **Error! Reference source not found.**.

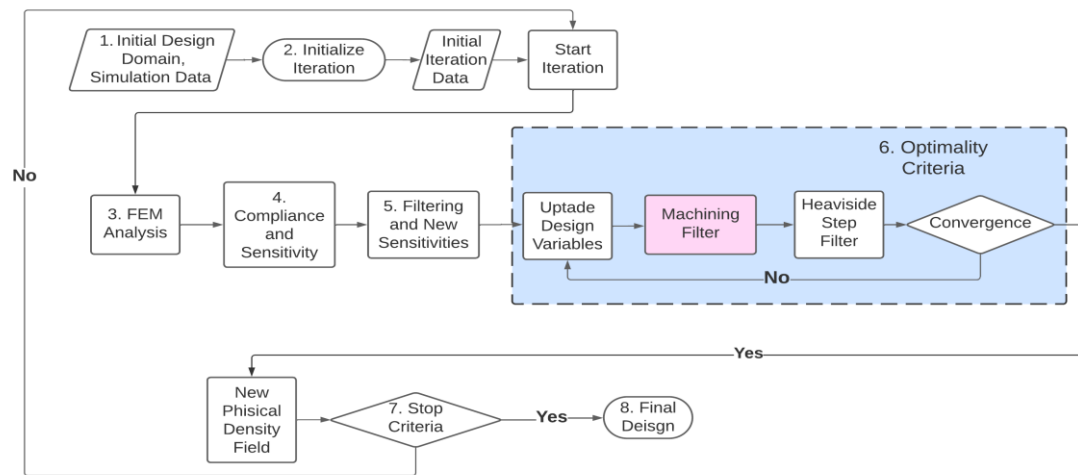


Figure 10. Flowchart of the TO method with the machining filter highlighted in pink for clarity in later explanations.

Three additions are done to the code: the machining filter, the Heaviside step filter, and the ellipsoid-shaped density filter. These implementations are next explained and positioned inside the OC routine.

2.4. Machining Filter

The new machining filter developed for this study utilizes a method described in section 2.3.2 to identify elements. Non-machinable elements are identified, and an individual machinability radius, R_{mach} , is assigned to each of them. The determination of R_{mach} involves analyzing the neighboring elements of the element under analysis, using the filter radii, R , described in Equation (4) and Table 1. This process is done individually for each cartesian axes, so the designer can obtain machinable parts in a concrete direction,

The analysis begins with the smallest practical radius of 1.5. If a neighboring element is deemed machinable through direct access or neighbor access, this radius is adopted, and the elements influenced by this radius are not analyzed further until the next iteration for computational efficiency. If there are no machinable elements within the neighborhood defined by the radius, the next radius in Table 1 is examined, and this process continues until at least one machinable element is found. This process MATLAB® code is shown in Appendix B and C.

Once the non-machinable elements are detected as explained in the epigraph 0, and the filter radii are obtained, the filtering process can begin. It is graphically described in **Error! Reference source not found.**, and it is also broken down step-by-step next.

The process starts with the blueprint design, which is the resultant density field of the design variables update in the OC loop. This is represented in Figure 11a, where blue elements are the elements with a density over the chosen threshold. Then the filter radii are obtained using the MATLAB® code of the Appendix A considering that, in this case, a machinable result in the x

direction is desired. Elements highlighted in pink in Figure 11b are the ones affected by the filter radii.

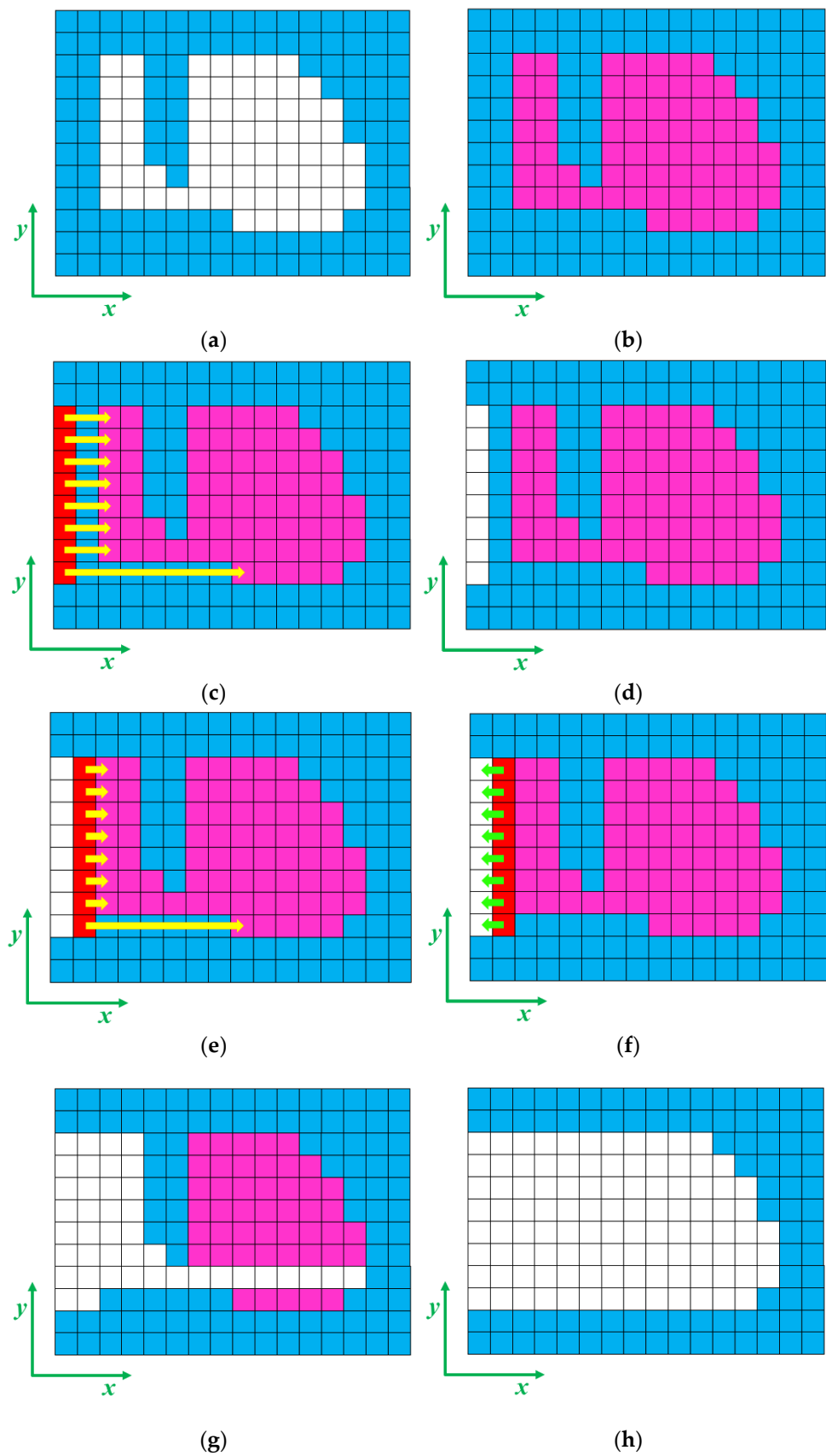


Figure 11. Machining filter step-by-step in x direction. (a) Initial blueprint design. (b) Highlighted in pink, the elements covered by the filter radii. (c) First column filtering. (d) Design after first column filtering. (e) Elements detection for subsequent columns (second column represented). (f) Filtering of

subsequent columns (second column represented). (g) Design after second column filtering. (h) Final design after filtering.

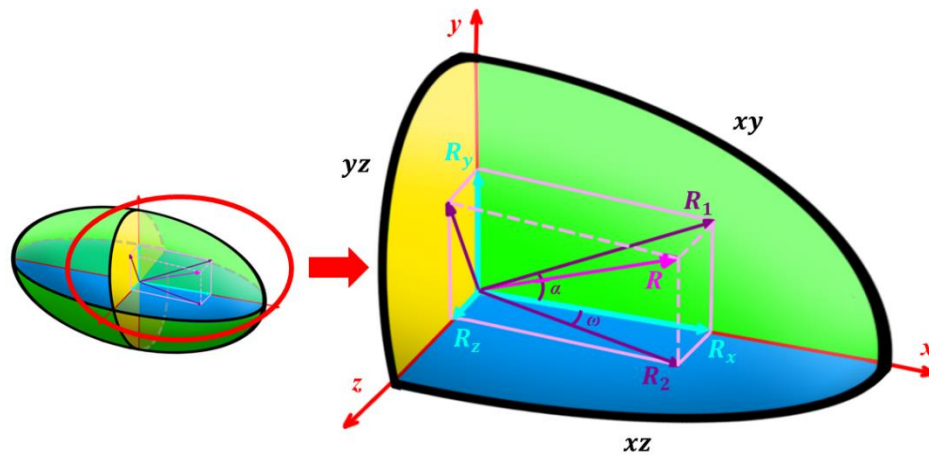


Figure 12. Geometrical parameters of the ellipsoid for obtaining its radius from the cartesian coordinates of the elements being analyzed.

Now, two conditions are evaluated: the element density (and if this density is above the density threshold) and the existence of pink elements in the machining direction at any point of the density field. If it exists an element that meets both conditions, its density is turned into zero. This is done for giving a first access to the filter for working as will be now described. This first column filtering is represented in Figure 11c, where elements in red are the elements that meet the mentioned conditions and the yellow arrows point the pink elements in the machining direction; and Figure 11d, where the design after the first column filtering is represented.

The rest of the columns are filtered in a similar way: the elements to filter are selected if they meet the conditions described above, but in this case, they take the density of the element before them in the machine direction, the $(x-1, y)$ element. In Figure 11e, the second column element detection is represented in a similar way as in Figure 11c; and in Figure 11f, green arrows represent the elements from whom the filtered elements are taking their new density.

Figure 11g represents the density field after the second column filtering. The process continues until there's no elements that meet the filtering conditions. In Figure 11h the final design after all the filtering process is represented. The MATLAB® code of the filter itself is shown in Appendix C.

Finally, a kernel filter as the described in Equation 2 is applied as damping tool. Enforce some elements to have a concrete density can derive on a disintegrated density field and thus in a difficult convergence of the OC routine. This acts as the density filter [19,36] presented in section 1 and it's also where the ellipsoid-shaped filter is added (epigraph 4.3).

2.5. Heaviside Step Filter

To achieve a manufacturable part with minimal discretization of the element densities, a chain rule consisting of close filters through a Heaviside function as the one described by Schevens (2016), Andreassen (2007) and Pellens (2019), is used first applying a dilatation, and lastly an erosion step. Sigmund [28] demonstrated the use of close filters for reducing the discreteness of the result while preserving the original shape (non-filtered shape). The utilized Heaviside Step function corresponds to the smooth approximation proposed by Wang (2011):

$$\bar{\rho}_e = \frac{\tanh(\beta\eta) + \tanh(\beta(\bar{\rho}_e - \eta))}{\tanh(\beta\eta) + \tanh(\beta(1 - \eta))} \quad (9)$$

where β corresponds to the Heaviside smoothness parameter: a β approaching zero filters the densities linearly without making any changes, while β trending to infinite causes a step function in

the value defined by the parameter η . When $\eta = 1$, the expression corresponds to an erosion filter, while when $\eta = 0$, it corresponds to a dilation filter. The aim of using first a dilation filter and later an eroding filter is to avoid a non-volume preserving strategy.

The sensitivities must be modified to ensure accurate calculation of the new design variables in every OC iteration. This is done through the chain rule described in Equation (10):

$$\frac{\partial c}{\partial x} = \frac{\partial c}{\partial \bar{x}} \frac{\partial \bar{x}}{\partial \hat{\bar{x}}} \frac{\partial \hat{\bar{x}}}{\partial \hat{\hat{x}}} \quad (10)$$

2.6. Ellipsoid-Shaped Density Filter

The main issue when extrapolating results from the sensitivity analysis to other design spaces is the mesh dependency of the density-based TO [17]. As the shape of the optimized part is directly dependant on the density filter described in Equation (2). This is caused by the relation of the filter size and the design domain size: using the same filter size with different design domains leads to different resultant shapes.

To adress this characteristic, the original sphere-shaped filter is taken as an ellipsoid with equal semi-axes, so they can be individually varied with the three cartesian dimmensions of the initial design domain. As the used weighting factor is conic weights [36], the radius of the ellipsoid at the same latitude and longitude than the analyzed element, R_{ijk}^{fil} , must be obtained for calculating the weight of the neighbor elements.

For obtaining this radius, six initial parameters are needed: a , b and c , which are the semi-axes dimmensions given by the user as three filter radii R_x^{min} , R_y^{min} and R_z^{min} respectively; and the relative coordinates of the element being analysed in the Kernel filter with respect to the base element: R_i , R_j and R_k .

The angles α and ω in **Error! Reference source not found.** are the geocentric latitude and longitude respectively, they can be obtained with trigonometry, being:

$$\begin{aligned} \alpha &= \tan^{-1}(R_y/R_x) \\ \omega &= \tan^{-1}(R_z/R_x) \end{aligned} \quad (11)$$

The coordinates of the ellipsoid surface at the same latitude and longitude than the analysed element, R_x , R_y and R_z , can be geometrically calculated. Combining these values with the general formula of an ellipsoid, Equation (12) is obtained for the calculation of the desired radius:

$$R_{ijk}^{fil} = \frac{abc}{\sqrt{c^2 \cos^2(\omega) (b^2 \cos^2(\alpha) + a^2 \sin^2(\alpha)) + a^2 b^2 \cos^2(\alpha) \sin^2(\omega)}} \quad (12)$$

The three implementations are presented and highlighted in green in the flowchart presented in **Error! Reference source not found.** The flowchart represents the level 6 of the **Error! Reference source not found.** and the blue block of the **Error! Reference source not found.**. The machining filter is divided into the three phases described in Appendices A.1, A.2 and A.3; and both density filters are computed with the ellipsoid-shaped neighborhood.

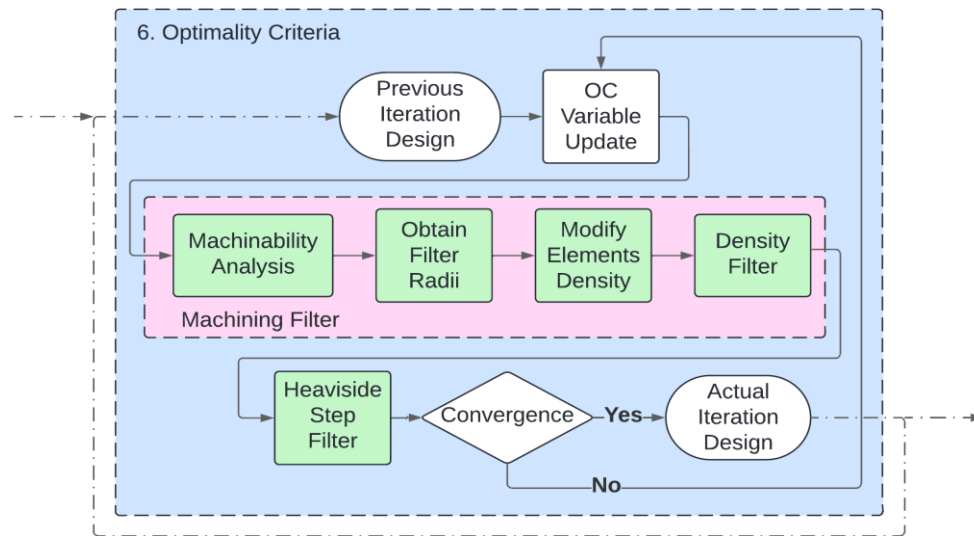


Figure 13. OC routine with the machining filter highlighted in pink and the colored in green.

3. Results

This section is divided in two parts. The first one is a validation of the proposed additions to the original TO methodology. In this subsection a concrete case of the sensitivity analysis is chosen, and the machining filter, the ellipsoid-shaped density filter and the Heaviside step filter are applied separately and at once to show the impact in the results and verify their performance. The second subsection is an epigraph explaining the conclusions reached in this study, highlighting the most useful points for the potential designers that read this paper.

3.1. Validation

In this subsection, six different validation cases are analyzed based on the subcase $R = 3$ and $p = 6$, whose data and shape can be seen in Appendices B.1 and B.2. All the figures are colored rainbow-like in the machining direction, starting in blue and finishing in red. This is done for having a clearer view of the results and the difference between the resultant shapes of the validation cases.

- The case (a) of Figures A1–A4 of Appendices C.1-4. represents the standard case without any additional filter added.
- The case (b) of Figures A1–A4 of Appendices C.1-4. is the resultant shape of the standard case in combination with the Heaviside step filter with $\eta = 0.5$.
- In Figures A1–A4 (c) of Appendices C.1-4., the ellipsoid-shaped density filter is applied to the standard case. Fixing the column of $p = 6$ in Appendices B.2-4, the radius of each cartesian direction is chosen in order to obtain the features in that axis of the case in the sensitivity analysis. In this case $R_x^{\min} = 6$, $R_y^{\min} = 5$ and $R_z^{\min} = 3.5$ are used as filter radii for obtaining a more manufacturable result with the features of $R = 6$ and $p = 6$ (Appendix B.4) in x direction, $R = 5$ and $p = 6$ (Appendix B.4) in y direction and $R = 3.5$ and $p = 6$ (Appendix B.3) in z direction.
- In (d) case, the machining filter is applied to the standard case.
- The case (e) showcases the combination of case (b) and case (c), this is the standard case with Heaviside step filter and ellipsoid-shaped density filter combined.
- Finally, the case (f) evidences the resultant shape of the combination of case (e) with the machining filter. In this case the three additions described in section 4 are applied.

For the cases (c), (d), (e) and (f), that are the cases where the Heaviside filter is applied, the searching methodology described in [37] is applied, updating the design variables every 25 iterations or when the updating conditions are met. Finally, a case (g) is showcased where the number of

iterations for updating the design variables is reduced to 10 with the conditions of case (f) for reducing the total number of iterations and search for an optional different design. A video showcasing case (g) and the AVD described in [37] is available in the Data Availability Statement section.

Relevant data of the validation are presented in Table 2, where the number of iterations before convergence, the measure of non-discreteness, the machinability and the compliance (as objective function) of all the verification cases are showed.

Table 2. Comparative relevant data of the verification cases.

	Case (a)	Case (b)	Case (c)	Case (d)	Case (e)	Case (f)	Case (g)
Iterations	1018	1738	346	2086	1087	2639	790
M_{nd}	21.5065	0.8477	28.3512	3.901	1.1824	5.2443	6.1459
Machinability	89.577	94.381	95.751	100	95.479	100	100
Compliance	38.0111	19.5020	59.1714	136.4004	21.1516	46.4197	34.3937

3.2. Conclusions

Through this study, several conclusions can be drawn regarding the parameters to be considered for designers when using TO as design methodology, as well as how to fix the issues found. The proposed solutions can be interpreted as potential implementations for commercial software packages in the future, as they haven’t implemented this kind of methodologies yet.

In the sensitivity analysis, modification over the penalization and filter radius are analyzed in a concrete 3D case. 92 simulations are showcased, being the most complete study of this kind found in the bibliography. The results are commented in section 3, but here some notes are added regarding the aim of the paper: helping designers. The interesting thing in a TO when designing functional parts is having a low interpretation of the optimized shape (this is low M_{nd}), high machinability and the lower objective function as possible for the prescribed constraints (compliance and volume fraction respectively in the case study).

The main issue detected in the sensitivity analysis is the fact that a low measure of non-discreteness M_{nd} and high machinability can’t be achieved at the same time. This can be seen in **Error! Reference source not found.**, but specially in **Error! Reference source not found.**, where the described relation is evidenced with the discrete data of the results. As the volume fraction is the same for all the simulations, the objective is clearly dependant on the penalization, obtaining better results for low penalizations as **Error! Reference source not found.** and Appendix B.1 evidence; but worse non-discreteness results are obtained for that low penalization factors (**Error! Reference source not found.**). Machinability finds a trend in the filter radius, being more manufacturable the results with high value of that radius as Figure 7 shows.

Here, the necessity of additions for obtaining the best possible results in terms of interpretation and manufacturability is detected. For this paper, three additions have been studied and applied (being two of them a novelty): the machining filter, the ellipsoid-shaped density filter and the Heaviside step filter. The machining filter have been developed for obtaining machinable parts in a concrete direction; the ellipsoid-shaped density filter has the aim of obtain results with concrete features in certain selective directions (looking for example the results of the sensitivity analysis) and the Heaviside step filter is applied for obtaining low interpretation of the results or, in other words, low measure of non-discreteness.

Table 2 and Appendix C evidence the successful addition of the three filters separately and combined. In the cases where the machining filter is applied (cases (d) and (f)), a machinability of 100% is achieved with a higher computation cost (more iterations until convergence and more time per iteration). The cases where Heaviside step filter is used (cases (b), (d), (e) and (f)), M_{nd} is sensibly lower than “standard” cases. Finally, the cases where the ellipsoid-shaped density filter is applied (cases (c) and (e)), the machinability is a bit better in comparison to those cases without the filter, but the interesting thing is achieving the described desired features of different cases seen in the sensitivity analysis.

Considering this, and specially looking to cases (f) and (g), the behaviour and effectiveness of the developed filters is verified, as a low M_{nd} is obtained while having a 100% of machinability. The incrementation of the iterations until convergence is a normal consequence of adding extra constraints (filters in this case), as well as the increase of the objective function, but modifying the AVD [37] parameters, it can get to convergence while keeping a high discreteness, a 100% machinability and low objective function compared to non-constrained designs. Geometrical restrictions introduce dramatic limitations into the possible optimum shapes, but it's the only way to mathematically ensure that an optimum design is reached and allow the designers to use the obtained shapes with low interpretation of the results.

In summary, the sensitivity analysis carried out in this study is a good tool for designers to choose the initial parameters of their TOs, and they can be useful for getting to some geometrical features through the ellipsoid-shaped density filter. The Heaviside step filter is a valuable tool for obtaining low interpretable designs (with high discretization of the density field), ensuring a good results interpretation stage. The presented machining filter have demonstrated to be an interesting tool for obtaining machinable parts in a certain direction, and it could be a powerful tool in combination with the other additions presented in the paper. Nevertheless, a convergence or searching strategy should be used (Adaptive Variable Design, that have proven its effectiveness; use of other optimization algorithms or a user custom strategy, etc.).

Author Contributions: Conceptualization, A.V.; methodology, A.V., J.M. and A.B.; software, A.V.; validation A.V., J.M. and A.B.; investigation, A.V.; resources, C.C.; writing—original draft preparation, A.V., J.M. and A.B.; writing—review and editing, A.V., J.M. and A.B.; supervision, J.M. and A.B.; project administration, C.C.; funding acquisition, C.C. All authors have read and agreed to the published version of the manuscript.

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Data Availability Statement: The original data presented in the study are openly available the appendices section of this paper. Further questions about the implementation, understanding and use of the data can be submitted to the corresponding author. A video of the case (g) has been generated and is available in the next Dropbox link: <https://www.dropbox.com/scl/fi/oxgbuo7mamzugsgxqe7m5/Case-g.avi?rlkey=ocer0yqjskmoxidv2sf6hu0ey&st=sbuxk78y&dl=0>

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Conflicts of Interest: The authors declare no conflicts of interest.

Appendix A

Appendix A.1. Machinability MATLAB® Code

```
function [machnbr,grayflag,coreflag,machflag,neigflag,periflag,eroflag] =
machnumberx(xPhys,nelx,nely,nelz,th)

machflag=zeros(size(xPhys)); % If machflag=1 means there's accesibility
coreflag=zeros(size(xPhys)); % If coreflag=1 means its not frontier
neigflag=zeros(size(xPhys)); % If neigflag=1 means acces through neighbourhood
periflag=zeros(size(xPhys)); % If periflag=1 means its the perimeter of the initial
volume
eroflag=zeros(size(xPhys)); % If eroflag=1 means its removable solid

for k1=2:(nelz-1)
```

```

    for i1=2:(nelx-1)
        for j1=2:(nely-1)
            if xPhys(j1,i1,k1)>=0.5
                % Count non frontier elements
                if all(xPhys(j1,i1,(k1-1:k1+1))>=th) & all(xPhys(j1,(i1-1:i1+1),k1)>=th)
& all(xPhys((j1-1:j1+1),i1,k1)>=th)
                    coreflag(j1,i1,k1)=1;
                end
                % Access x direction
                if machflag(j1,i1,k1)==0 && (all(xPhys(j1,1:i1-1,k1)<th) ||
all(xPhys(j1,i1+1:nelx,k1)<th))
                    machflag(j1,i1,k1)=1;
                end
                % Access through neighbour x direction
                if (all(xPhys(j1+1,1:i1,k1+1)<=th) || all(xPhys(j1-1,1:i1,k1+1)<=th) ||
...
                    all(xPhys(j1+1,1:i1,k1-1)<=th) || all(xPhys(j1-1,1:i1,k1-1)<=th)
|| ...
                    all(xPhys(j1+1,i1:nelx,k1+1)<=th) || all(xPhys(j1-
1,i1:nelx,k1+1)<=th) || ...
                    all(xPhys(j1+1,i1:nelx,k1-1)<=th) || all(xPhys(j1-1,i1:nelx,k1-
1)<=th))...
                    && machflag(j1,i1,k1)==0
                    neigflag(j1,i1,k1)=1;
                end
            end
        end
    end
end
end

for k1=1:nelz
    for i1=1:nelx
        for j1=1:nely
            % Count frontier elements
            if (i1==1 | i1==nelx | j1==1 | j1==nely | k1==1 | k1==nelz) &
xPhys(j1,i1,k1)>=th & machflag(j1,i1,k1)==0 & neigflag(j1,i1,k1)==0 &
coreflag(j1,i1,k1)==0
                periflag(j1,i1,k1)=1;
            end
        end
    end
end
end

```

```

% Percentaje of accesible elements
machnbr=((nnz(machflag(:))+nnz(neigflag(:)))*100)/(sum(xPhys(:)>=th)-sum(coreflag(:))-
nnz(periflag(:)));

thmat=zeros(size(xPhys));
thmat(find(xPhys>=th))=1;
grayflag=thmat-(machflag+coreflag+neigflag+periflag); % Flag for frontier elements

end

```

Appendix A.2. Filter Radii MATLAB® Code

```

function rfil = machradiix(nele,nelx,nely,nelz,xTilde,j1,i1,k1,th)

% RADIUS IN X DIRECTION

    raccsflag=0; raccnflag=0; racceflag=0; raccwflag=0;
    raccs=0; raccn=0; racce=0; raccw=0;
    rneigsflag=0; rneignflag=0; rneigeflag=0; rneigwflag=0;
    rneigs=0; rneign=0; rneige=0; rneigw=0;

    % RADIUS ACCESS IN X DIRECTION

    % x south direction
    radpos=0; j2=j1-1;
    if xTilde(j2,i1,k1)>=th
        raccs=NaN;
    else
        while raccsflag==0
            if j2>=1 & all(xTilde(j2,(1:i1-1),k1)<th)==1
                raccsflag=1;
            elseif j2>=1
                radpos=radpos+1;
            end
            raccs=j2-j1;
            if (j2>1 & xTilde(j2,i1,k1)<th) & raccsflag==0, j2=j2-1; else raccsflag=1; end
        end
    end

    % x north direction
    radpos=0; j2=j1+1;
    if xTilde(j2,i1,k1)>=th
        raccn=NaN;
    else

```



```

while raccnflag==0
    if j2<=nely & all(xTilde(j2,(1:i1-1),k1)<th)==1
        raccnflag=1;
    elseif j2<=nely
        radpos=radpos+1;
    end
    raccn=j2-j1;
    if (j2<nely & xTilde(j2,i1,k1)<th) & raccnflag==0, j2=j2+1; else raccnflag=1; end
end
end

% x east direction
radpos=0; k2=k1-1;
if xTilde(j1,i1,k2)>=th
    racce=NaN;
else
    while racceflag==0
        if k2>=1 & all(xTilde(j1,(1:i1-1),k2)<th)==1
            racceflag=1;
        elseif k2>=1
            radpos=radpos+1;
        end
        racce=k2-k1;
        if (k2>1 & xTilde(j1,i1,k2)<th) & racceflag==0, k2=k2-1; else racceflag=1; end
    end
end

% x west direction
radpos=0; k2=k1+1;
if xTilde(j1,i1,k2)>=th
    raccw=NaN;
else
    while raccwflag==0
        if k2<=nelz & all(xTilde(j1,(1:i1-1),k2)<th)==1
            raccwflag=1;
        elseif k2<=nelz
            radpos=radpos+1;
        end
        raccw=k2-k1;
        if (k2<nelz & xTilde(j1,i1,k2)<th) & raccwflag==0, k2=k2+1; else raccwflag=1; end
    end
end
end

```

```

% RADIUS NEIGHBOURHOOD IN X DIRECTION

% x south direction
radpos=0; j2=j1-1;
if xTilde(j2,i1,k1)>=th
    rneigs=NaN;
else
    while rneigsflag==0
        if j2>=1 & all(xTilde(j2,(1:i1),k1)<th)==1
            rneigsflag=1;
        elseif j2>=1
            radpos=radpos+1;
        end
    end
    rneigs=j2-j1;
    if (j2>1 & xTilde(j2,i1,k1)<th) & rneigsflag==0, j2=j2-1; else rneigsflag=1; end
end
end

% x north direction
radpos=0; j2=j1+1;
if xTilde(j2,i1,k1)>=th
    rneign=NaN;
else
    while rneignflag==0
        if j2<=nely & all(xTilde(j2,(1:i1),k1)<th)==1
            rneignflag=1;
        elseif j2<=nely
            radpos=radpos+1;
        end
    end
    rneign=j2-j1;
    if (j2<nely & xTilde(j2,i1,k1)<th) & rneignflag==0, j2=j2+1; else rneignflag=1; end
end
end

% x east direction
radpos=0; k2=k1-1;
if xTilde(j1,i1,k2)>=th
    rneige=NaN;
else
    while rneigeflag==0
        if k2>=1 & all(xTilde(j1,(1:i1),k2)<th)==1

```

```

        rneigeflag=1;
    elseif k2>=1
        radpos=radpos+1;
    end
    rneige=k2-k1;
    if (k2>1 & xTilde(j1,i1,k2)<th) & rneigeflag==0, k2=k2-1; else rneigeflag=1; end
end
end

% x west direction
radpos=0; k2=k1+1;
if xTilde(j1,i1,k2)>=th
    rneigw=NaN;
else
    while rneigwflag==0
        if k2<=nelz & all(xTilde(j1,(1:i1),k2)<th)==1
            rneigwflag=1;
        elseif k2<=nelz
            radpos=radpos+1;
        end
    while rneigw=k2-k1;
    if (k2<nelz & xTilde(j1,i1,k2)<th) & rneigwflag==0, k2=k2+1; else rneigwflag=1; end
end
end

fil=[raccs raccn racce raccw rneigs rneign rneige rneigw];
minValue = min(abs(fil(:)));
minIndex=find(abs(fil)==minValue);
if mean(size(minIndex))>1
    prefil=fil(minIndex);
    rfil=prefil(1);
elseif isnan(minValue)
    rfil=0;
else
    rfil=fil(minIndex);
end

end
end

```

Appendix A.3. Machining Filter MATLAB® Code

% MACHINABILITY ANALYSIS

```

[machnmbr,grayflag,coreflag,machflag,neigflag,periflag,eroflag] =
machnumberx(xTilde,nelx,nely,nelz,th);

checkflag=zeros(size(xPhys)); % Flag for not repeating elements while filtering
solidflag=zeros(size(xPhys));

if loop>=25
    for k1 = 1:nelz
        for i1 = 1:nelx
            for j1 = 1:nely
                if i1>1 && i1<nelx && j1>1 && j1<nely && k1>1 && k1<nelz
                    rfil = machradiix(nele,nelx,nely,nelz,xTilde,j1,i1,k1,th);
                    if rfil>0, j2=[j1:min(j1+rfil,nely)]; k2=[k1:min(k1+rfil,nelz)];
                    elseif rfil<0, j2=[max(1,j1+rfil):j1]; k2=[max(1,k1+rfil):k1];
                    else, j2=j1; k2=k1; end
                    for k2=k2(1:end)
                        for j2=j2(1:end)
                            if xTilde(j2,i1,k2)<th && checkflag(j2,i1,k2)==0 &&
grayflag(j1,i1,k1)==1
                                checkflag(j2,i1,k2)=1;
                            end
                        end
                    end
                end
            end
        end
    end

if loop>25
    for k2=1:nelz
        for j2=1:nely
            for i1=1:nelx
                if xTilde(j2,i1,k2) >= th && all(xTilde(j2,1:i1-1,k2)<th) &&
any(checkflag(j2,i1:end,k2)==1) && i1 > 1
                    xTilde(j2,i1,k2) = xTilde(j2,i1-1,k2);
                    checkflag(j2,i1,k2) = 1;
                    solidflag(j2,i1,k2) = 1;
                elseif xTilde(j2,i1,k2) >= th && i1 == 1 && any(checkflag(j2,i1:end,k2)==1)
                    xTilde(j2,i1,k2) = 0;
                    checkflag(j2,i1,k2) = 1;
                    solidflag(j2,i1,k2) = 1;
                end
            end
        end
    end
end

```

end
end
end
end

Appendix B

Appendix B.1. Numerical Results of the Sentitivity Analysis

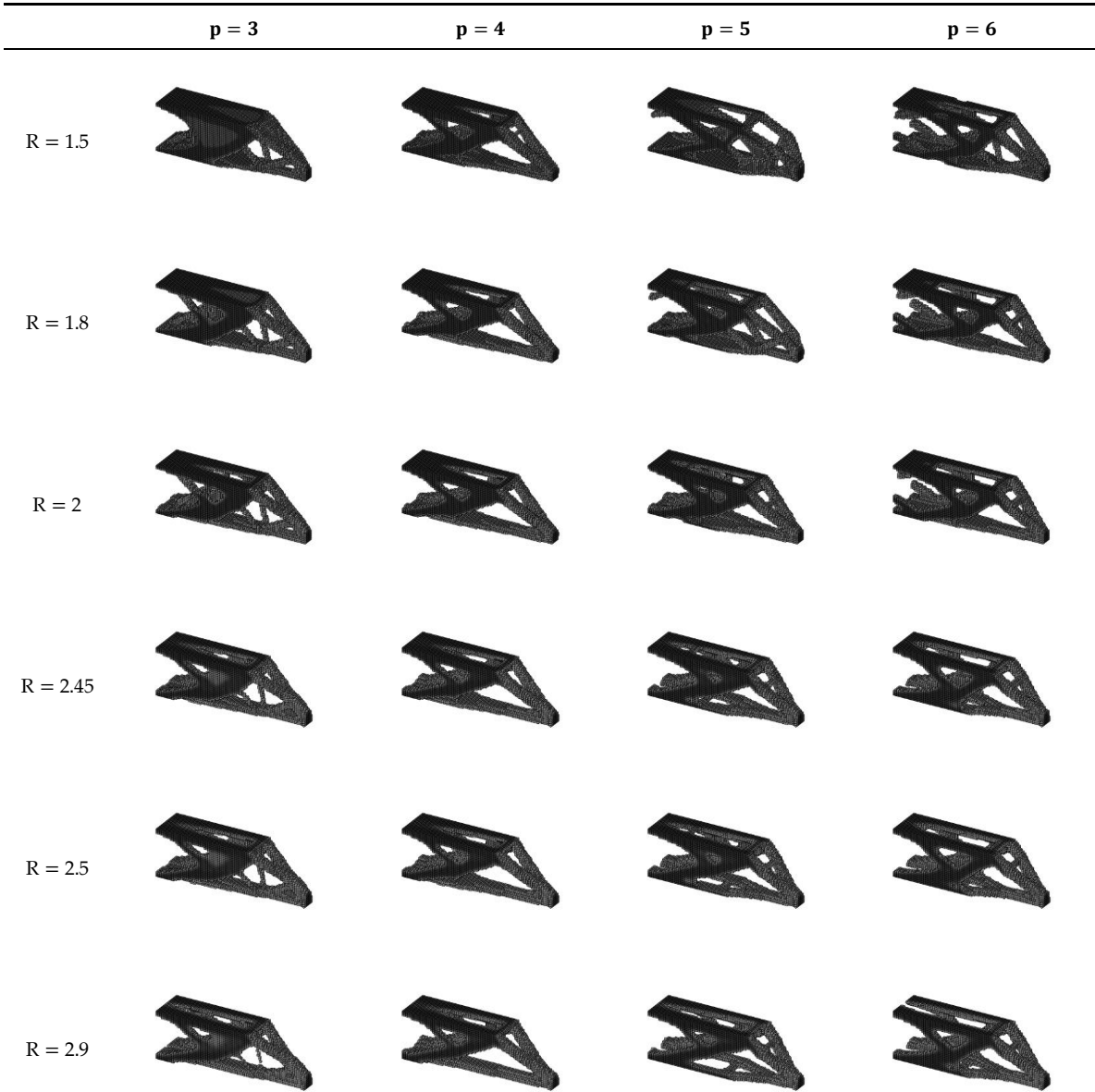
Sim. ID	Penalization	Filter Radius	Iterations	It. Time [s]	Objective	Non-Discreteness	Machinability
1	3	1,5	567	13,884	21,943	14,744	76,959
2	4	1,5	816	11,939	23,102	11,78	83,592
3	5	1,5	1677	11,938	26,6456	13,481	88,424
4	6	1,5	1122	11,919	27,5479	12,866	91,694
5	3	1,8	820	11,656	23,545	17,052	83,195
6	4	1,8	714	11,851	24,3423	14,309	88,123
7	5	1,8	765	11,813	27,4922	16,05	80,631
8	6	1,8	777	11,387	28,5529	14,818	89,976
9	3	2	1122	11,944	26,6273	17,28	85,983
10	4	2	624	11,229	25,0757	15,325	89,987
11	5	2	1048	11,618	27,2029	15,431	86,971
12	6	2	750	14,152	29,3505	15,861	94,034
13	3	2,45	1043	10,221	25,2856	20,039	91,066
14	4	2,45	845	9,914	27,9992	18,703	91,904
15	5	2,45	953	11,012	30,9835	18,375	87,688
16	6	2,45	1336	10,713	32,7265	18,054	90,878
17	3	2,5	1009	10,158	25,6883	20,603	91,927
18	4	2,5	1345	10,036	28,412	19,025	92,277
19	5	2,5	1338	11,255	31,8068	18,811	88,728
20	6	2,5	1253	11,784	33,5945	18,666	90,139

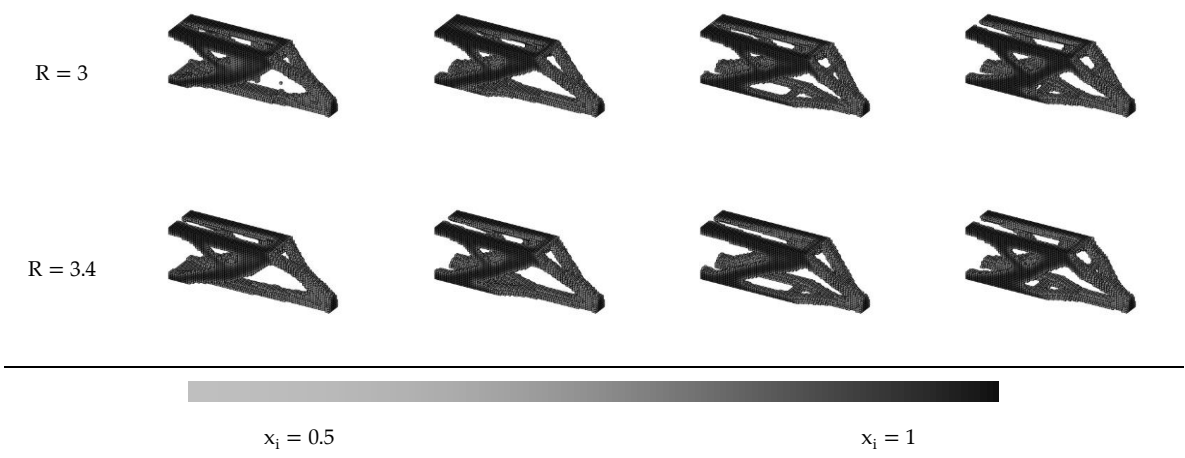
Sim. ID	Penalization	Filter Radius	Iterations	It. Time [s]	Objective	Non-Discreteness	Machinability
21	3	2,9	989	10,025	27,5435	22,934	92,992
22	4	2,9	1212	10,278	30,833	21,347	93,632
23	5	2,9	1224	10,549	35,4921	21,445	89,727
24	6	2,9	1082	11,362	37,2027	21,114	89,466
25	3	3	752	10,991	27,9193	23,285	89,966
26	4	3	1017	10,241	31,6539	21,926	92,385
27	5	3	1003	10,203	36,2139	21,896	90,352
28	6	3	1018	10,88	38,0111	21,506	89,577
29	3	3,4	628	10,129	30,2176	25,09	92,742
30	4	3,4	1089	10,766	35,8477	24,691	92,343
31	5	3,4	752	10,997	41,52	24,442	91,737
32	6	3,4	449	11,209	45,7168	24,655	89,955
33	3	3,47	1232	10,237	30,586	25,537	93,644
34	4	3,47	979	10,822	36,5405	25,151	92,466
35	5	3,47	545	9,888	42,4405	24,893	91,731
36	6	3,47	697	10,344	46,9585	25,162	89,736
37	3	3,5	1068	10,577	30,7718	25,73	93,368

38	4	3,5	890	10,576	36,8576	25,353	92,852
39	5	3,5	675	10,72	42,7654	25,069	91,84
40	6	3,5	686	10,207	47,569	25,362	91,116
Sim. ID	Penalization	Filter Radius	Iterations	It. Time [s]	Objective	Non-Discreteness	Machinability
41	3	3,8	695	14,976	32,6765	27,575	95,482
42	4	3,8	704	14,976	40,071	27,215	91,743
43	5	3,8	652	9,903	47,0157	26,87	91,478
44	6	3,8	817	11,018	54,1068	27,279	90,762
45	3	4	836	10,748	34,034	28,765	95,363
46	4	4	614	11,769	42,5076	28,438	90,458
47	5	4	626	10,367	50,9945	28,469	89,762
48	6	4	584	9,966	58,6735	28,232	92,896
49	3	4,2	617	12,87	35,3255	29,798	95,279
50	4	4,2	743	10	44,8544	29,507	92,683
51	5	4,2	915	12,313	54,5237	29,452	90,514
52	6	4,2	934	10,778	63,434	29,257	93,449
53	3	4,4	647	10,607	36,9453	31,019	95,679
54	4	4,4	501	10,435	47,8507	30,796	93,079
55	5	4,4	744	10,185	58,8192	30,438	94,141
56	6	4,4	736	9,681	69,2076	30,209	93,676
57	3	4,5	1149	9,686	37,6527	31,513	96,252
58	4	4,5	415	10,5277	49,3779	31,372	92,43
59	5	4,5	422	10,185	61,3949	31,067	93,922
60	6	4,5	580	9,973	72,4609	30,8	96,268
Sim. ID	Penalization	Filter Radius	Iterations	It. Time [s]	Objective	Non-Discreteness	Machinability
61	3	4,8	1019	10,463	40,1734	33,049	97,36
62	4	4,8	573	10,778	54,4644	33,106	93,393
63	5	4,8	699	10,476	70,182	32,808	96,059
64	6	4,8	528	10,727	84,8949	32,616	94,806
65	3	5	810	9,537	41,7655	33,98	96,755
66	4	5	907	9,736	57,4949	33,979	92,887
67	5	5	851	10,553	75,8592	33,789	92,671
68	6	5	606	11,068	93,0405	33,562	95,795
69	3	5,2	779	9,313	43,6644	35,008	97,477
70	4	5,2	1091	9,651	61,3489	35,032	92,34
71	5	5,2	1051	9,884	83,3058	34,979	92,605
72	6	5,2	337	9,899	104,346	34,754	95,6
73	3	6	535	9,224	44,6408	37,613	100
74	4	6	505	9,413	82,3081	39,494	96,195
75	5	6	1149	10,701	106,722	37,338	84,818
76	6	6	1015	10,892	138,359	37,181	85,325
77	3	8	473	9,941	73,6699	47,258	100
78	4	8	396	10,078	167,952	48,21	98,516
79	5	8	452	10,246	261,633	46,87	89,919
80	6	8	576	10,743	432,583	46,606	89,592
Sim. ID	Penalization	Filter Radius	Iterations	It. Time [s]	Objective	Non-Discreteness	Machinability

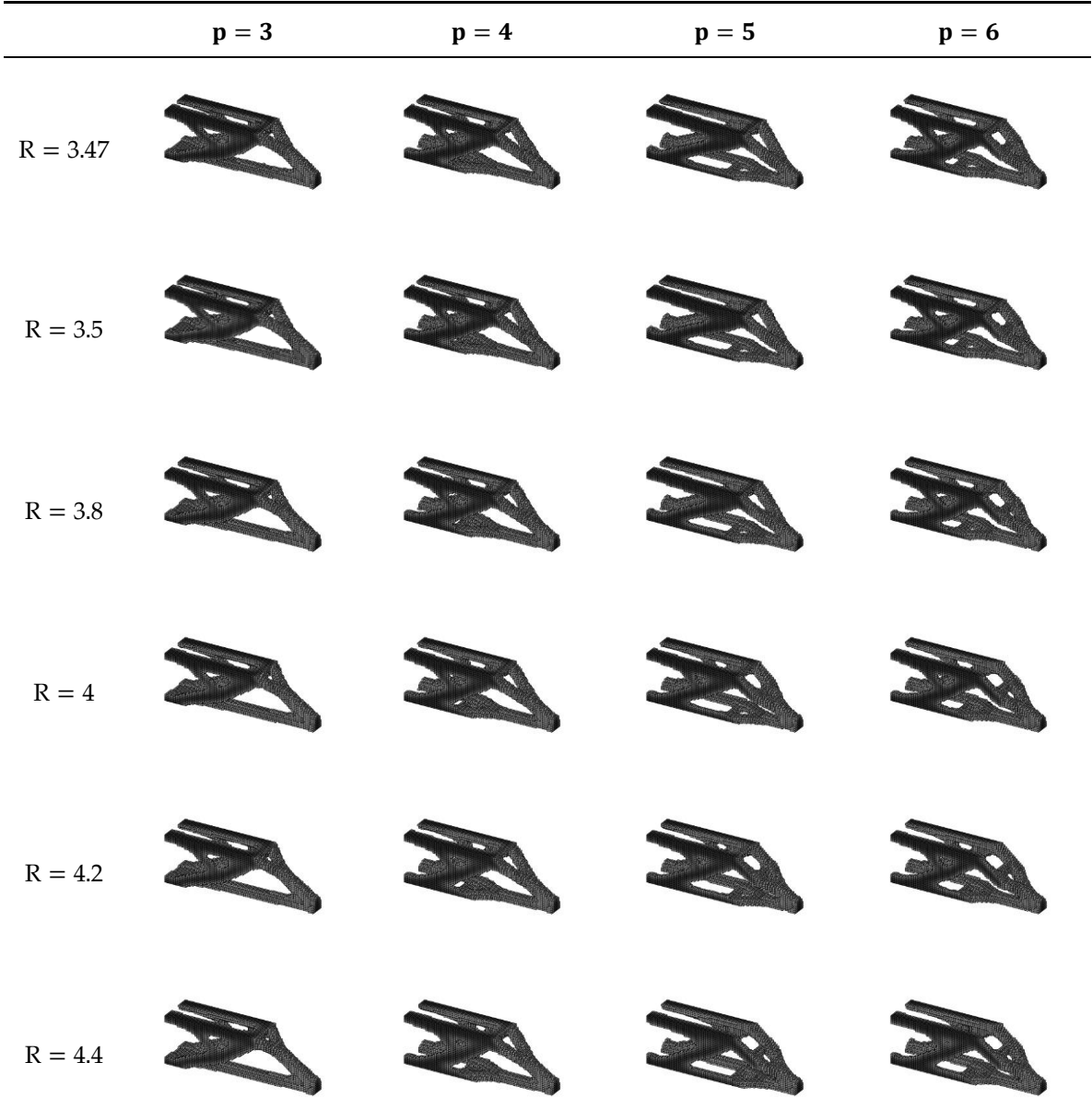
81	3	10	381	10,982	118,666	53,618	100
82	4	10	1144	13,666	284,998	53,272	98,14
83	5	10	1811	14,05	219,921	38,502	100
84	6	10	3986	17,087	314,672	37,534	100
85	3	15	238	19,017	230,282	59,919	100
86	4	15	1083	20,813	363,853	51,092	100
87	5	15	5000	NaN	NaN	NaN	NaN
88	6	15	5000	NaN	NaN	NaN	NaN
89	3	20	119	30,664	281,498	61,537	100
90	4	20	189	33,328	1150,9	61,91	NaN
91	5	20	5000	NaN	NaN	NaN	NaN
92	6	20	5000	NaN	NaN	NaN	NaN

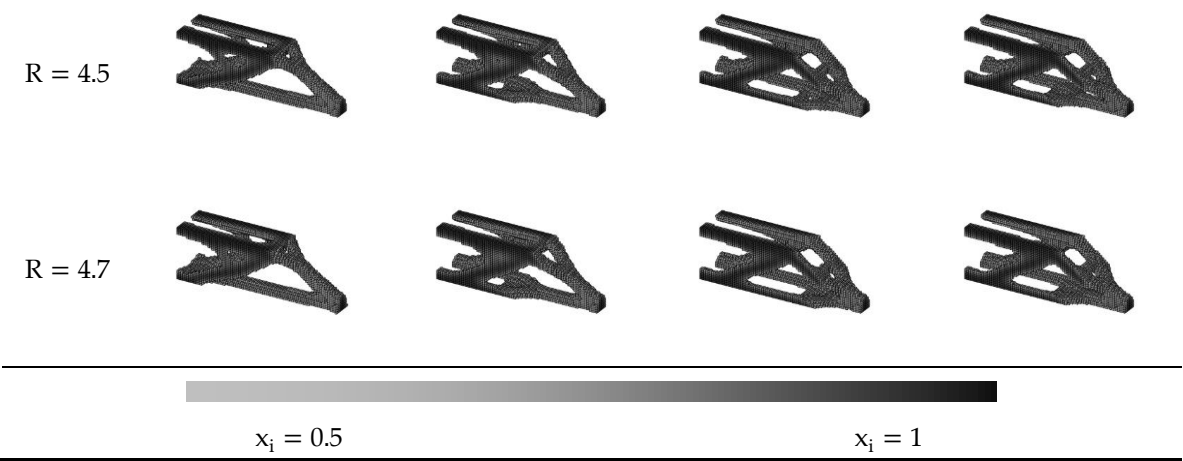
Appendix B.2. Results of the sensitivity analysis for $1.5 \leq R \leq 3.4$



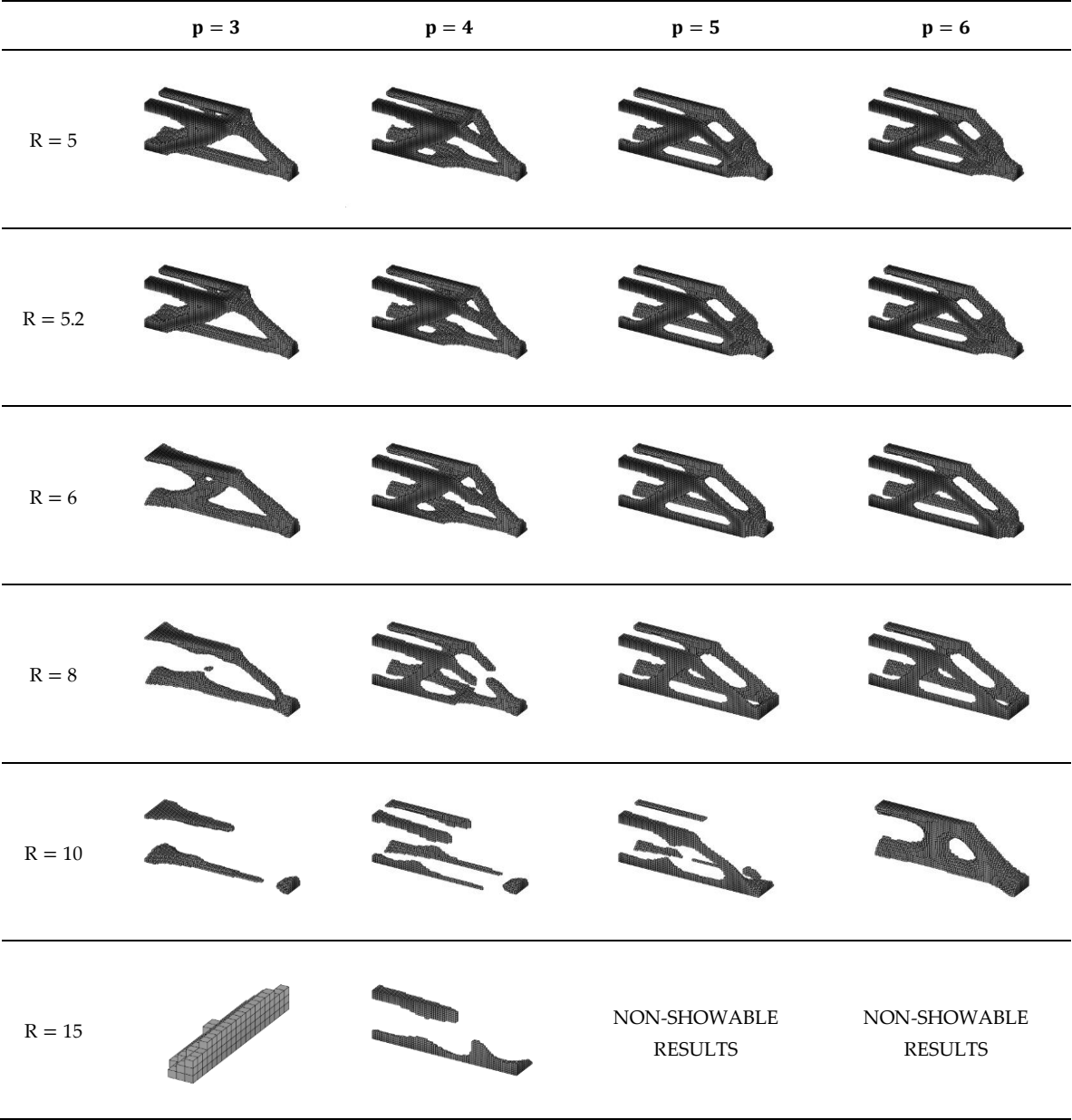


Appendix B.3. Results of the Sensitivity Analysis for $3.47 \leq R \leq 4.7$



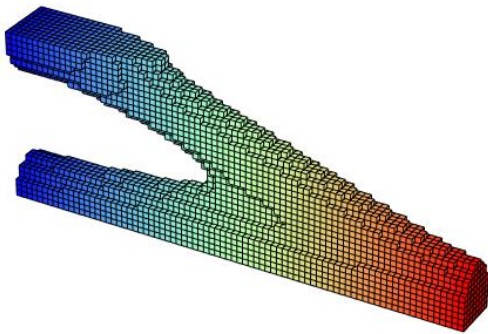


Appendix B.4. Results of the Sensitivity Analysis for $5 \leq R \leq 20$



Appendix C.1. Verifications: Superior Perspective View

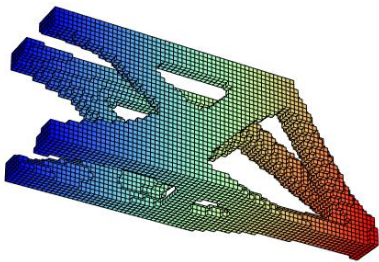




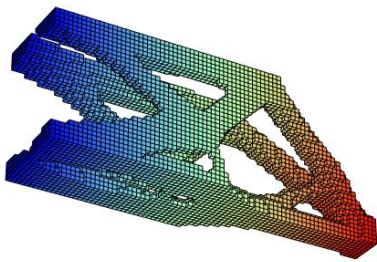
(g)

Figure A1. Superior perspective view of the different validation cases. (a) Standard case. (b) Heaviside step filter. (c) Ellipsoid-shaped filter. (d) Machining filter. (e) Ellipsoid-shaped filter + Heaviside step filter. (f) Machining filter + ellipsoid-shaped filter. (g) Machining filter + ellipsoid-shaped filter + changes in adaptive design variable method [37].

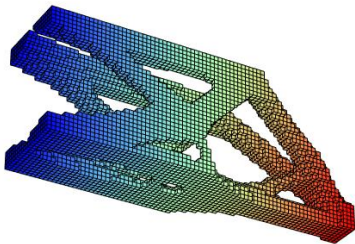
Appendix C.2. Verifications: Inferior Perspective View



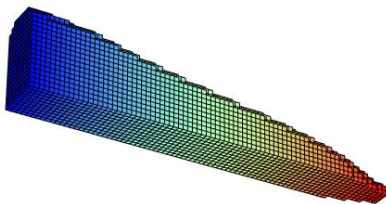
(a)



(b)



(c)



(d)

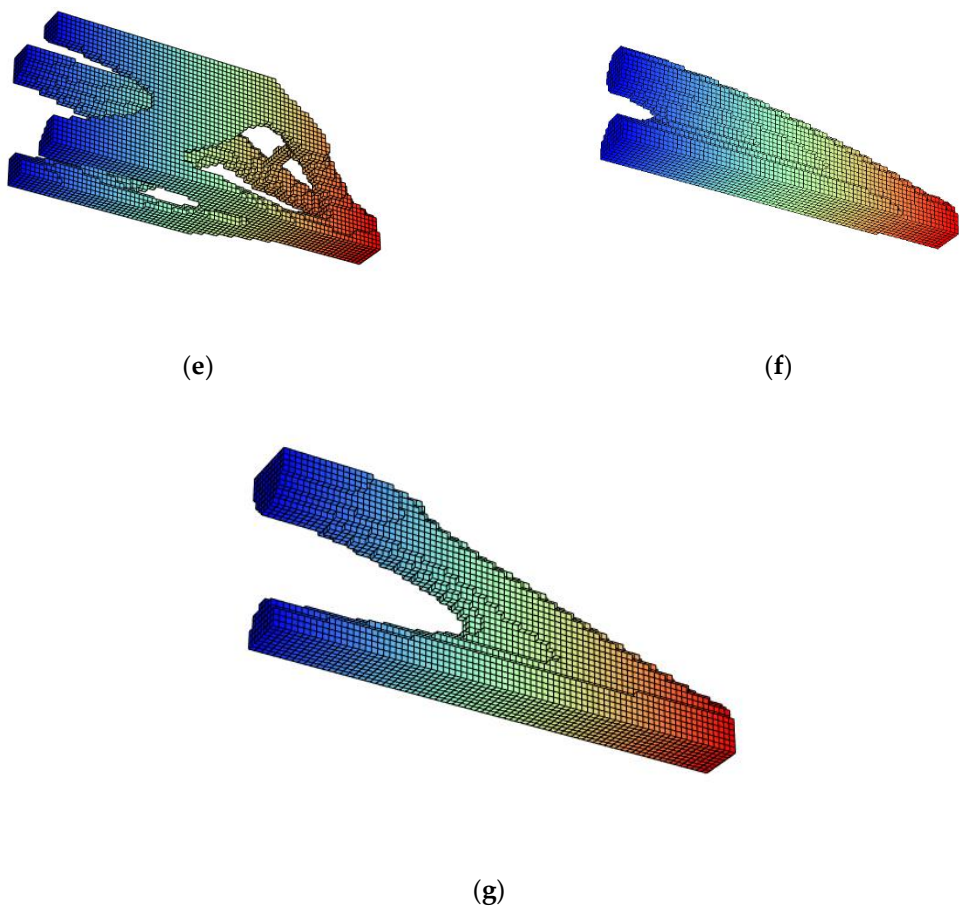
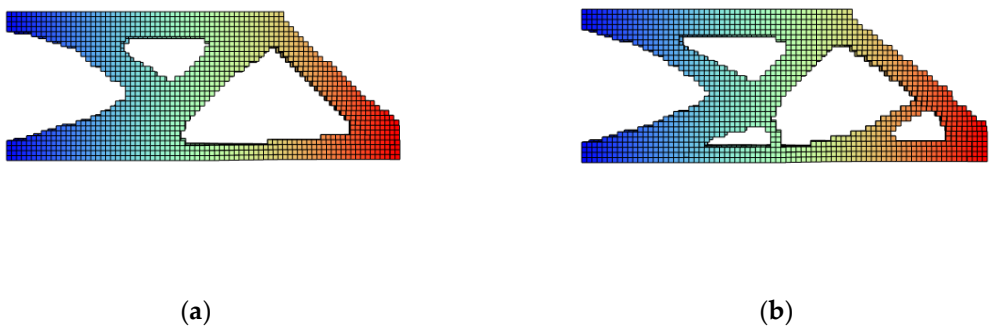


Figure A2. Inferior perspective view of the different validation cases. (a) Standard case. (b) Heaviside step filter. (c) Ellipsoid-shaped filter. (d) Machining filter. (e) Ellipsoid-shaped filter + Heaviside step filter. (f) Machining filter + ellipsoid-shaped filter. (g) Machining filter + ellipsoid-shaped filter + changes in adaptive design variable method [37].

Appendix C.3. Verifications: Lateral View



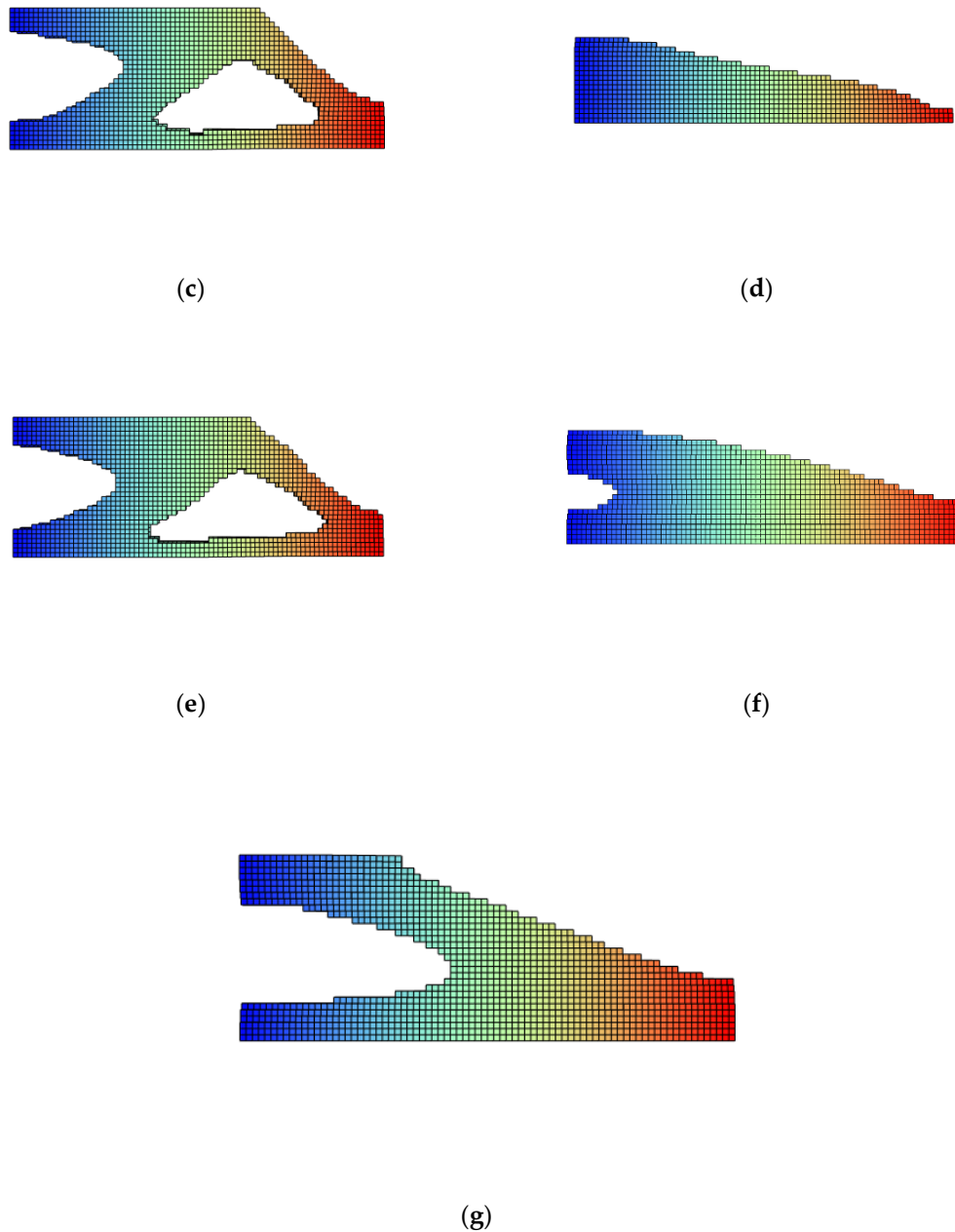


Figure A3. Lateral view of the different validation cases. (a) Standard case. (b) Heaviside step filter. (c) Ellipsoid-shaped filter. (d) Machining filter. (e) Ellipsoid-shaped filter + Heaviside step filter. (f) Machining filter + ellipsoid-shaped filter. (g) Machining filter + ellipsoid-shaped filter + changes in adaptive design variable method [37].

Appendix C.4. Verifications: Front View

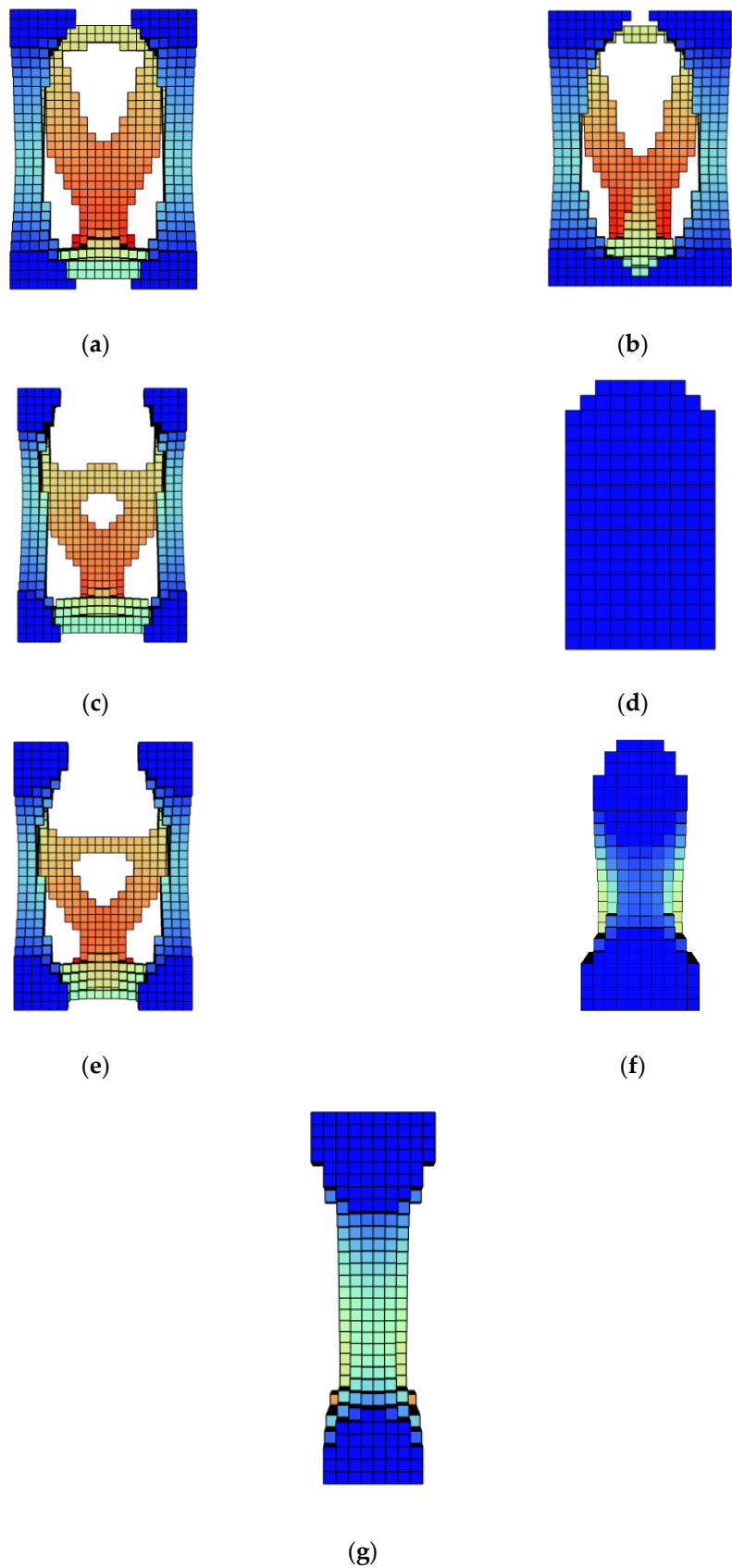


Figure A4. Front view of the different validation cases. (a) Standard case. (b) Heaviside step filter. (c) Ellipsoid-shaped filter. (d) Machining filter. (e) Ellipsoid-shaped filter + Heaviside step filter. (f) Ellipsoid-shaped filter. (g) Ellipsoid-shaped filter.

Machining filter + ellipsoid-shaped filter. (g) Machining filter + ellipsoid-shaped filter + changes in adaptive design variable method [37].

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