Design for Additive Manufacturing: Cellular Structures in Early-Stage Aerospace Design

Max M. J. Opgenoord · Karen E. Willcox

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Abstract The immense design freedom offered by additive manufacturing yields tremendous benefits, but it also raises the question of how to best go about designing components to exploit this design freedom. In early design phases, design tools need to utilize this freedom, but do so with low computational cost and with a good estimate of the final weight to correctly assess the part's influence on the overall product design. This is especially important for aerospace designs. In this paper, a design methodology is devised for cellular structures to be manufactured using additive manufacturing techniques, specifically for such early design phases. This design methodology uses adaptive meshing techniques to design the topology of the cellular structure, after which the struts of that cellular structure are optimized separately to reduce the dimensionality of the problem. The method is demonstrated for a small generic bracket and an aircraft bracket design from the literature. For these problems, our method is fast enough to evaluate many thousands of design options. This results in identifying promising candidate designs for further detailed design work.

Keywords Design for additive manufacturing \cdot cellular structures \cdot aerospace \cdot early-stage design

M. Opgenoord Department of Aeronautics & Astronautics Massachusetts Institute of Technology Cambridge, MA, USA E-mail: mopg@mit.edu

K. Willcox Institute for Computational Engineering and Sciences University of Texas at Austin Austin, TX, USA E-mail: kwillcox@ices.utexas.edu

1 Introduction

Additive manufacturing builds up products in a layerwise fashion, eliminating most of the geometric limitations present in subtractive manufacturing methods. However, designing for additive manufacturing is still not a straightforward problem. This is especially true for the design of lightweight structures, such as cellular structures. Typically, only the detailed design of those structures is considered, using expensive design tools. However, in early design phases, we also require a good estimate of the weight and performance of such a part, but typically have lower computational budget available. This paper describes a methodology to design the topology of the cellular structure such that it is aligned with the stress direction and such that its density is based on the stress magnitude throughout the part—all with low computational effort to permit incorporation in early design phases.

Research into design for additive manufacturing has mostly focused on the detailed design of (smaller) structural components, where the allowable design space and the loads are known and fixed. An example of such a detailed design study is the work by Aage et al. (2017) who performed a 1-billion-element topology optimization study on the internal wing structure of a commercial jetliner, where the resulting wing structure mirrored typical bone structures found in a bird wing. However, when additive manufacturing is used for larger parts and where the part's structural properties affect its loading (e.g., a wing structure), the effect of additive manufacturing on the overall product design also has to be considered in early-stage design phases. This is especially important in aerospace designs, for instance when additive manufacturing techniques are used for the wing—for which several design methods for additive manufacturing have been published (Stanford and Dunning, 2015; Kolonay and Kobayashi, 2015; Townsend et al., 2018). In early-stage design phases, potentially thousands of designs need to be evaluated quickly on the system level in a multi-disciplinary fashion (Martins, 2017). In such a system level design study, the loads into the structure are continually changing, requiring repeated redesigns of the structural components. Structural design tools for early-stage design phases therefore need to be low-cost, but at the same time result in designs that are sufficiently good representations of the final (detailed) design. Most topology optimization methods are more suited for detailed design studies due to their cost and lower-resolution representations of the same optimization problem can be far off from the final result. We therefore develop a design methodology specifically for early-stage design phases, with low computational cost and a good approximation of the final structural properties.

An example of a bio-inspired structure that is of interest to structural engineers is the *cellular structure*.¹ Cellular structures occur in almost any living tissue in nature, displaying optimized properties such as stiffness-to-weight ratio, strength-to-weight ratio (Gibson, 1997; Mazur et al., 2016), thermal conductivity, acoustic absorption, and gas permeability (Rehme, 2010). Furthermore, cellular structures can be more resistant to buckling (Sigmund et al., 2016), and are attractive from a fail-safe design perspective as it is typically straight forward to design for multiple load paths through a part. They can also be tailored to exhibit a specific (even negative) Poisson's ratio (Clausen et al., 2015).

In order to design structural parts for additive manufacturing in detailed design studies, topology optimization methods are often used. Such topology optimization methods are based on structural finite element methods where either each part of the domain is assigned a density determined by an optimizer in so-called Solid-Isotropic-Material-with-Penalization (SIMP) methods (Bendsøe and Sigmund, 2013; Bendsøe and Kikuchi, 1988), or where the optimizer can control the parameters of a level-set to yield a sharp boundary for the final part (Sethian and Wiegmann, 2000; Osher and Santosa, 2001; Van Dijk et al., 2013). To more efficiently generate cellular structures using such approaches, the result from a topology optimization problem on a coarse mesh can be projected down to finer length scales to yield a cellular structure (Pantz and Trabelsi, 2008; Groen

and Sigmund, 2017). Alternatively, notions of structural components such as struts can be embedded in a continuum topology optimization problem (Guo et al., 2014; Norato et al., 2015; Watts and Tortorelli, 2018).

A different approach designs cellular structures directly, through for instance ground structure optimization methods (Dorn et al., 1964; Hemp and Chan, 1966; Vanderplaats and Moses, 1972; Kirsch, 1989). A similar approach was used by Zhou (2012) to design truss-like continuum structures, optimizing for the density and orientation of members of a truss-like structure. These methods numerically approximate optimal Michel structures (Michell, 1904) using a finite number of struts, yielding a structure more similar to an animal bone structure. However, these methods are inherently limited by the original mesh on which the ground structure is generated. This can be partially alleviated by tailoring a limited number of unit cells to the stress state in different parts of the structure and using those to generate a new ground structure mesh (Graf et al., 2009).

Although additive manufacturing offers larger design freedom compared to conventional manufacturing techniques, some manufacturing constraints remain. Because the parts are built up in a layer-wise fashion they need support from the layer below them, limiting the overhang angles between subsequent layers. Even for powder-based techniques, such overhang limits usually remain to prevent warping. Selective Laser Sintering (SLS), however, does not have such overhang constraints as it is powder-based and the temperature gradients are lower. Critical overhang angles are typically $45^{\circ} \pm 10^{\circ}$ (Rehme, 2010; Schmidtke et al., 2011), but this number is strongly dependent on the manufacturing process and the material. Furthermore, the geometric resolution is limited—and dependent on manufacturing process and material—which also limits the minimum wall thickness.

Including such manufacturing constraints in structural optimization techniques is critical, otherwise the optimized model needs to be manually altered to adhere to the manufacturing constraints, resulting in a loss of optimality—as illustrated in Fig. 1. Moreover, dedicated design for additive manufacturing—including its manufacturing constraints—is of major importance for the introduction of additive manufacturing as an economically viable production technique (Vaneker, 2017). The manufacturing constraint most commonly used in topology optimization is a minimum or maximum length scale constraint (Poulsen, 2003; Guest et al., 2004; Guest, 2008). Recently, topology optimization methods started considering other manufacturing constraints, such as overhang constraints (Gaynor and Guest, 2016; Langelaar, 2016; Qian, 2017). For cellular structures, the com-

¹ With "cellular structure" we refer to a structure that consists of nodes and struts connecting those struts, where all struts can have different cross-sectional areas. When there is a repeating pattern to a cellular structure, it is typically referred to as a "lattice structure."



Fig. 1: Manufacturing constraints need to be included in the optimization to ensure the final manufactured part has optimal performance.



Fig. 2: Complete approach for designing a manufacturable load-dependent cellular structure.

mon approach is to find that printer orientation where the maximum number of struts are manufacturable and remove struts that do not comply with the manufacturing constraints (Shidid et al., 2016).

In this paper, we develop a methodology for designing optimal and manufacturable cellular structures to be manufactured using additive manufacturing. A major goal here is to reduce the computational burden of designing such cellular structures to allow for inclusion in early design stages, which is enabled by the use of low-order methods and adaptive meshing techniques. The design methodology is split in two main parts: (1) the design of the structural topology (Section 2), and (2) finding the optimal thickness of each strut (Section 3)—this approach is also illustrated in Fig. 2. This approach is not globally optimal, but allows for a large reduction in computational design effort. The designed cellular structure is then converted into a solid CAD model to allow for rendering and manufacturing (Section 4). The methodology is demonstrated for several brackets in Section 5 and concluding remarks are provided in Section 6.

2 Design of Cellular Topology

To design lightweight cellular structures, the cellular topology—i.e., the location of the nodes and the connectivity between those nodes—needs to be aligned with the load direction and needs to be fine in areas of high stress, while it can be coarser in locations of lower stress. Moreover, the cellular topology should be defined such that it is actually manufacturable.

In this work, a Riemannian metric field (Section 2.1) is used to define the cellular topology, where the metric is informed by the stress tensor throughout the available design space of the problem, which is found by solving a linear elasticity problem on a fully solid design domain (Section 2.2). The metric field is based on this stress tensor, subject to corrections for direction and positive definiteness (Section 2.3). Moreover, information about manufacturing constraints and the build direction can be included in the Riemannian metric (Section 2.4).

2.1 Riemannian Metric Fields

A cellular structure can be described from a discrete or continuous viewpoint. In the discrete viewpoint, the cellular structure consists of nodes and struts, each with a certain length, thickness, and direction. In the continuous viewpoint, the properties of the cellular structure such as the local size—i.e., the volume that each element of the cellular structure encompasses—and directionality of the cellular topology can be described as a continuous field throughout the structure, as shown in Fig. 3. The latter is a more elegant way of looking at the problem of designing a load-dependent cellular topology and also relaxes the intractability of solving a discrete optimization problem for the optimal topology. This continuous approach is demonstrated in the following.

The goal here is to design a load-dependent cellular topology—i.e., the location of the nodes and direction of the struts of the cellular topology are tailored to the specific loading condition(s) to which the part is subjected. To design such a load-dependent cellular topology, we leverage mesh adaptation techniques from the CFD community. In CFD applications, meshes need to be locally refined in areas with large gradients, such as boundary layers and shock waves. Furthermore, these are areas where the gradient in one direction is large, while being small in a perpendicular direction, requiring the mesh to exhibit a certain anisotropy. One way of prescribing anisotropy is through the notion of Riemannian metric spaces, which were first used by Diaz et al. (1997) and Hecht et al. (1997) for mesh adaptation purposes in fluid flow problems. Such a metric is a tensor field that at any point in the domain describes mesh size and anisotropy. Mesh adaptation using a metric computed from solution error is demonstrated by, amongst others, Loseille and Alauzet (2011a,b), and Yano and Darmofal (2012). Metric-based anistropic mesh adaptation has also been used to increase the efficiency of SIMP-based topology optimization methods (Ejlebjerg Jensen, 2016).

Most of these adaptive methods are developed for meshes with simplicial elements, which are also appropriate for structural applications. Triangles—or tetrahedra in three-dimensional structures—have good static structural properties and are therefore appropriate core elements for a cellular structure. Using such a metric to describe the triangulation is a natural way to include information of load paths into the cellular structure, and also to include manufacturing constraints, as we show in Section 2.4.

We now present the formal mathematical definition of a Riemannian metric field, closely following the notation by Loseille and Alauzet (2011a) and Loseille and Alauzet (2011b). A Riemannian metric field $\{\mathcal{M}(\mathbf{x})\}_{\mathbf{x}\in\Omega}$ with \mathbf{x} the physical coordinate—is a smoothly varying field of semi-positive definite (SPD) matrices on the *d*dimensional domain $\Omega \subset \mathbb{R}^d$. The edge length of a segment **ab** from $\mathbf{a} \in \Omega$ to $\mathbf{b} \in \Omega$ under this Riemannian metric is given by

$$\ell_{\mathcal{M}}\left(\mathbf{ab}\right) = \int_{0}^{1} \sqrt{\mathbf{ab}^{\top}\mathcal{M}\left(\mathbf{a} + \mathbf{ab}s\right)\mathbf{ab}} \, \mathrm{d}s$$

The metric-conforming triangulation \mathcal{T}_h is then such that all edges are close to unit length under the Riemannian metric field, $\{\mathcal{M}(\mathbf{x})\}_{\mathbf{x}\in\Omega}$, typically satisfying

$$\frac{1}{\sqrt{2}} \le \ell_{\mathcal{M}}(e) \le \sqrt{2} \quad \forall e \in \operatorname{Edges}(\mathcal{T}_h)$$

A geometric interpretation of a metric and its corresponding triangulation are illustrated in Fig. 4. As a further example of a metric field, consider the metric field in Fig. 5, which was used to generate the cellular structure in Fig. 3.

2.2 Solid Mechanics Solver

To find the continuous stress tensor, we solve the linear elasticity equations over the whole domain. The full



Fig. 3: Cellular structure and continuous fields of its properties.

linear elasticity equations are (Shames and Pitarresi, 2000)

$$\nabla \cdot \underline{\boldsymbol{\sigma}} + \mathbf{f} = 0 \qquad \qquad \forall \mathbf{x} \in \Omega \qquad (1a)$$

$$\underline{\boldsymbol{\epsilon}} = \frac{1}{2} \left[\nabla \mathbf{u} + \left(\nabla \mathbf{u} \right)^{\mathsf{T}} \right] \quad \forall \ \mathbf{x} \in \Omega \tag{1b}$$

$$\underline{\boldsymbol{\sigma}} = \underline{\boldsymbol{C}} : \underline{\boldsymbol{\epsilon}}, \qquad \forall \mathbf{x} \in \Omega \qquad (1c)$$

$$\mathbf{u} = \overline{\mathbf{u}}, \qquad \forall \mathbf{x} \in \partial \Omega_D \qquad (1d)$$

$$\underline{\boldsymbol{\sigma}} \cdot \mathbf{n} = \mathbf{t}, \qquad \forall \mathbf{x} \in \partial \Omega_N \qquad (1e)$$

where Ω is the domain of the problem, $\underline{\sigma}$ is the Cauchy stress tensor, $\underline{\epsilon}$ is the infinitesimal strain tensor, \mathbf{u} is the displacement vector, \underline{C} is the fourth-order stiffness tensor, and \mathbf{f} is the body force per unit volume. As for the boundary conditions, $\overline{\mathbf{u}}$ is the specified displacement of a boundary $\partial \Omega_D$ (Dirichlet boundary condition), \mathbf{t} is the specified stress normal to a boundary $\partial \Omega_N$ (Neumann boundary condition), \mathbf{n} is a normal vector (positive outward). Note that the : operator indicates a double contraction, such that Eq. (1c) can be written as $\sigma_{ijkl} = C_{ijkl} \epsilon_{kl}$ in index notation. \underline{C} is defined as

$$C_{ijkl} = \lambda_1 \delta_{ij} \delta_{kl} + \mu_s \left(\delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk} \right)$$

where λ_1 is Lamé's first parameter, and μ_s is the shear modulus (or rigidity), both of which are elastic moduli, defined as

$$\lambda_1 = \frac{E\nu}{(1+\nu)(1-2\nu)}$$
$$\mu_s = \frac{E}{2(1+\nu)},$$

4

where E is the Young's modulus, and ν is Poisson's ratio.

The linear elasticity equations are solved using a finite element method (FEM), specifically a hybridizable discontinuous Galerkin discretization, following the approach by Soon et al. (2009).

More details on that discretization and the solver are provided in Appendix B of (Opgenoord, 2018).²

2.3 Load-Dependent Cellular Topology Based on Metric Field

To tailor the cellular topology to the load direction and areas of high stress, we base the topology on the stress tensor throughout the solid domain. The stress tensor $\underline{\sigma}$ indicates which areas of the domain are highly loaded. Intuitively, it is desirable to have a finer cellular structure near highly loaded areas, and to have a coarser cellular structure near areas with low stress. Furthermore, the structure should align with the stress direction. Therefore, we adapt the structure to the stress tensor using metric-based mesh adaptation techniques (Diaz et al., 1997; Yano and Darmofal, 2012).

Our proposed method for computing the metric from the stress tensor applies corrections to ensure positive definiteness of the metric and to ensure that the direction of the metric aligns with the stress direction. The stress tensor is not necessarily positive definite because it distinguishes between compression and tension (Fig. 6), whereas a metric is required to be positive definite. For adaptation purposes, however, it is not important to distinguish between compressive and tensile stresses since it is the magnitude of the stresses that drives the need for local refinement.³ Therefore,

 $^{^2\,}$ The source code for this solver has been made available at github.com/mopg/luteos.jl.

 $^{^{3}\,}$ In designing the topology of the cellular structure, buckling is ignored.





(c) Example triangulation in 2D conforming only to the metric defined at the center of ellipse



(d) Example triangulation in 3D conforming only to the metric defined at the center of ellipsoid

Fig. 4: Geometric interpretation of a metric. Note that the triangulation corresponding to the metric is not unique.

to guarantee the metric is positive definite, we take the magnitude of the principal stresses to be positive, such that this metric is positive definite—i.e., it has a positive determinant. In practice, this is implemented by taking the eigenvalue decomposition of the stress tensor and taking the absolute value of the eigenvalues to compute the metric—while ensuring the eigenvalues stay above a minimum threshold to avoid a semi-definite metric.

Secondly, to align the topology with the stress direction, the eigenvalues of the metric are altered. The magnitude of the local determinant of the metric is inversely proportional to the local size of the mesh. Therefore, when the determinant of the stress tensor is large in a particular part of the domain, the element sizes of the topology are small in that part as well, which is



Fig. 5: A metric is a field of SPD tensors, illustrated using ellipses here. This metric field is used to generate the cellular structure in Fig. 3.

desired. If the stress is high in one particular direction, the local topology should be elongated along that direction, and thus the metric should be smaller in that direction. Therefore, the eigenvalues of the final metric are taken to be the inverse of the eigenvalues of the metric based on the stress tensor—while ensuring that the area under the metric stays constant. This is illustrated for two-dimensional problems in Fig. 7.

For three-dimensional problems, we take the inverse of the principal stresses while scaling the metric such that the determinant of the metric stays the same. For three-dimensional problems, the metric is therefore computed as

$$m_1 = \sqrt[3]{S^2} \frac{1}{\sigma_1}, \qquad m_2 = \sqrt[3]{S^2} \frac{1}{\sigma_2}, \qquad m_3 = \sqrt[3]{S^2} \frac{1}{\sigma_3},$$

where $S \equiv \sigma_1 \sigma_2 \sigma_3$ and m_1, m_2 , and m_3 are the eigenvalues of the final metric \mathcal{M} . The metric \mathcal{M} is used to define the topology.

While not implemented in this paper, this approach could extend to multiple load cases as well. In that case, the metric \mathcal{M} could be based on a weighted combination of the stress fields corresponding to those load cases. The incorporation of manufacturing constraints into that metric will be the same as is described in the next subsection.

2.4 Manufacturing Constraints

Although additive manufacturing allows for large design freedom, few AM processes can manufacture any type of cellular structure. Most processes are for instance limited in terms of the overhang tolerated for parts. Typically, extrusion-based manufacturing processes have more stringent overhang constraints than



Fig. 6: The stress tensor is not necessarily positive definite, but an updated metric for adaptation purposes can be made positive definite by taking the absolute value of the principle stresses.



Fig. 7: Using scaled eigenvalues for the final metric yields a topology with correct orientation. The top row illustrates what the topology would look like if the eigenvalues of the final metric were taken to be the absolute principal stresses, the bottom row illustrates the desired direction obtained by using scaled eigenvalues based on the principal stresses. \mathcal{M}' is the metric \mathcal{M} expressed in the coordinate system of the principal stresses.

powder-based manufacturing techniques, but almost all additive manufacturing processes have some form of overhang constraints.

The two most significant manufacturing constraints are the minimum geometric feature due to limited printer resolution, and the overhang angle constraint. The latter constraint is imposed because a new layer needs to have support from the structure below, as shown in Fig. 8. Even in powder-based manufacturing methods this constraint remains due to heat transfer and warping issues. We also include bridge constraints, which limit the maximum horizontal "bridge" that can be printed.

There are several different ways to include the manufacturing constraints. The minimum geometric feature size limit is straightforward to include by setting the minimum mesh size through the metric. For the overhang angle constraint, we optimize the local metric such that it is as close as possible to the original metric in the Frobenius norm, while adhering to the manufacturing constraints. In order to solve that optimization problem, the orientation and shape of the metric needs to be described geometrically. Here, we choose quaternions (Hamilton, 1844; Wie, 2008) to describe the axis orientation of the metric, and the length of the major axes (h_1, h_2, h_3) to describe the shape of the ellipsoid (Fig. 9). Quaternions are chosen over Euler or Tait-Bryan angles, because these have a singularity which can limit the convergence of the optimization algorithm. The optimization problem is described mathematically



Fig. 8: In this work, overhang angle constraints and bridge length constraints are considered.

as

$$\min_{\mathbf{q}, h_1, h_2, h_3} \|\mathcal{M} - \mathcal{M}_{\text{org}}\|_F$$
subject to
$$\frac{\mathbf{n}_1 h_1 \pm \mathbf{n}_2 h_2}{\|\mathbf{n}_1 h_1 \pm \mathbf{n}_2 h_2\|} \cdot \mathbf{n}_{\text{printer}} \ge \cos \beta_{\text{max}}$$

$$\frac{\mathbf{n}_1 h_1 \pm \mathbf{n}_3 h_3}{\|\mathbf{n}_1 h_1 \pm \mathbf{n}_3 h_3\|} \cdot \mathbf{n}_{\text{printer}} \ge \cos \beta_{\text{max}}$$

$$\|\mathbf{q}\| = 1,$$
(2)

where $\mathbf{q} = [q_1, q_2, q_3, q_4]^{\mathsf{T}}$ are the quaternions, and $\mathcal{M} = \mathbf{R} \mathbf{A} \mathbf{R}^{\mathsf{T}}$ with \mathbf{R} obtained from \mathbf{q} as

$$\boldsymbol{R} = \begin{bmatrix} 1 - 2q_3^2 - 2q_4^2 & 2q_2q_3 - 2q_1q_4 & 2q_2q_4 + 2q_1q_3 \\ 2q_2q_3 + 2q_1q_4 & 1 - 2q_2^2 - 2q_4^2 & 2q_3q_4 - 2q_1q_2 \\ 2q_2q_4 - 2q_1q_3 & 2q_3q_4 + 2q_1q_2 & 1 - 2q_2^2 - 2q_3^2 \end{bmatrix}$$

and $\boldsymbol{\Lambda}$ the eigenvalue matrix obtained from h_1, h_2, h_3 as

$$\boldsymbol{\Lambda} = \begin{bmatrix} 1/h_1^2 & 0 & 0 \\ 0 & 1/h_2^2 & 0 \\ 0 & 0 & 1/h_3^2 \end{bmatrix}.$$

The normal vectors of the major axes of the ellipse— \mathbf{n}_1 , \mathbf{n}_2 , \mathbf{n}_3 —are the columns of \mathbf{R} . $\mathbf{n}_{\text{printer}}$ is the zaxis of the printer expressed in the coordinate system of the part. Note that this optimization problem is non-convex, due to the $\|\mathbf{q}\| = 1$ constraint. A general nonlinear optimization solver is therefore used to



(a) Metric, defined by h_1 , h_2 , h_3 as well as **q** (from which \mathbf{n}_1 , \mathbf{n}_2 , \mathbf{n}_3 follow).



(b) Overhang constraints can be thought of as a cone in the direction of the printer's z-axis.

Fig. 9: Geometry of optimization problem for including manufacturing constraints in the local metric.

solve this problem, in this case an interior point optimizer, Ipopt (Wächter and Biegler, 2006). A geometric interpretation of this optimization problem is that the metric has to be oriented such that it points into the manufacturable cone (Fig. 9), while deviating as little as possible from the initial metric.

This approach to include manufacturing constraints into the metric is purely geometric, without a direct relationship to the loads through the cellular structure. An example of the result of the optimization in 2D is illustrated in Fig. 10, where in this case the metric is mostly rotated. However, even though this is a geometric approach, the impact on the load bearing capability of the structure is typically limited, since we find at each point in the domain a metric that is closest



Fig. 10: 2D example of including manufacturing constraints by optimizing the local metric.

to the original metric obtained from the stress tensor, which ensures the topology is still as closely aligned with the stress direction as possible. We demonstrate this by showing the relative specific compliance of a cellular structure—column under compression—when we change the build direction. For this cellular structure the cross-sectional areas for the struts is constant throughout the structure. We see that the specific compliance is fairly constant for small β angles, but even for large printer angles (40°) the specific compliance is only 20% higher. We show the influence of printer angle on the direction of the cellular topology of this column in compression in Fig. 12, which clearly shows that the topology is oriented along the printer z-axis. Of course, when the build direction is too different from the load direction, the load-bearing capability of the structure will suffer more. The build direction should therefore be aligned as much as possible with the main load directions in the structure. A natural extension of this work would be to wrap an optimizer around our design methodology to find the best build direction, as is for instance done by Langelaar (2016).



Fig. 11: Influence of printer angle on compliance of column in compression (same column as in Fig. 12).



Fig. 12: The cellular structure is aligned with the printer direction for this column under compression.

Once the metric is optimized to include manufacturing constraints, the topology of the cellular structure is generated using a metric-based mesh generator. Throughout this work, we use feflo.a by (Loseille and Lohner, 2010; Loseille, 2014; Loseille et al., 2017). To generate the initial mesh for which we compute the stress tensor, any mesh generator can be used, in this work typically commercial mesh generation software is used.

Even though the cellular topology is now tailored to the loads and aware of manufacturing constraints through the metric, not all struts may follow manufacturable directions. Several heuristic approaches can be taken to alter the cellular structure to ensure 100% manufacturability. One could find struts in the structure that violate the manufacturing constraints, and then perform edge swaps on them which in many cases may make the new strut manufacturable, see Fig. 13. Such edge swaps are commonly performed in meshing algorithms. For a 3D structure, however, we only perform these edge swaps on boundaries, as in the interior swapping edges may result in struts colliding with one another.

For struts that are perpendicular to the printer's z-axis, we need to ensure that they have enough support from below. In this work, we support the "horizontal" strut by finding a node under the strut and then adding struts between that node and the horizontal strut (Fig. 13).

The remaining struts that are not manufacturable could simply be removed, as is demonstrated by Shidid et al. (2016), but that does not necessarily guarantee manufacturability—one could remove a strut that was critical to support another strut—nor is there any guarantee that the resulting structure is still stable. For now, we take Shidid's approach, but a more integral approach is a topic of future work. In case the resulting structure is unstable, the overall density of the cellular structure can be altered or the build direction can be changed.

3 Optimal Cross-Sectional Area of Struts

Once the cellular topology is defined, we can apply truss optimization techniques to optimize the size of each strut in that cellular structure. We therefore use pinned-joint theory to relate nodal loads to the strut loads and stresses, thereby assuming uniaxial loading of the struts and stresses in the nodes. This is a simplification that is made to reduce the computational effort for this early-stage design method. In later design stages, the struts themselves can be allowed to take up bending loads (using frame analysis) while also including the stresses in the nodes.

The truss optimization problem for the minimumweight solution with the cross-sectional area of the struts as design variables, can be cast into a linear programming form. Take **a** and **l** to be the vectors containing the areas and lengths, respectively, of the struts. Then, for a cellular structure with n nodes and m struts in physical dimension d, we minimize the volume of the cellular structure (\mathcal{V}) by solving (Kirsch, 1993)

$$\min_{\mathbf{a},\mathbf{f}} \quad \mathcal{V} = \mathbf{l}^{\mathsf{T}} \mathbf{a} \tag{3}$$

subject to $C\mathbf{f} = \mathbf{F}$

$$-\sigma_C a_j \le f_j \le \sigma_T a_j$$
$$a_j \ge 0$$
$$f \in \mathbb{R}^m$$



Fig. 13: The manufacturability of the cellular topology can be improved through edge swaps and by adding internal support.

where **f** is a vector containing the forces in the struts, and **F** are the nodal forces—the definitions of these variables is also illustrated in Fig. 14. σ_C and σ_T are



Fig. 14: A truss structure optimization is performed to minimize the weight of the cellular structure.

the compressive and tensile stress limits, respectively. $C \in \mathbb{R}^{nd \times m}$ is the connectivity matrix between nodes and struts, which is directly determined by the topology of the cellular structure. Each column of C is the projection of a strut on the degrees of freedom of the nodes that are connected to that strut. Therefore, the matrix entry corresponding to the *i*th node and *j*th strut for a three-dimensional cellular structure is expressed as

$$C_{3(i-1)+1, j} = \mathbf{n}_{j} \cdot \hat{\mathbf{x}}$$
$$C_{3(i-1)+2, j} = \mathbf{n}_{j} \cdot \hat{\mathbf{y}}$$
$$C_{3(i-1)+3, j} = \mathbf{n}_{j} \cdot \hat{\mathbf{z}},$$

where $\hat{\mathbf{x}}$, $\hat{\mathbf{y}}$, and $\hat{\mathbf{z}}$ are the unit vectors in the x, y, and z directions, respectively. As an example, consider Fig. 15 where strut k is orientated in the xy-plane and makes a 60° angle with the xz-plane.



Fig. 15: The matrix C defines the connectivity between the struts and nodes of the cellular structure.

This plastic formulation in Eq. (3) can be reformatted as a standard linear program and can therefore be solved very quickly. This formulation, however, can only be used for single load cases with no minimum area constraints, because the solution has to be a statically determinate structure, seeing as no stress-strain compatibility conditions are taken into account (Kicher, 1966). However, we require constraints to be placed on the cross-sectional areas of the struts. Therefore, a formulation which satisfies the stress-strain relations has to be used (Ohsaki, 2011)

$$\min_{\mathbf{a},\mathbf{u},\mathbf{f},\boldsymbol{\sigma}} \quad \mathcal{V} = \mathbf{l}^{\mathsf{T}} \mathbf{a} \tag{4}$$
subject to $\mathbf{C}\mathbf{f} = \mathbf{F}$ (force balance)
 $\mathbf{B}\boldsymbol{\sigma} + \mathbf{C}^{\mathsf{T}}\mathbf{u} = \mathbf{0}$ (stress-strain compatibility
 $f_j = \sigma_j a_j$ (stress definition)
 $-\sigma_C \leq \sigma_j \leq \sigma_T$ (stress limits)
 $a_{\min,j} \leq a_j \leq a_{\max,j}$
 $\mathbf{u} \in \mathbb{R}^{n\,d}, \ \mathbf{f} \in \mathbb{R}^m$

where $B \in \mathbb{R}^{m \times m}$ is a diagonal matrix where the *j*th diagonal entry corresponds to the deformation per unit

force of the jth strut,

$$B_{jj} = \frac{l_j}{Ea_j},$$

where l_j and a_j are the length and cross-sectional area of the *j*th strut, respectively, and *E* is the Young's modulus. This elastic formulation has been known to suffer from problems with *vanishing constraints* as $a_j \rightarrow 0$, resulting in σ_j becoming independent of f_j (Rozvany, 2001). Fortunately, we specifically want to prescribe a minimum area for each strut to ensure support for each strut, and therefore no vanishing constraints occur in this problem. Finally, note that we ignore self-weight in this optimization, i.e., we assume f = 0 in Eq. (1).

The optimization problem in Eq. (4) can be extended to include buckling constraints. The critical Euler buckling load $f_{\rm crit}$ of the *j*th strut can be expressed as

$$f_{\text{crit},j} = \frac{\pi E a_j^2}{4(k_e l_j)^2},$$

where k_e is the column effective length factor. This constraint is also nonlinear in a_j , which further adds to the nonlinearity of Eq. (4). This constraint is added to the optimization statement in Eq. (4) as

$$f_j \ge -f_{\operatorname{crit},j}.$$

Multiple load cases can also be included in the strut diameter optimization problem (4) by adding additional force-balance, stress-strain compatibility, stress definition, and stress limit constraints, as well as additional displacement, force, and stress optimization variables corresponding to each load case.

As written, Eq. (4) is not convex, due to the $f_j = \sigma_j a_j$ constraint. When the minimum compliance problem is solved instead—maximizing stiffness, rather than minimizing weight—the design problem can be written as a convex optimization problem (Freund, 2004). However, we specifically require the minimum-weight solution, which results in a non-convex optimization problem. The problem is therefore solved as a generic nonlinear optimization problem using Ipopt (Wächter and Biegler, 2006).



Fig. 16: Cellular structure representation with 6 nodes and 13 struts.

4 Geometry Generation for Cellular Structures

After the cellular topology is designed and the optimal cross-sectional area for each strut in the cellular structure is found, the design of the cellular structure is essentially a list of nodes coordinates and a list of cross-sectional areas of struts connecting those nodes (Fig. 16). This representation is a cheap representation of a complicated model, and is ideal to use in designing the part. However, in order to prototype a part, this information has to be translated into a solid representation of the part (Fig. 17). Note that fillets in the joints are not taken into account in this geometry generation algorithm. Such fillets, however, are crucial to avoid large stress concentrations in the nodes of the cellular structure.

We have developed a geometry generation algorithm that can quickly transform a large heterogeneous cellular structure representation into a watertight solid representation (in either .stl format or a parametric CAD format).⁴ We consider a heterogeneous cellular structure to be a cellular structure without repeating elements (element density varies over the structure and is dependent on the direction), and consider a homogeneous cellular structure to be a structure with repeating elements. In standard mesh terminology, a heterogeneous structure can also be referred to as anisotropic, and a homogeneous one as isotropic. McMillan et al. (2015) explored similar ideas for geometry generation, but their algorithm is restricted to using period unit cells, whereas we specifically require an algorithm for heterogeneous cellular structures. Our geometry generation algorithm is described in detail in Appendix C of (Opgenoord, 2018).



Fig. 17: Solid geometry for node 1 of Fig. 16.

5 Example Designs

The design methodology in Sections 2 and 3 is demonstrated here for several test cases. A simple bracket is designed in Section 5.1, which is used as a test case to compare the performance of a homogeneous cellular structure to that of a heterogeneous cellular structure and to describe the influence of the strut crosssectional area optimization on the mass of a part. We further demonstrate the design methodology on a test case from General Electric in Section 5.2, which allows us to compare our methodology to methods from the literature.

5.1 Generic Bracket

As a first example to demonstrate our design methodology, we design a generic bracket, for which the load case is shown in Fig. 18. In the following, we compare the performance of a tailored cellular topology to that of a homogeneous cellular topology and also characterize the influence of printer angle on the manufacturability of the part. We consider the design domain and load case shown in Fig. 18. The part is to be manufactured using standard ABS plastic. The Von Mises stress throughout the part is shown in Fig. 19 for that load case. As expected the Von Mises stress is highest near the "neck" of the structure, both because that part has the lowest cross-sectional area and because of the discontinuity in the geometry of the bracket, which leads to stress concentrations.

From the stress tensor computed for the load case in Fig. 18, we compute a metric field. The cellular topology is then generated using a metric-based mesh generator, in this case feflo.a (Loseille and Lohner, 2010). The cross-sectional area of the struts in that resulting cellular are optimized to yield the minimum-weight structure, subject to stress and buckling constraints. In

 $^{^4\,}$ The source code for the geometry generation algorithm is made available at github.com/mopg/intrico.jl.



Fig. 18: Load case for bracket design example. The largest dimension is in z-direction (150 mm).



Fig. 19: Von Mises stress through domain of bracket for load case in Fig. 18.

the optimization we limit the diameter of the struts to be between 3 mm and 5 mm. This lower limit is imposed since the bracket is to be manufactured using Fused Filament Fabrication (FFF), which requires a minimum strut diameter of around 3 mm (Redwood et al., 2017). Under these constraints, the resulting optimized part weighs only 0.140 kg for a cellular structure with 401 struts. A render of this part is shown in Fig. 20.

To investigate the importance of the cellular topology, we compare the minimum weight of the heterogeneous cellular structure—which is aligned with stress directions—to the minimum weight of a homogeneous cellular structure which is not tailored to the stress directions. For fair comparison, the heterogeneous and homogeneous cellular structures have the same number of struts. For the homogeneous cellular structure then, we also optimize the cross-sectional area of each strut. For the optimization problem of the homogeneous cellular structure, the diameter constraints of the



(a) Heterogeneous cellular structure



(b) Homogeneous cellular structure

Fig. 20: Comparison of optimized heterogeneous cellular structure and optimized homogeneous cellular structure.

struts have to be relaxed slightly compared to the heterogeneous cellular structure to ensure a feasible design. The diameter of the struts is therefore constrained to be between 3 mm and 6 mm. The optimized homogeneous cellular structure weighs 0.209 kg—a 37% increase compared to the heterogeneous cellular structure. These designs are also compared in Fig. 20. We see in Fig. 20 that the homogeneous cellular structure has a few thick struts while most struts are assigned the minimum strut diameter. This could be expected, because there are only few struts that are aligned with the stress direction and therefore there are only a few struts that take up most of the load. The heterogeneous cellular structure is therefore also expected to exhibit better load-bearing capability in case one or more struts fail. Also note that the manufacturability of the homogeneous cellular structure is only 59%, whereas for the heterogeneous cellular structure it is 100%—this again is expected, because the heterogeneous cellular structure is actually designed with manufacturing constraints in mind. To demonstrate the manufacturability of the heterogeneous cellular structure, the bracket

is manufactured using an FFF desktop printer—the printed part is shown in Fig. 21.



Fig. 21: This heterogeneous cellular structure is manufactured using the Markforged Mark II desktop printer.

As a final comparison, we investigate the influence of optimizing the cross-sectional area of each strut in the cellular structure on the performance of a part. We compare the previous designs to cellular structures with a uniform cross-sectional area for all struts for both the heterogeneous and homogeneous cellular structures; we call these designs uniform cellular structures. The minimum-weight uniform cross-sectional area for the uniform cellular structures is found by a bisection method, which finds the minimum cross-sectional area for which one or more struts are at the stress limit. We compare these designs in Table 1, where we see that the uniform cellular structures are at least 51% heavier than their non-uniform counterparts.

	Mass, kg	Manufacturability, %
Heterogeneous	0.152	100
Heterogeneous	0.230	100
(uniform)		
Homogeneous	0.232	59.2
Homogeneous	0.367	59.2
(uniform)		

Table 1: Manufacturability and performance for heterogeneous and homogeneous cellular structures.

Finally, we also compare the influence of the design for manufacturability on the actual manufacturability of the resulting cellular structure. For this comparison, we design a cellular structure for various manufacturing directions by varying β for the design and then check the manufacturability of the part under different β manufacturing angles. Manufacturability here is defined as the ratio between the number of struts that follow manufacturable directions to the total number of struts in the part. The results in Table 2 show that whenever the part is to be manufactured in the same direction as it is designed, the manufacturability is 100%, validating the approach. However, when we vary the angle at which the part is manufactured, the manufacturability quickly drops to levels as low as 53%. Incorporating manufacturing constraints for the cellular structure into the design is therefore crucial.

Not only can we include critical design constraints early, the algorithm is fast as well, enabling inclusion in early design phases. Generating the inhomogeneous heterogeneous cellular structure only takes 28.5 seconds⁵ to go from outer geometry definition and loading condition to a generated .stl file.

	β (manufacturing)						
	in $\%$	-45°	-25°	0°	25°	45°	
	-45°	100	97.1	99.3	87.7	60.0	
	-25°	77.5	100	98.6	94.2	67.8	
β	0°	78.5	97.2	100	94.9	74.1	
(design)	25°	72.9	94.7	98.7	100	78.3	
	45°	53.4	86.5	98.8	96.6	100	

Table 2: Manufacturability (%) for manufacturing and designing a bracket with different manufacturing orientations.

5.2 Aircraft Engine Bracket

In 2013, General Electric (GE) launched a competition to redesign an aircraft engine bracket to take advantage of additive manufacturing capabilities. Although the deadline for the competition has long since passed, this design problem still serves as a useful test case for this work. Tang et al. (2015) used this bracket case study to demonstrate their lattice optimization algorithm, the result of which we compare our methodology against. Note that in this study only one load case is used—a 42.5 kN bearing load, see Fig. 22. This bracket is to be manufactured using Titanium Ti-6Al-4v, which has a tensile yield stress of 880 MPa, a compressive yield stress of 970 MPa, and a Young's modulus of 113.8 GPa.

 $^{^5\,}$ 28.5 seconds on one core of a 2015 Macbook Pro (3.1 GHz Intel Core i7). Note that this is including solving the FEM problem, which takes 15.5 seconds.





(b) Load case for GE bracket.

Fig. 22: The load case for the GE aircraft bracket is a 42.5 kN force applied at a 42° angle from vertical.

The Von Mises stress through this bracket is shown in Fig. 23, where it can be seen that this bracket is fairly lightly loaded, with the maximum stress in the hooks around 350 MPa. The rest of the structure is even more lightly loaded, and therefore provides a good opportunity to minimize the weight using a cellular structure.

The results in Fig. 23 are used to generate the cellular topology, after which the cross-sectional area for each strut is optimized for minimum weight, subject to buckling and compatibility constraints. In this optimization, the radius of the struts is constrained to be between 0.3 mm and 1 mm, the same constraints as used by Tang et al. (2015).⁶

The resulting cellular structure is shown in Fig. 24, which has a mass of 0.408 kg—a 24% reduction compared to the cellular structure from Tang et al. (2015), which had a mass of 0.54 kg. The difference is explained by the hooks and bolt holes being solid in that work, as well as by their use of a fine homogeneous cellular topology. It should also be noted that Tang et al. (2015) did not include buckling of the struts in their analysis, while that has a large influence on the mass—



(a) The hooks are the highest stressed part of the bracket.



(b) The internals of the bracket experience minimal stress. Note that this figure's color legend has a different scale to figure (a).

Fig. 23: Von Mises stress through domain of GE bracket for load case in Fig. 22.

without including buckling, our analysis shows that the bracket would weigh $0.286 \ kg$. Use of stress-aligned cellular structures thus yields a large mass reduction. Note that the overall winner of the GE bracket challenge had a weight of $0.327 \ kg$ for a design that included four load cases. In that design, the outer topology of the bracket was allowed to change as well. In our design that topology stays fixed—it is expected that changes in outer topology can yield a further mass reduction in some cases. Allowing the outer topology to change as well in the optimization problem is a natural extension of this work, for instance using an approach similar to Ringertz (1985). Such a topology optimization approach could be used as the next step in a design phase to get higher performance for higher computational cost.

6 Conclusion

This paper presented an early-stage design methodology for cellular structures to be manufactured using additive manufacturing. In early-stage design phases it is important to characterize a part's performance and weight accurately with low computational cost, as potentially thousands of designs need to be evaluated on a system level. The design methodology therefore only uses one FEM solution, which is used to inform

⁶ Note that these constraints on the radius of the struts violate the minimum size requirements of the GE engine bracket, but are consistent with Tang et al. (2015).



Fig. 24: Optimized cellular structure for aircraft bracket. This part weighs 0.408 kg.

a Riemannian metric field. That metric field is employed to construct the cellular topology using adaptive meshing techniques, such that the cellular topology is fine near areas of high stress, coarse near areas of low stress, and is aligned with the stress direction. This approach also allows for including manufacturing constraints, such as minimum feature size and overhang angle constraints. Once the cellular topology is designed, the cross-sectional area of each strut is optimized for minimum weight subject to stress, compatibility, and buckling constraints.

This design methodology is applied to a small bracket to show the manufacturability of the part, where it was found that the design for manufacturability improves the manufacturability of the cellular structure substantially. Lastly, an aircraft bracket from the literature is redesigned to achieve a 24% weight improvement.

The approach demonstrated in this paper should be considered as the first step in design methodology for (large) additive manufactured structures. The next step (in the preliminary design phase) could be an approach where the outer topology of the part is allowed to change. Finally, in the detailed design phase, a levelset or density-based topology optimization approach can be used.

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Conflict of Interest

On behalf of all authors, the corresponding author states that there is no conflict of interest.

Replication of Results

Parts of the results in this paper are reproducible using several packages which have been made available online. $luteos.jl^7$ is used to solve the FEM problems in this paper, which uses the mesh data structure package $divido.jl^8$ Once the cellular structure is designed, the solid geometry is generated using $intrico.jl^9$

It is not possible to make the entire code-base available as it heavily relies on a metric-based mesher throughout this paper feflo.a is used—which is not publicly available and still in development.

⁷ Available at github.com/mopg/luteos.jl.

⁸ Available at github.com/mopg/divido.jl.

⁹ Available at github.com/mopg/intrico.jl.

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