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# Paper:

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ORIGINAL ARTICLE



# Part orientation optimisation for the additive layer manufacture of metal components

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Abstract In a number of additive layer manufacturing processes, particularly for metals, additional support structure is required during the build process to act as scaffolding for overhanging features and to dissipate process heat. Such structures use valuable raw materials and their removal adds to post processing time. The objective of this study was to investigate whether a simple, single objective optimisation technique could be used to find the best orientation of the part, that would minimise the volume of support needed during the build. Not only reducing waste but potentially providing an effective and consistent approach for inexperienced users to orient components during manufacture. Software was developed using MatLab with an unconstrained optimisation algorithm implemented to search the different rotations of the part and identify the configuration with the least requirement for support volume. The algorithm was gradient based, and so multiple starting points were used to identify a global minimum. The efficacy of the algorithm is illustrated with three different case studies of increasing complexity. Additionally, the component of the final study was manufactured, which allowed a comparison between the algorithm's results and the orientations chosen by experienced operatives. In two of the three case studies, the software was able to find good solutions for the support volume minimisation. For the manufactured part,

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only one of the results matched the orientation chosen by the operators, the other was orientated in a similar way but the difference added significantly to the required support volume. Future developments of the software would benefit from incorporating the expertise of the manufacturing operative.

Keywords Optimisation  $\cdot$  Additive manufacture  $\cdot$  Part orientation  $\cdot$  SLM

# **1** Introduction

Additive layer manufacture (ALM), the process of building components in layers directly from a computer-aided design (CAD) geometry, has made increasing progress in both polymer and metal manufacture. Refinements to the process to improve production efficiency and accuracy have been studied in some detail over the past 20 years [1, 4, 12, 13]. The majority of the published research has focussed on plastic technologies. This paper describes research aimed at optimising the performance of metallic processes, e.g. selective laser melting (SLM) and direct metal laser sintering (DMLS); but, will first review some of the lessons learnt from other processes and materials.

Considering the work of Phatak and Pande [14], five distinct parameters were identified for fused deposition modeling (FDM) in a multi-objective optimisation, namely:

- Surface roughness
- Interior material
- The volume of the support structure
- Contact surface area
- Build height

The surface roughness measure used was the degree of 'staircasing' present in any particular orientation. Adjacent

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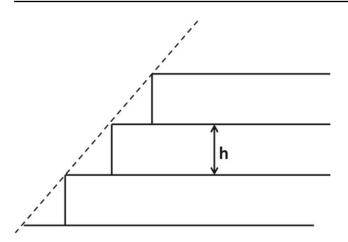


Fig. 1 Staircasing effect caused by adjacent layers of material of height 'h'

layers create a non smooth stepping effect when building an inclined plane (see Fig. 1). In FDM, the layer thickness, h, is of the order of 200  $\mu$ m; at least four times the thickness typical of SLM. A number of authors [11, 15, 20] use adaptive slicing to reduce this effect, using thinner layers in more critical areas. However, adaptive slicing is not readily available in all ALM technologies, and although it may be effective when building a single part, it becomes increasingly complex for multiple parts built simultaneously.

In metal ALM, staircasing may not be the dominant factor in terms of roughness. Maximum powder particle diameter can be larger than the layer thickness (e.g. 60  $\mu$ m powder compared to 50  $\mu$ m slice thickness) and so entrainment at the surface of partially sintered powder may have a greater impact than staircasing [17].

The surface roughness creates multiple stress concentrations which can have a negative effect on the fatigue performance of the component [6]. Research in this area indicates that post processing is necessary to increase fatigue life, particularly for titanium alloys. This may include machining to improve surface quality, heat treatment to reduce residual stresses and hot isostatic pressing (HIP) cycles to minimise internal defects [9, 22].

Most commonly, metal ALM parts are manufactured for the aerospace and medical industries and so post-processing is often essential, for example, to machine locally critical surfaces like those requiring close contact with other components, or channels for blood flow and other sensitive fluids. Under these conditions, the control of the as-built surface roughness although still important takes a lower priority. For some medical applications, high levels of surface roughness are desirable to improve bonding between an implant and living tissue, for example [5]. The freedom to reduce interior material or simply hollow out a component is more applicable to plastic parts than metals. For metal parts, strength tends to be more critical and so material removal cannot be undertaken without considering the structural integrity of the component.

Support material is required to act as scaffolding for overhanging features of design. Any parts that are inclined at a relatively small angle to the horizontal are not selfsupporting and so additional structures must be added temporarily to hold the feature in place until solidification occurs. The support material also acts as a heat sink, increasing the conduction from the melt-pool to the substrate [7]. The required angle at which this is needed varies according to the process used and can range from as low as 20° [19], with a more typical range being  $30-45^{\circ}$  [2, 21] to the horizontal. The inclusion of supports uses additional raw material and creates waste which is difficult to recycle [3]. This increases the time and costs of building the component and also the degree of post processing required to remove the supports. Distortion of the part caused by residual stress can also be reduced by using appropriate supports [21]. The area where the support contacts the part tends to show high roughness values, but this too can be reduced at post processing.

Changing a part's orientation to reduce the support volume may increase the build height and consequently the time taken to manufacture. Some authors [12, 13] have undertaken complex, multi-objective optimisation to find solutions that allow for the impact of surface roughness, part orientation and build height. In the data available to the authors, it is not clear that there is a significant correlation between build height and manufacturing time. Thus, from the factors considered by Phatak and Pande [14], the optimisation of support volume remains one of the most critical factors for improving the efficiency of metallic ALM processing.

A number of authors have proposed methods of design which completely avoid the need for support structures [3, 8], but these techniques are in their infancy and have only been tested on 2D structures in the literature. Strano et al. [16] optimised the support volume for 3D structures by calculating the support at every 5° rotational angle about the x and y axes, subsequently choosing the lowest value. This technique was also used with polymer ALM by Masood et al. [10]. However, this systematic approach may not find the most optimum orientation with a 5° resolution, particularly when considering very complex structures. Increasing the resolutions can make solving the problem very time consuming.

This paper will present a simple single objective optimisation of the build orientation of the part by minimising the support structure only. Using an unconstrained optimisation algorithm in MatLab, inexperienced operators can find an effective positioning of parts for ALM build. The use of the software provides greater accuracy and consistency than other methods and the lack of complexity in the approach enables computationally efficient solutions.

The work focusses solely on the requirements of metal powder bed SLM but may prove useful to other ALM processes.

# 2 Methodology

#### 2.1 Support volume calculation

Figure 2 shows a flowchart of the optimisation software for minimising the support volume. The script reads the geometry in the form of a sterolithography (stl) file. The stl file represents the surfaces of the geometry in terms of nonoverlapping triangular faces. The coordinates and order of the vertices of the triangles are obtained as well as the normal vectors for each face. Testing the angle of the normal

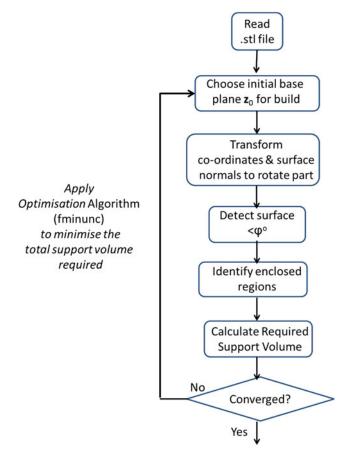


Fig. 2 Flowchart for optimisation script for part orientation

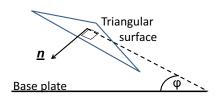


Fig. 3 Single triangular surface from the stl file showing angle to the base plate

 $(\pi/2 - \phi)$  for each face relative to the building base-plate identifies those areas of the part that require support (see Fig. 3). The critical value of  $\phi$  is dependent on the ALM process and equipment being used. The value taken in these calculations was 45°; this is the minimum angle recommended for the Renishaw AM250 SLM, which is used in each of the test cases of this paper. For each of the triangles where  $\phi$  is less than 45°, the volume of support required was obtained from the irregular triangular prism formed as the triangle was projected perpendicular to the base plate (see Fig. 4). The three vertices  $v_i$  give the coordinates for the calculation of the base area *B*. The vertical heights  $z_i$  for i = 1, 2, 3, can then be used to calculate the volume of the prism.

$$V = (z_1 + z_2 + z_3)\frac{B}{3}$$
(1)

For some triangles, other surfaces of the component lie within the field of view of the base, and so it is sufficient to have support only from those surfaces to the next feature. The total support volume takes account of this in the

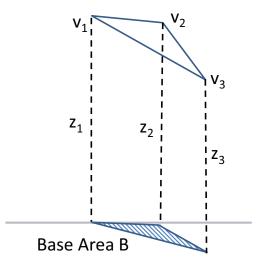


Fig. 4 Illustration of support volume calculated from projection of triangle to build plate

calculation. It is important to note that the calculated support volume considers only geometric factors. It does not assess the need for additional support based on the internal stresses generated during the build nor the requirement for heat dissipation.

# 2.2 Optimising part orientation

In order to determine the optimum orientation for building the part while minimising the support material, the volume calculation was incorporated into an optimisation function in MatLab.

The optimisation problem was unconstrained and can be simply stated as:

$$min: f(\mathbf{z}) = TotalSupportVolume$$
(2)

where the design variable, z, was the normal of the base plane of the build. An initial estimate  $z_0$  of the design variable was made. This was the normal vector for the new base-plate. A new orthogonal coordinate system was generated from  $z_0$  and the co-ordinates and surface normals from the stl file were transformed to determine the new position of the part. Effectively, the part was being rotated while maintaining the x-y plane as the base plate. When necessary, the rotated part was also translated to ensure that all features lay above the base plate. Following the flow of Fig. 2, the surfaces needing support were identified and the volume of support material was calculated, making allowance for those surfaces where the support needed to extend only to the next feature.

Using the unconstrained optimisation function, fminunc, in Matlab, the value of the design variable continued to be modified until the total support volume converged to a minimum. No information was available regarding the behaviour of the gradients of this problem; consequently, fminunc defaulted to a line search algorithm. The final design variable vector was generated together with the transformation matrix,  $U_{opt}$  which transformed the original orientation to the calculated optimum one. Since the  $U_{opt}$  transformations preserved all vector lengths, then the inverse of the transformation matrix was equal to its transpose,  $(U_{opt})^T$ , called  $V_{opt}$ . This could be expressed as the product of three rotations about the x, y and z axes, with angles  $\alpha$ ,  $\beta$  and  $\gamma$ , respectively, i.e.

$$V_{opt} = \begin{pmatrix} \cos \gamma & -\sin \gamma & 0\\ \sin \gamma & \cos \gamma & 0\\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \cos \beta & 0 & \sin \beta\\ 0 & 1 & 0\\ -\sin \beta & 0 & \cos \beta \end{pmatrix} \begin{pmatrix} 1 & 0 & 0\\ 0 & \cos \alpha & -\sin \alpha\\ 0 & \sin \alpha & \cos \alpha \end{pmatrix}$$
(3)

Multiplying the matrices and comparing terms gave solutions for  $\alpha$  and  $\beta$ , which were sufficient for orienting the part in the ALM equipment. The rotation  $\gamma$  about the z axis was not needed as the support volume was independent of this rotation.

As the line search employs a gradient-based method, a global solution could not be guaranteed in this optimisation. In an attempt to compensate for this, multiple starting points were used to improve the likelihood of identifying the best optimum.

# **3** Results and discussion

# 3.1 Support volume accuracy

The Renishaw AM250 SLM equipment used Marcam Engineering AutoFAB software to set up the part orientation and manufacturing parameters. Using stl files for a number of different components, a comparison was made between the support volume calculated by the AutoFAB software and the MatLab calculations. The results are shown in Fig. 5. The graph shows a good linear correlation between the two sets of data, with the MatLab calculation being typically, approximately three times higher than the AutoFAB. The AutoFAB supports are not completely dense, typically 1 mm support strands are separated by 2 mm gaps (see Fig. 6), which may account for the gradient of the trend line being close to one third.

There is some known loss of accuracy in the script calculations:

(i) The triangular surfaces of the stl file give an approximation to any curved surfaces. The degree of error can be reduced by increasing the resolution of the

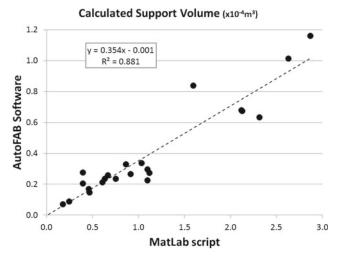
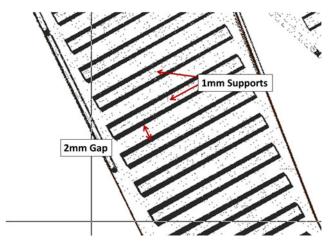


Fig. 5 Comparison of volume of support material calculated by AutoFAB and MatLab optimisation software



**Fig. 6** Example of AutoFAB support structure in cross section showing 1 mm supports with a 2 mm gap

file when created; however, increasing the number of surfaces increases the time taken to find an optimum.

(ii) A very simplistic approach has been used to test whether the support should extend to the base of the build or only to the next feature of the part below. The script used an algorithm for the intersection of a 3D ray with a triangle [18] to determine if the centroid of the face lies within the triangle of any of the surfaces at a lower height. However, there are many likely configurations where the centroid may indeed lie within the lower triangle but the upper triangle may only partially overlap the one below or vice versa. When this is the case, the volume calculation would be in error. Testing all the possible options would add considerably to the computational time and so has not been considered.

Despite these observations, the support volume predictions were considered more than adequate for purpose and the optimisation script has been trialled on a number of test cases.

#### 3.2 Cylindrical half pipe

Initial tests were undertaken using the simple geometry of half a cylindrical pipe (see Fig. 7). The pipe was 0.06 m long, with an outer diameter of approximately 0.1 m and wall thickness of 0.0145 m, consisting of only six surfaces. The stl file for the pipe had 252 triangular faces. An initial calculation of the support required in the orientation shown below was  $9.9 \times 10^{-5} \text{ m}^3$ .

Twenty different starting points for the optimisation were generated using Latin Hypercube sampling. Each of the three components of the initial base plate normal vector were sampled over the range -1 to +1. The best solution found required no support volume for the part. The

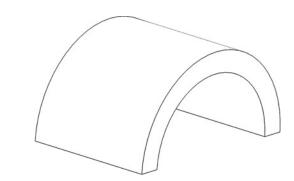


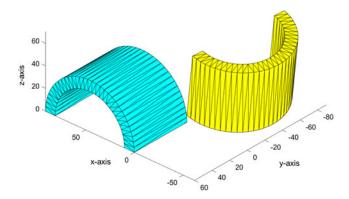
Fig. 7 Half a cylindrical pipe

optimised plane had unit normal [0, -1, 0], i.e. the x-z plane. The orientation of the optimised part is shown in yellow in Fig. 8, together with the original part in cyan. The optimisation algorithm had successfully chosen the best orientation. This solution was found by 11 of the 20 starting points, though in 3 of these the optimised normal vector was [0, 1, 0] confirming that when the pipe was rotated 180° about the *y*-axis, there is also an optimum solution. The fastest of these solutions converged in 13 iterations.

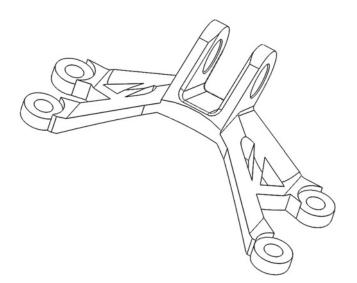
#### 3.3 Engine bracket

The second, more complex test case shown in Fig. 9 was a jet engine bracket composed of 110 surfaces which generated 6208 triangular faces in the stl file. The best solution found using multiple starting points was an optimised support volume of  $0.88 \times 10^{-5} \text{ m}^3$ . The solution was found in 11 iterations with 848 function counts.

The AutoFAB software offers two default orientations, namely '*least height*', where the part requires the smallest number of layers, and '*least footprint*', where the area of the projection of the part on the base is at a minimum. For



**Fig. 8** Optimised solution for half pipe (*yellow*) compared to original orientation (*cyan*)



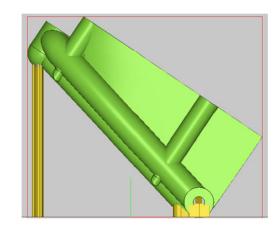


Fig. 11 Part A—best orientation from MatLab script shown in Auto-FAB software with support added

Fig. 9 Jet engine bracket design

structures used in the build. This has enabled the results of the optimisation software to be assessed against the build orientation chosen by experienced operators.

this bracket design, the fully dense support volume for these orientations was  $8.75 \times 10^{-5} \text{ m}^3$  and  $14.44 \times 10^{-5} \text{ m}^3$  for least height and least footprint, respectively. The optimised result gave a 90 % reduction in support volume over the best of these two orientations.

# 3.4 Instrument casing section

The third case study was the orientation of a casing design for an electronic component. This casing was manufactured independently by Renishaw and the as-built parts are shown in Fig. 10. Data was provided detailing the support **Part A** The best orientation found by the MatLab optimisation software indicated a rotation of  $-56^{\circ}$  about the *y*-axis only. This gave a total support volume of  $1.57 \times 10^{-5} \text{ m}^3$ . Using these rotational angles, the volume calculated by the AutoFAB software was  $0.359 \times 10^{-5} \text{ m}^3$  with an estimated build time of 33.47 h. The ratio of AutoFAB to MatLab support was 0.229 somewhat less than the gradient of the line of best fit in Fig. 5. Figure 11 shows the stl file for Part A with the support structure attached in AutoFAB.

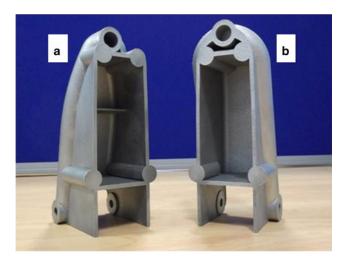


Fig. 10 ALM build casing parts showing two distinct halves a and b



Fig. 12 ALM part a affixed to base with support material still attached

Table 1Comparison ofsupport volume predictionswith manufactured volumes

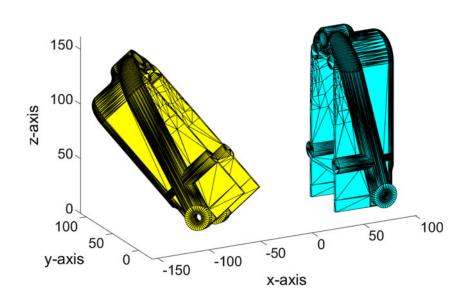
	Best optimum using MatLab script	In AutoFAB		
		Using best MatLab orient.	Using Renishaw orientation	Percent improve. by Renishaw
Support volum	$me(x10^{-5} m^3)$			
Part A	1.57	0.359	0.33	8 %
Part B	2.53	0.848	0.196	77 %
Time to build	(h)			
Part A		33.47	34.17	-2 %
Part B		37.7	36.83	2 %

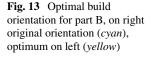
Stl files of the actual support material used were supplied by Renishaw after the build was completed. A photograph of part A with the support material in place is shown in Fig. 12. The positioning of the part is equivalent to a rotation of 0 and  $-46^{\circ}$  about the *x*- and *y*-axes, respectively. Subsequently, the support volume calculated for the built part by AutoFAB was  $0.33 \times 10^{-5} \text{ m}^3$ , with a build time of 34.17 h. The improvement in support volume was 8 %, with 2 % increase in the build time, as compared to the MatLab optimisation result.

**Part B** A similar exercise was carried out for part B. All the results are summarised in Table 1. The predicted optimum

occurred at rotational angles of  $18^{\circ}$  about the *x*-axis and  $45^{\circ}$  about the *y* (see Fig. 13). When built, however, this part was rotated  $45^{\circ}$  about the *y*-axis only and required only the support material shown (see Fig. 14).

In summary, it can be seen that for part A, the orientation predicted by the algorithm was similar to the build. Only a small reduction in support volume was achieved with the manufactured part and this took slightly longer to build. For part B, the optimiser did not find the best solution. The tilt of the part by just  $18^{\circ}$  from the base plate about the *x* axis made a very large difference to support volume, with little impact on the build time.





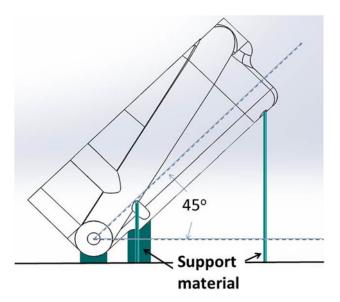


Fig. 14 Geometry of part B in as-built position showing support material attached

#### 3.5 Computational efficiency

Figure 15 shows the time taken to find the best orientation of a series of parts minimising the support volume. The data is plotted as a function of the number of triangular surfaces formed in the stl file for the component. The graph shows a strong positive correlation; though in reality, the time is more closely related to the number of downward facing surfaces in any orientation, with these not easily predicted from the initial file. The graph, therefore, gives a good indication of the expected time to solution.

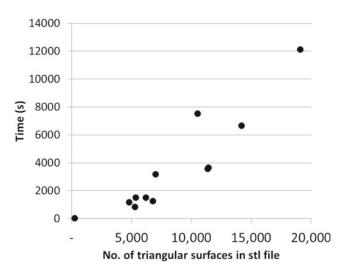


Fig. 15 Time taken to find global optimised build orientation solutions for different geometries

#### **4** Conclusions

Software has been developed to predict the optimal orientation for minimising support volume in SLM or DMLS additive layer manufacture. The procedure has been tested on three different geometries and shown to bring improvements over the standard orientations recommended by preprocessing software. When compared to the build positioning selected by experienced operators, the software did not always find the best result. Although this code would enable consistent and efficient builds to be undertaken by inexperienced users, incorporating the expertise of the manufacturers in future developments would further enhance its efficacy.

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