# Towards predicting process parameters for selective laser melting of titanium alloys through the modelling of melt pool characteristics

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# Abstract

Various researchers have investigated the use of experimental melt pool characterization to speed up the optimisation process of selective laser melting parameters and found this to be possible. From their studies, it has become clear that the incorporation of modelling into this approach could provide efficient and predictable results that would minimise the experimental validation work required. This paper reports on progress made towards characterising the melt pool through simulation. The development of a numerical model is discussed. Subsequently, experimental validation of the numerical model is presented. This is done through a comparison of the simulation results with the experimentally determined cross-sectional geometry of single tracks created with various sets of process parameters. Melt pool size and shape are considered. Based on these results, it is concluded that using an identical simulation setup, verified simulation method, and verified material properties, it was possible to accurately determine the melt pool geometry for some, but not all process parameters.

Keywords: Selective laser melting, melt pool characteristics, Ti6Al4V, modelling, process parameter optimisation

#### 1. Introduction

The selective laser melting (SLM) process involves a large number of process variables, with many of these being interdependent. This makes the process of fully optimising the process parameters for new materials challenging, as well as expensive and timeconsuming (Sun, et al., 2017). The reason for this is the extensive experimental work required to produce statistically convincing data for proving the repeatability and reliability of the process (Nicoletto, et al., 2018]). Therefore, a novel approach such as modelling is needed to reduce the experimental work required for the optimisation process.

Gong, et al. (2014) attempted to rapidly optimise process parameters by characterising cross-sectional melt pool dimensions and geometry. Their study concluded that the characterisation of the melt pool provided helpful information in determining ideal process parameters. In particular, it was seen that the geometry of the melt pool could be used to accurately predict the optimal hatch spacing distance. Lastly, and of greatest importance to this study, it was concluded that the incorporation of simulation work could predict process parameters with greater efficiency. Recently, multiple new numerical models have been created (Rauniyar & Chou, 2019) (Majeed, et al., 2019) (Ridolfi, et al., 2019). With the accuracy of these models continuously improving, the possibility of using numerical modelling to predict process parameters becomes a greater possibility. This paper forms part of a broader study that is attempting to achieve the above-mentioned. The setup of a numerical model of an SLM single track on a bare Ti6Al4V

plate with no powder and the experimental validation thereof are presented and discussed.

#### 2. Material and methods

For the validation of the model, a systematic approach was used. Firstly, the complexity of the model was reduced as much as possible. This was done by removing the powder layer from the substrate and not considering multiple layers or tracks. Consequently, the SLM process was reduced to the movement of a laser along a bare metal surface in a straight line (single track). This removed the added complexity of powder layer thickness, hatch spacing, particle spreading and laser absorptivity inherent to a powder bed.

When considering single tracks in SLM, there are only two userinput parameters that greatly influence the melt pool dimensions. These are laser power and scanning speed. To validate a numerical model that will be used to determine unknown melt pool geometries at various process parameters, the accuracy of the model at the extremes of these parameters needed to be investigated. Therefore, it was decided to simulate four distinctly different melt pool phenomena with widely different process parameters. The first of these was weak penetration, which can be identified by partial melting of the powder layer. Secondly, an optimal process resulting in a uniform melt pool was considered. The third phenomenon that was considered was balling, which results in separation of the melt pool due to high scanning speeds (Gibson, et al., 2015) Lastly, the keyhole phenomenon (King, et al., 2014), caused by a too high laser energy density, was evaluated. The process parameters used in this study to simulate the above-mentioned phenomena are given in Table I.

Parameter set	Laser power (W)	Scanning speed (m/s)	
Weak penetration	80	1.2	
Optimal (Yadroitsava, et al., 2015)	170	1.2	
Balling	350	2.0	
Keyhole melting	340	0.6	

#### Table I: Various parameter sets used for modelling

# 2.1 Numerical model

The numerical model was developed using the additive manufacturing (AM) package of the commercial software FLOW-3D (FLOW-3D, 2021). This software was used to apply a finite volume method (FVM) model to determine the cross-sectional width and depth of a melt pool created by the SLM process.

FLOW-3D numerically solves the equations of mass-, momentumand energy-conservation. The free surface of the melt pool is

Table II: Material properties used in the model

tracked by utilising the volume-of-fluid (VOF) technique. More detailed explanations of the equations and techniques can be found in papers utilising the same software (Cheng, et al., 2019) (Wu, et al., 2017). This approach enables the accurate simulation of the geometry of the melt pool, as well as the interface of the free surface with the surrounding air. The simulation was set up assuming incompressible flow of the melt.

A simulation domain with width x breadth x depth of  $2400 \times 760 \times 395 \ \mu m$  was used for all but one simulation. As the keyhole parameters lead to a deeper melt pool, a depth of  $495 \ \mu m$  was used for this simulation. A track length of  $1600 \ \mu m$  was simulated for all process parameters to ensure that steady-state was reached within the simulation. The top boundary was set as a constant pressure outflow, and all other boundaries were defined as walls at room temperature (293 K). This most closely represents physical reality.

After conducting a study of mesh dependency, it was found that using a cell size of 5  $\mu$ m for the smaller area in which the melt pool formed, and a larger cell size of 20  $\mu$ m in the surrounding areas to

Table II: Material properties used in the model				
Material property	Value	Sources		
Convective heat transfer coefficient	100 W/mK	(Cengel & Ghajar, 2015)		
Surface tension coefficient	1.493-4x10 <sup>-4</sup> (T-T <sub>m</sub> )	(Egry, et al., 2010) (Mohr, et al., 2020)		
Density	Temp dep.	(Mills, 2002) (Schmon, et al., 2017)		
Viscosity	Temp dep.	(Mohr, et al., 2020) (Lu, et al., 2017)		
Specific heat	Temp dep.	(Boivineau, et al., 2006) (Milosevic & Aleksic, 2012)		
Thermal conductivity	Temp dep.	(Mills, 2002)(Mohr, et al., 2020)		
Saturation temperature	3315 K	(Brandes & Brook, 1992)		
Latent heat of vaporisation	9830 kJ/kg	(Tan, et al., 2019)		
Specific heat ratio	1.67	(Ford & Lee, 2001)		
Vapour specific heat	695 J/kgK	(Brandes & Brook, 1992)		
Absorptivity	0.41	(Boley, et al., 2016) (Xie, et al., 1997)		
Emissivity	0.32	(Mohr, et al., 2020)		



Figure 1: Graphical representation of the simulation domain. The use of larger cell sizes on the edges can be seen here





Figure 2: Micrographs of the cross-sections of a.) weak penetration b.) optimal c.) balling and d.) keyhole melting process parameters

account for thermal diffusion would be sufficient. The domain and its meshing can be seen in Figure 1

Physics models activated in the setup included temperaturedependent material properties, volumetric thermal expansion, evaporation, solidification, gravity and all modes of heat transfer. The heat transfer modes included conduction between different phases (solid, liquid and vapour), convective cooling through the air above the bed, as well as radiative cooling to the surroundings. The material property values used in the model, as well as references to the sources in which the values were found, are provided in Table II.

The FLOW-WELD package was used to simulate the laser energy applied to the top free surface of the substrate. The laser used in the experimental work had a Gaussian heat distribution with a spot diameter of 80  $\mu$ m. The same spot diameter was used in the setup of the simulation. Laser power and scanning speed for the individual simulations were set according to the parameter sets given in Table I. Multiple Fresnel reflections of the laser were enabled. Tracking of the laser was stopped after 99% of the laser energy had been absorbed to account for most of the heat applied to the free surface due to the laser.

This setup resulted in a mesh with 2,4 million cells for the keyhole mesh and 1,8 million cells for all other meshes. The simulation

times varied substantially due to the different parameter sets and domain sizes used. The total simulation time varied between six hours for the weak penetration parameter set and more than twenty hours for the keyhole parameter set. This was achieved by using a computer with an Intel 10980XE processor with 16 cores and 128 GB of memory.

# 2.2 Experimental procedure

Samples were created with an EOS M280 machine according to the process parameter sets in Table I on a bare Ti6Al4V plate. No specific surface finish was obtained before processing. For each parameter set, fifteen single tracks, each with a length of 20 mm, were created. After the tracks had been created, samples were cut from the plate using wire electrical discharge machining to expose the cross-sections of the molten tracks. After the samples had been cut, they were mounted, polished and etched with Kroll's reagent according to standard sample preparation techniques to expose the microstructure of the samples. From these samples, measurements of the depth and width of all single tracks were recorded using optical microscopy.

Table III: Depth and width of the melt pool for the different parameter sets

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Parameter set	Weak penetration	Optimal	Balling	Keyhole
Average depth (µm)	23,2	75,7	66,7	152,4
Standard deviation (µm)	1,10	1,52	2,20	67,75
RSD (%)	4,74	2,01	3,29	44,44
Average width (µm)	73,9	117,9	106,2	195,0
Standard deviation (µm)	2,32	1,65	1,52	9,65
RSD (%)	3,15	1,40	1,43	4,95





Figure 3: Cross-section results of the simulations of a.) weak penetration process parameters b.) optimal process parameters c.) balling process parameters and d.) keyhole melting process parameters

# 3. Results and discussion

# 3.1 Experimental results

In Figure 1 cross-sectional micrographs from the various process parameter sets can be seen, while Table III shows the depth and width measurements determined from the cross-sectional micrographs.

For most of the parameter sets, the creation of the melt pool was steady, with only small deviations in the cross-sectional melt pool geometry. This is confirmed by the relative standard deviations (RSD) in Table III that are all less than 5%, except for the keyhole depth. For the keyhole parameter set, however, the depth of the melt pool varied considerably, ranging from 57  $\mu$ m to almost 250  $\mu$ m. Therefore, Figure 1(d) does not represent a typical result for the keyhole mode melting. This figure also shows keyhole induced porosity.

# 3.2 Simulation results

Figure 2 shows the cross-sections of the simulations that were run. The blue colour indicates the metal substrate, and the red colour the molten region. The coloured region between the blue and red regions indicates a region of partial melting.

The variation of the dimensions of the simulated melt pools from the experimentally determined dimensions is shown in Table IV.

#### 3.2.1 Weak penetration parameter set melt pool

When comparing the experimental melt pool shape with the simulation result, it is clear that the melt pool depth of the simulation is considerably deeper than the actual depth (Table III and IV). It differs by more than 70% from the experimentally

obtained values. The width of the melt pool compares well, with less than 10% difference between the simulation and experimental results. It was initially thought that the difference in depth might be due to enhanced conductivity as a result of a too coarse mesh. Subsequent simulations, however, indicated that although reducing the cell size to 3  $\mu$ m did result in the melt pool depth decreasing, it was still more than 60% deeper than the experimental values. Possible reasons for this discrepancy are discussed in Section 3.3. The latter simulation resulted in a melt pool width of 76  $\mu$ m, differing from experimental results by less than four per cent.

# 3.2.2 Optimal parameter set melt pool.

The melt pool dimensions for the optimal process parameters were very close to those for the experimental samples, with dimensions varying by less than three per cent, falling well within two standard deviations of the experimental results. When critically analysing the difference in melt pool geometries, it was found that although the melt pool shapes correlate well, there were still slight variations, such as the difference in the angle of the sidewall of the melt pool, as well as a melt pool that protruded higher above the surface for the simulation.

#### 3.2.3 Balling parameter set melt pool

When comparing the melt pools of the balling parameter set, these varied by approximately twenty per cent from the experimental values. The peculiar melt pool shape (protruding upper part) obtained by the balling parameter set can be seen in the simulation shown in Figure 2(c). This result is, however, believed to be by chance, since no other cross-section along the track had this shape. Apart from this, the dimensions of the melt pool along the length of the simulated track remained stable.

Table IV: Simulation dimensions and variation from experimental dimensions for the different process parameter sets

Parameter set	Weak penetration	Optimal	Balling	Keyhole
Depth (µm)	40	75	76	259
Variation from experimental (%)	72.4	-0.9	13.9	70.4
Width (µm)	80	115	126	179
Variation from experimental (%)	8,3	-2.5	18.9	-8.2



#### 3.2.4 Keyhole parameter set melt pool

When comparing the melt pool for this parameter set, it is clear that the geometry obtained from the simulation is not close to the experimental results. The width of the melt pool surface only varied by eight per cent from the values obtained experimentally. This is quite an encouraging result. When comparing the depth of the melt pool, the obtained result differed from the experimental mean by 70.4%. However, this value still falls well within two standard deviations of the experimentally obtained results. This is due to the large variation in the melt pool depth due to the instability of the keyhole melting process (Refer to Table III). Although deep penetration into the substrate occurs, which is a characteristic of keyhole mode melting, the melt pool geometry is not representative of the experimental results. It should be noted that no keyhole simulation studies could be found at the time of writing where this melt pool cross-section shape was obtained. This is being investigated further.

# **3.3** Effect of unknown and uncertain simulation parameters on melt pool geometry

During this investigation, it was noted that there were a few key simulation input parameters that were unclear at this stage. Most notably, these include the accommodation coefficient and evaporation pressure coefficients derived from the Clausius Clapeyron equation, and the temperature-dependent absorption of the material. These variables affect the rate of evaporation of the metal, the local pressure above the melt pool caused by evaporation and the energy absorbed by the substrate, respectively. The main reasons for the discrepancies in the results are thought to be the unknown evaporation pressure coefficients and temperaturedependent absorption (Communications, 2021). A study by Ye, et al. (2019) demonstrated that the absorptivity of a Ti6Al4V baseplate without powder can vary between 0.26 and 0.66 depending on the laser power applied to the bed. This finding is significant and needs to be accounted for to ensure accurate simulation in the future. Secondly, there was no known methodology to determine the evaporation pressure coefficients for simulations and it was not clear whether these values changed when different laser power and scanning speeds were applied. Accurate use of evaporation pressure coefficients and absorptivity is critical, as additional simulation work indicated that melt pool dimensions can vary widely by changing only these parameters.

## 4. Conclusions

This study proved that accurate results can be obtained for certain process parameters using numerical modelling. However, the study also clearly demonstrated that accurate results for all process parameters could not be obtained by using an identical simulation setup, verified simulation method, and verified material properties. These inaccuracies currently hinder the possibility of using the software to accurately determine process parameters. The reasons for these inaccuracies are still under investigation. However, certain simulation parameters have been identified that are both unknown, and notably affect melt pool dimensions. A deeper understanding of these parameters, as well as how they can be determined accurately, is required. While the process of characterising melt pool geometry through simulation holds merit, more research is required before its full benefit could be derived.

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