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Analytical-method for polishing-surface prediction considering tool flexibility and grain-material interaction

5 Amaia LEGIDO PÉREZ DE SAN ROMÁN ^{a,b}, Julien CHAVES-JACOB ^a, Jean-Marc LINARES ^a, Pedro José ARRAZOLA ^b

^a Aix Marseille Univ, CNRS, ISM, Inst Movement Sci, Marseille, France

^b Escuela Politécnica Superior de Mondragón Unibertsitatea, Loramendi 4, 20500 Mondragón, Spain

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Abstract

To improve surface roughness of components without altering geometrical dimensions, polishing operations are commonly used. These operations may be time-consuming and expensive. Thus, the prediction of polished surface quality is a key issue
15 to reduce the cost of these operations. Resulting surface quality of polishing operations is the outcome of a large number of local cutting phenomena (grains-material). The control of the multi-scale physical phenomena is a challenge when it comes to simulate these operations. In this paper, an Analytical-Method for Polishing-Surface Prediction (AMPSP) that considers the tool flexibility and each grain-material interaction is
20 proposed. The objective is to predict accurately the polished surface topography and the material removal rate, and to keep the history of all the local cutting phenomena, in order to define a digital twin of polishing operation. Experimental validation demonstrates that this AMPSP predicts the material removal rate (less than 35% of error) and the surface topology (less than a few percent). AMPSP will enable engineers
25 to quickly and accurately predict the polished topology obtained with 5-axis toolpaths.

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Keywords: Finishing; Polishing; Modelling; Digital Twin.

List of symbols and units

	a	length of major semi-axis of the contact ellipse (mm)
	a_N	nominal length of major semi-axis of the contact ellipse (mm)
	b	length of minor semi-axis of the contact ellipse (mm)
35	b_N	nominal length of minor semi-axis of the contact ellipse (mm)
	D_{max}	maximum diameter of the abrasive grains (mm)
	D_{mean}	mean diameter of the abrasive grains (mm)
	D_i	diameter of grain i (mm)
	E^*	contact modulus of the abrasive grain and the workpiece (MPa)
40	E_T	macroscopic Young's modulus of the tool (MPa)
	E_W	Young's modulus of the workpiece (MPa)
	F_a	mean force applied to each grain (N)
	F_i	force applied to the grain i (N)
	F_m	force between tool and workpiece (N)
45	F_N	applied nominal force (N)
	h_i	micro depth of cut of grain i (mm)
	HB	Brinell hardness number (N/mm ²)
	i	abrasive grain
	K_p	Preston coefficient
50	m	toolpath position
	N	spindle rotation speed (rpm)
	N_{grain}	standard grit number ISO 6344-1, 2 and 3 (<i>Tissier (2001)</i>)
	N_N	number of grains in a nominal contact ellipse
	P	pressure or force per unit area (MPa)
55	P_0	Hertzian mean pressure (MPa)
	P_{bi}	height of lowest point of grain i (mm)
	P_i	pressure applied to i grain (MPa)
	Q	material removal rate (mm ³ /min)
	r	minor torus radius (mm)
60	R	major torus radius (mm)
	v	relative velocity between the tool and the surface (mm/min)
	v_T	macroscopic Poisson's ratio of the tool
	v_W	Poisson's ratio of the workpiece
	V_f	feed rate (mm/min)
65	X_i, Y_i	coordinates of a grain i in the local contact framework (mm)
	Z_{xy}	height of surface point (mm)
	<i>Greek letters</i>	
70	$\Delta\omega$	spindle rotation angle discretization (°)
	θ_1, θ_2	tilt and lead angles (°)
	Δt	discretised time (s)
	σ	standard deviation of grain diameters (mm)
75	ω	spindle rotation angle (°)
	δ_N	nominal tool offset (mm)
	δ_i	depth of cut of each grain i (mm)
	δ_{imax}	maximum depth of cut (mm)

80 1) Introduction

Polishing is a common finishing stage of advanced workpieces. Polishing operations, unlike grinding, use flexible tools and are employed to improve surface roughness without significantly modifying surface form. Thus, the control and prediction of the obtained surface roughness is an important issue for polishing operations. To predict and control the surface quality of polishing operations, a method has been developed in this work, to define a digital twin of polishing operation. In literature, numerous grinding and polishing digital twins have been implemented. Polishing and grinding are complex operations that involve multi-scale phenomena. The result of these processes is the outcome of a multitude of local cutting phenomena performed by grains upon material. In order to devise digital twins of finishing operations (polishing or grinding), three types of methods can be distinguished in literature:

- Empirical methods: they are based on a mathematical model best-fitted by experimental results. For example, Moumen *et al.* (2016) study flank polishing operations with abrasive caps using a 5-axes milling machine. By means of experimental testing, they obtain an empirical model able to reduce polishing cost by optimising polishing parameters (spindle speed, feed rate, toolpath shape, radial engagement, and grain size). Zhao *et al.* (2014) perform single-factor belt polishing experiments to build an empirical model in order to predict surface roughness and obtain optimum process parameters in belt polishing (abrasive size, contact force, belt linear velocity and feed rate). The use of empirical models is restricted to the interpolation of the tested parameters for the process in question.

- Semi-empirical methods: they use an elementary experimental result that is analytically moved along a toolpath to predict the overall result. These methods use experiments to determine the behavioural law of a physical phenomenon and then, implement this law in a digital twin. For instance, Cheung *et al.* (2011) experimentally characterise the Tool Influence Function (TIF), which is the footprint left by a tool during a punctual contact with a surface. Thereafter, following their method, this TIF is displaced along a toolpath to virtually cut the surface and thus predict the obtained surface topology. Lu *et al.* (2019) used a TIF experiment in dual-axis wheel polishing to derive the roughness topology in cases where either a single or dual-axis process was used. Thereafter, the obtained result was translated along the toolpath to predict the final surface topology. Through this methodology, they concluded that dual-axis polishing homogenises and reduces the generated surface roughness compared to single-axis wheel polishing.

- Analytical methods: they try to model the physical phenomena. These methods model cutting phenomena. Although they require more computing time, these kinds of models have more extensive applicability than empirical or semi-empirical models. Results can be transferred to more than one polishing condition.

Among all grinding and polishing models described in literature, three levels of analytical models are differentiated (see Fig. 1):

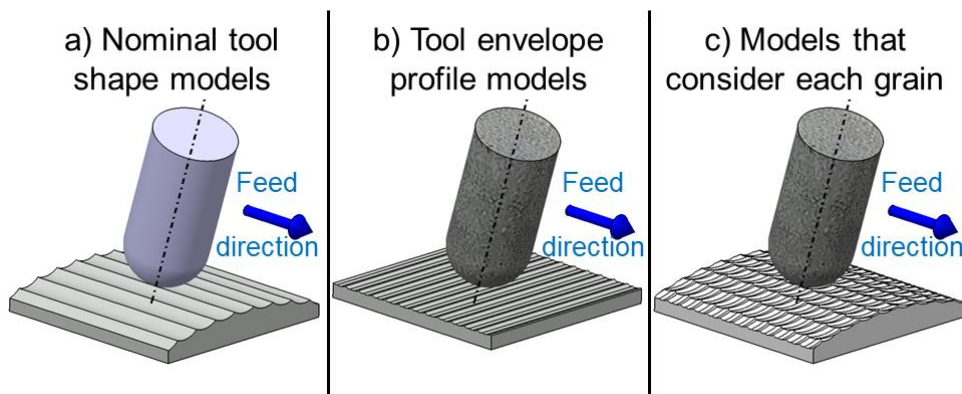
- **Nominal tool shape models:** the nominal geometry of the tool leaves its shape's footprint on the surface. These models are useful only for grinding methods since tool grains are not considered. Denkena *et al.* (2010) develop this kind of model in order to predict surface roughness in 5-axes grinding operations. Zhang *et al.* (2020) used this type of model to predict the profile roughness of internal tubes finished by a novel magnetic polishing process.

- 130 • **Tool envelope profile models:** a 2D profile is defined to simulate the tool envelope considering the tool grains. This profile is frequently used to predict surface roughness. This type of model is often more accurate than nominal tool shape methods in predicting surface roughness but is not suitable for waviness and form aspects. *Li et al. (2020)* develop this kind of model to predict the surface morphologies and surface roughness in the grinding of gallium gadolinium garnet single crystals. In order to achieve a more realistic and accurate determination of the tool envelope profile, *Wu et al. (2020)* propose to measure this profile while the tool is mounted on the machine. This measure is carried out using a laser displacement sensor system. *Uhlmann et al. (2016)* consider this type of model to predict surface roughness in 4-axes grinding processes.

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- 145 • **Models that consider each grain:** cutting phenomena of every tool grain are considered. All these local phenomena are added to get macroscopic results of surface roughness. *Darafon et al. (2013)* carry out a method that takes each grain into account and determines uncut chip thickness, instantaneous material removal rate, contact length and surface roughness. They consider a spherical geometry of grains and a stochastic abrasive wheel model. *Agarwal and Khare (2004)* develop the same kind of digital twin considering spherical grains and a random distribution of grain protrusion heights. They obtain a relationship between surface roughness and chip thickness. In the grinding method proposed by *Zhou et al. (2002)*, wear is taken into account by truncation of a Gaussian distribution model. The wheel wear volume is related to the change in the mean value of protrusion heights

150



155 Fig. 1. Three types of analytical model to predict the roughness

Polishing digital twins are uncommon in literature and numerous papers propose methods without taking into account tool flexibility. Accounting for vibration is a first step in this direction. *Jamshidi et al. (2019)* take into account the vibrations in a model that considers each grain to identify the active ones on a wheel. Thereafter, this model was completed to predict the grinding forces (*Jamshidi et al. (2020)*). *Xi and Zhou (2005)* develop a method considering each grain to predict surface roughness, based on the micro depth of cut calculated by Brinell theory. They determine the number of contact grains and the Gaussian probability function of grains in the tool from a given stone grit number. *Wang et al. (2009)* consider Hertzian pressure at the contact site between tool and workpiece in their method which calculates material removal values

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of all abrasive grains that cut. The elastic deformation of the tools is considered in the polishing method developed by *Qi et al. (2016)*, but this study is limited to the prediction of the material removal rate and does not include surface roughness determination. It follows Hertz elasticity contact theory and Brinell plasticity theory, and it assumes conic abrasive grain shape with spherical tip.

To conclude this overview from literature, there are very few models that consider each grain and tool flexibility, and none of them predicts the surface roughness. Thus, this paper presents an Analytical-Method for Polishing-Surface Prediction (AMPSP) that considers the tool flexibility, the initial surface and each grain-material interaction. The objective is to predict accurately the surface topography (to be able to identify all roughness criteria), the material removal rate, and to keep a history of all local interactions. In section 2, the AMPSP is presented and each computation step is detailed. In section 3, experimental tests are presented, to validate the mechanical model of tool/surface contact, the predicted material removal rate as well as the predicted surface topology.

2) AMPSP: Analytical-Method for Polishing-Surface Prediction

The main objective of the AMPSP is to predict analytically the topography of the resulting surface for polishing operations. Inputs of the AMPSP are presented in Fig. 2:

- Tool:
 - Nominal tool shape geometry.
 - Grain size distribution.
 - Macroscopic mechanical properties of the tool: Young's modulus E_T and Poisson ratio ν_T .(see section 3.1)
- Workpiece:
 - Mechanical properties of the workpiece: Young's modulus E_W and Poisson ratio ν_W .
 - Initial surface topography of the workpiece.
- Process parameters:
 - Toolpath.
 - Spindle and feed velocities, respectively: N and V_f .
 - Nominal tool offset. Usually this data is a common cutting parameter in polishing CAM software. This value is only used to help the operator estimate the suitable discretization step of the surface (see section 2.2). However, the real tool offset is calculated during the AMPSP process at each time step (see section 2.4.3).

Using provided inputs, the AMPSP carries out a simulation and gives as output the predicted surface topography of the polished surface. In Fig. 2, inputs and outputs of the AMPSP are presented schematically. Fig. 3 summarises the successive processing steps in the polishing method. The upper part of this chart shows the different inputs and their distribution through the different computational steps. The processes outlined in this flowchart are detailed in the subsequent sections.

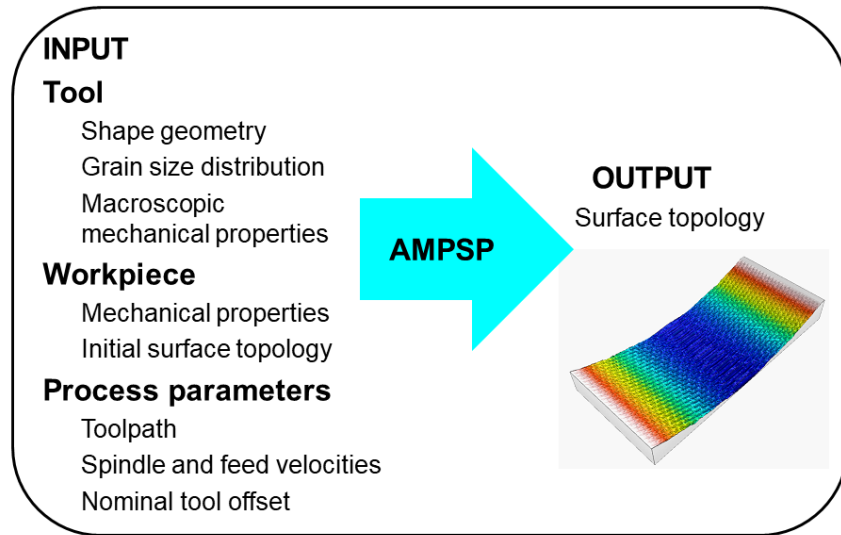


Fig. 2. Inputs and outputs of the AMPSP.

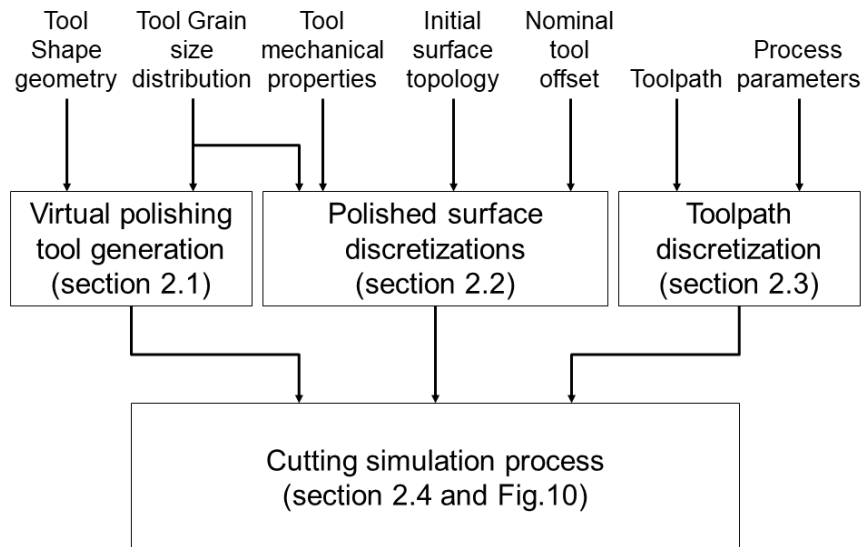


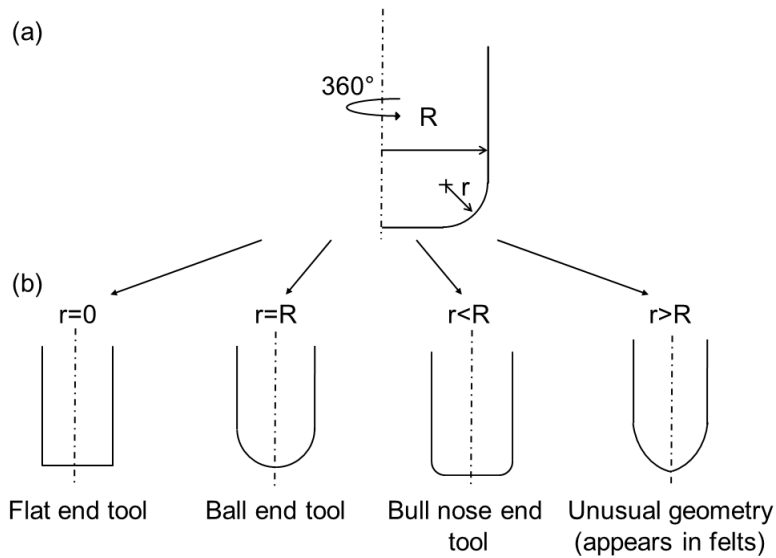
Fig. 3. Flowchart of the proposed method

2.1) Virtual polishing tool generation

215 In AMPSP, the tool is defined by a cloud of grains which may be distorted by the tool-surface contact. In this section, the steps used to generate the virtual tool are detailed. Firstly, the nominal tool shape is defined. Secondly, models are proposed to determine the grain distribution in relation to the standard grit number. Finally, the virtual tool is built by randomly positioning grains on the tool nominal surface.

2.1.1) Nominal tool shape generation

225 The first step for the tool shape generation is to define a tool profile that will be rotated around its axis. For that, a generic tool profile is proposed. Its geometry depends on two parameters, the major torus radius R and the minor torus radius r (Fig. 4 a). These parameters allow the generation of main tool geometries (Fig. 4 b).



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Fig. 4. (a) Proposed generic tool profile, (b) main tool geometries.

2.1.2) Grain size model based on standard grit number (*Tissier (2001)*)

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Once a tool profile geometry has been established, the tool's grain population and shape have to be defined. *Malkin et al. (2008)* demonstrate that the grain shape has a great influence on the surface roughness of the workpiece. In AMPSP method, the surface is discretised using a grid, which size is close to the footprint of a grain (see section 2.2). This discretization limits the effects of the chosen grain shape on the results. Thus, in the present work, the abrasive grains of the tool are modelled as spheres. With this model, the tool grains are defined by the position of their centre and their diameter.

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In addition, to specify the grain diameter population on a tool, a model was proposed. It is based on standard grit numbers ISO 6344-1, 2 and 3 (*Tissier (2001)*). For each grit number, standards provide percentages of grain population that remain in several sieve sizes. Many grinding and polishing digital twins developed in literature use Gaussian distribution to model tool grain population (*Xi and Zhou (2005)*, *Wang et al. (2009)*). Here, this kind of distribution is also employed. The grain distribution, associated to the grit number, is defined by (Fig. 5):

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- D_{mean} . It represents the mean diameter of the grains for each given grit number, as defined by standard.
- σ . It represents the standard deviation of Gaussian model in grains' diameter.
- D_{max} . It is the maximum grain diameter defined by standard. It is used to truncate the Gaussian grain distribution which is nominally infinite.

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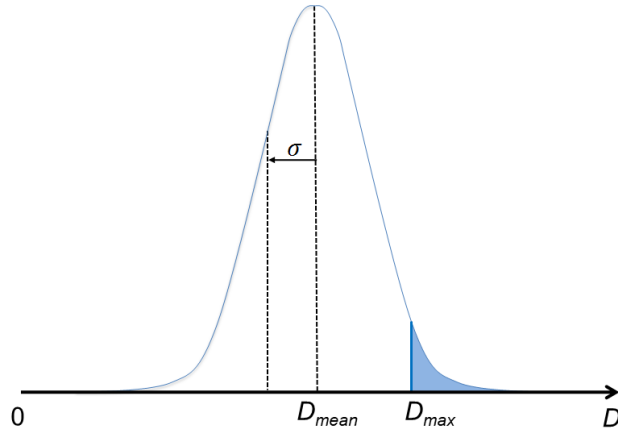


Fig. 5. Gaussian model of grain size distribution.

260 Eqs. 1-6 are proposed as best-fit values of D_{max} , D_{mean} and σ , provided by the standards (*Tissier (2001)*). The proposed model takes into account grit numbers for macro-grains from #60 to #220 and micro-grains from #240 to #1200. The equations of the model were obtained by best fitting the values given by standard. For each standard grit number N_{grain} , values of D_{max} , D_{mean} and σ can be obtained for both macro-grains (Eqs. 1-3) and micro-grains (Eqs. 4-6). To illustrate the relevance of proposed best-fit, the curve defined by eq. 5 (grey line) and the values defined by the standard ISO6344-1,2,3 (black spots) are plotted in Fig. 6. This figure highlights a suitability of the proposed model.

$$D_{max} = \frac{15}{N_{grain}} + 4.67 \times 10^{-2}(\text{mm}) \quad (1)$$

$$270 \quad D_{mean} = \frac{13}{N_{grain}} + 4.6 \times 10^{-3} (\text{mm}) \quad (2)$$

$$\sigma = \frac{1.5}{N_{grain}} + 4.4 \times 10^{-3}(\text{mm}) \quad (3)$$

$$D_{max} = \frac{31}{N_{grain}} - 2.6 \times 10^{-2}(\text{mm}) \quad (4)$$

$$D_{mean} = \frac{17}{N_{grain}} - 1.5 \times 10^{-2} (\text{mm}) \quad (5)$$

$$\sigma = \frac{1.5}{N_{grain}} - 1.2 \times 10^{-3} (\text{mm}) \quad (6)$$

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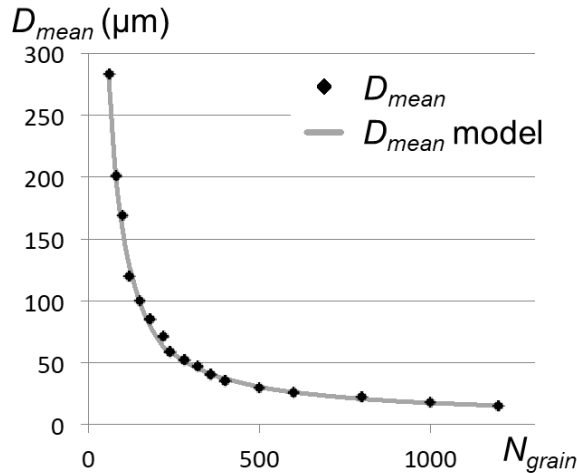


Fig. 6. D_{mean} defined by standard grit number (*Tissier (2001)*) and proposed interpolation.

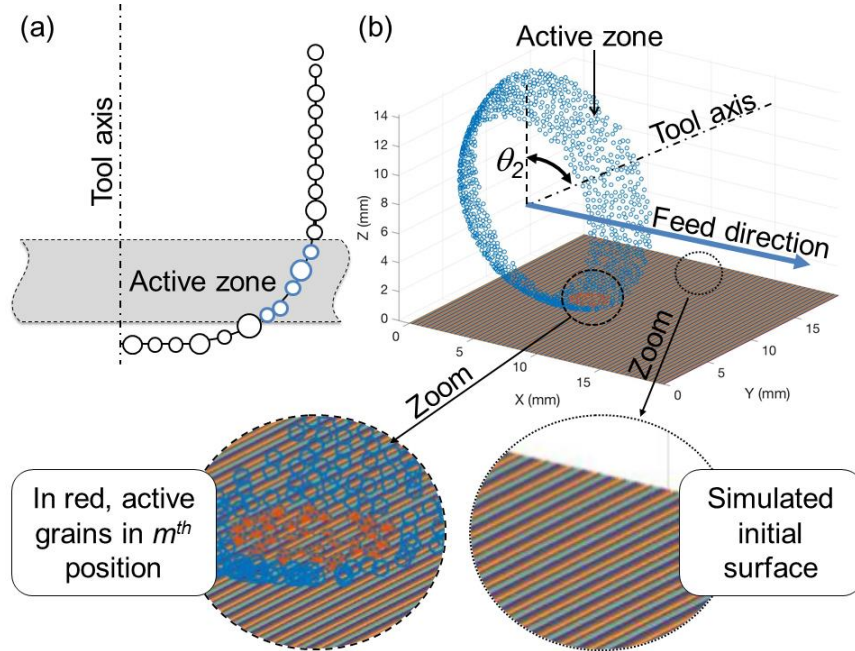
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2.1.3) Tool generation

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As seen previously, grains are modelled as spheres. A revolution of the tool nominal profile is performed to obtain the tool nominal surface. Thereafter, as proposed by *Xi and Zhou (2005)*, the centres of the grains are aligned along the nominal tool. Thus, the centres of the spheres which model the grains are positioned on the tool nominal surface and a grain diameter (D_i) is randomly selected using the distribution as defined in section 2.1.2. Each centre is randomly positioned, and a 3D collision test is carried out with all previously positioned grains. If no collision is detected, the position of the grain is validated, otherwise, the grain is removed. In this model, only the active grains are represented. The density of these grains depends on the tool type. *Chaves-Jacob et al. (2021)* propose an experimental procedure to determine this density. Fig. 7 represents an example of a tool generated in the AMPSP. In the AMPSP, only the cutting zone of the tool is generated in order to reduce computing time.



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Fig. 7. (a) Grain placement in tool profile, (b) generated tool example in 3D.

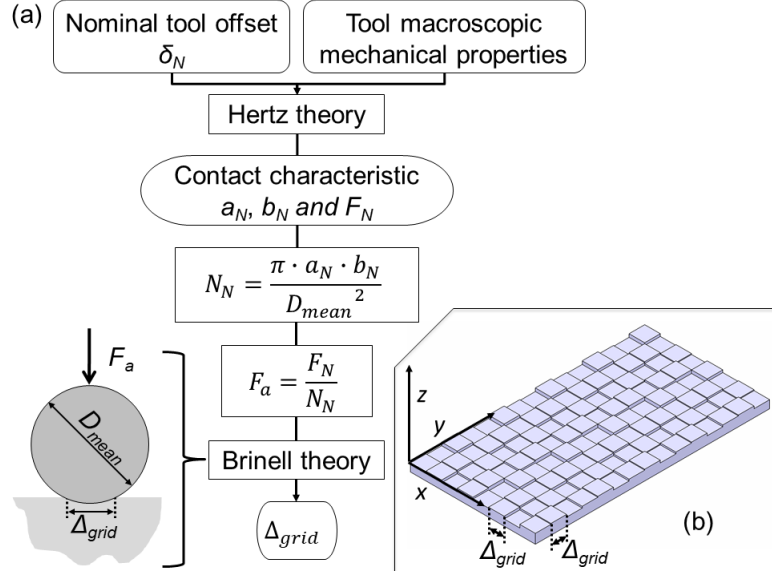
2.2) Polished surface discretization

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In the AMPSP, the surface of the workpiece is discretised using a grid (see Fig. 8 which defines the flowchart used to determine the grid size). The definition of the grid size is a trade-off between the predicted surface accuracy and the computing time. Here, the grid size, Δ_{grid} , is selected to be equal to the average footprint computed using Brinell theory with a mean grain diameter, D_{mean} , and an average force applied on this grain, F_a . (Fig. 8 a). The average force is estimated using the applied nominal force F_N , determined with the nominal tool offset and Hertz theory. In order to compute the average force, the number of grains is estimated by dividing the nominal contact surface area (computed with Hertz theory) by the surface occupied by an average grain. Brinell theory is applied in Eq. 7 to estimate the mean grain's micro depth of cut and the grid size respectively. Thus, as presented in Fig. 8 b, in AMPSP, the surface is modelled as a succession of horizontal tiles which all have an independent altitude Z_{xy} .

$$\Delta_{grid} = 4 \sqrt{\frac{D_{mean}^2}{2} - \left(\frac{D_{mean}}{2} - \frac{F_a}{D_{mean} \pi HB} \right)^2} \quad (7)$$



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Fig. 8. (a) flowchart to determine the surface discretization size; (b) surface discretization grid.

2.3) Toolpath discretization

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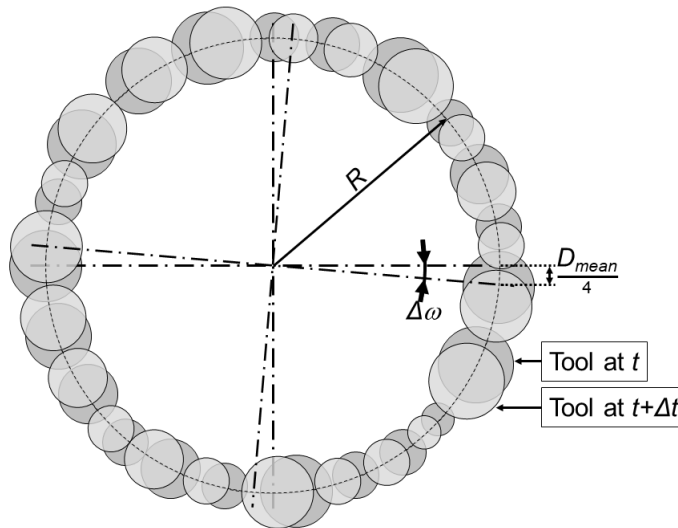
The AMPSP is developed for a 5-axes toolpath. Discretization of such a toolpath is required for the computation of the model. This discretization is once more a trade-off between predicted surface accuracy and computing time. In the proposed simulation, no interpolation is carried out between two simulated positions. To this end, the tool displacement between two simulated positions must be accurately controlled. To obtain

325 enough discretization, the peripheral displacement was fixed at a quarter of the value of the mean grain diameter, as illustrated in Fig. 9. This displacement corresponds to a spindle rotation of $\Delta\omega$. This angle was calculated considering the maximum tool radius R and the mean diameter of grains (Eq.8). This spindle rotation angle discretization $\Delta\omega$ induces a time discretization Δt that is calculated as a function of the spindle rotation

330 speed N (Eq. 9).

$$\Delta\omega = \tan^{-1} \left(\frac{D_{mean}}{4 \cdot R} \right) \quad (8)$$

$$\Delta t = \frac{\Delta\omega}{60N} \quad (9)$$



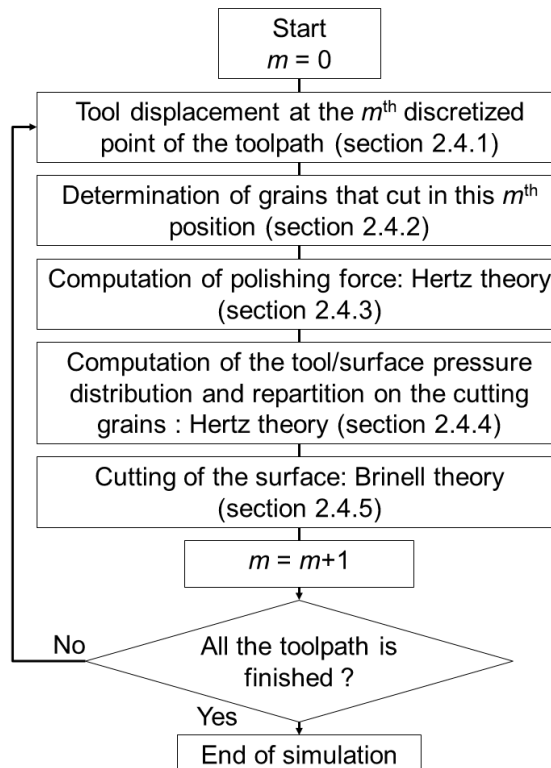
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Fig. 9. Tool rotation discretization model.

2.4) Cutting simulation process

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Fig. 10 illustrates the cutting simulation process and its steps at each discretised point of the toolpath. Cutting simulation ends when the sequence of discretised points of the toolpath reaches the last point. The following sections present the steps in detail.



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Fig. 10. Cutting simulation steps.

2.4.1) Tool displacement along the toolpath

Section 2.3 presented the determination of time discretization of the simulated
 350 polishing operation. The AMPSP is able to predict the surface topology obtained by a 5-
 axes toolpath. At each m^{th} discretised point of the simulation, the tool position and
 orientation are determined using a linear interpolation between the previous and next
 points of the simulated toolpath. In the simulation, the tool is modelled as a cloud of
 grains (see section 2.1.3). When the position of the tool and its axes orientation for the
 355 m^{th} point are determined, the cloud of points is translated and rotated according to this
 location. This operation allows the simulated tool to be set in the m^{th} position.

2.4.2) Determination of active grains in m^{th} position

In the AMPSP, the tool is considered a cloud of grains (see section 2.1.3). Once it is
 360 rotated and moved to the m^{th} position, the grains that contribute to the cutting process in
 this simulated position must be identified. For that purpose, the depth δ_i of the grain i in
 the simulated surface is calculated (Fig.11). Eq. 10 presents this computation. It is
 calculated by subtracting the height of the surface under the grain i Z_{xy} from the lowest
 365 point of this grain noted P_{bi} . In this way, when δ_i has a negative value, the grain will cut
 the surface at this simulated point (m^{th}) whereas when this value is positive, the grain
 will not touch the surface at this point.

$$\delta_i = P_{bi} - Z_{xy} \quad (10)$$

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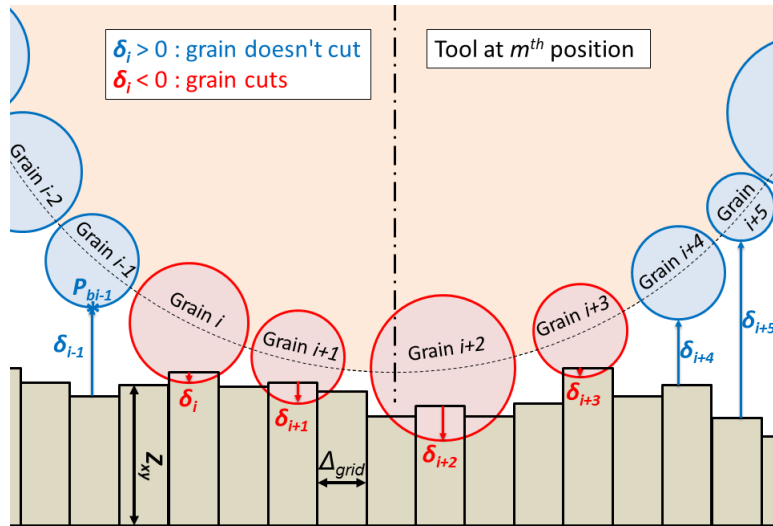


Fig. 11. Determination of grains that contribute to cutting process in simulated m^{th}
 point.

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2.4.3) Computation of polishing force in m^{th} position

There is a wide variety of technologies available in tool polishing processes.

Four main tool families can be defined:

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- Abrasive caps: sandpaper cap (covered by abrasives) is mounted on a flexible rubber support. These tools are used to remove wide volumes of material. They are generally used in rough polishing operations. However, these tools have a short lifetime.

- Flexible wheels: in this case, abrasive grains are embedded into a supple rubber matrix. These tools have a better surface compliance than abrasive caps, but with a lower material removal rate. They are usually used in semi-finishing polishing stages.

- Felts: loose abrasive grains are applied on a soft polishing felt. These tools have a very high surface compliance, but material removal rate is very low. These tools are limited to the finishing polishing stage.

- Shape Adaptive Grinding (SAG) proposed by *Beaucamp et al. (2015)*: this is a relatively new type of tool. Rigid pellets loaded with the abrasive grains are mounted on a flexible support. These tools have a high grinding ratio, and thus a long lifetime.

These four tool families involve different mechanical behaviours at the contact site between the tool and the polished surface due to the variety of physical phenomena. To take into account these differences in the AMPSP, different mechanical behaviours are considered. The Hertzian contact theory proposed by *Hertz (1881)* is used to model the elastic deformation of the tool (Fig.12). In this theory, the tool is considered a homogeneous material in the area of contact. Macroscopic Young modulus of tool E_T and macroscopic Poisson's modulus of tool ν_T have to be determined. An experiment is proposed in section 3.1 to determine these values.

To estimate the force F_m between the tool and the workpiece at the considered polishing time, the maximum value of depth δ_{imax} (computed in section 2.4.2) is used. The relation between force and depth is established with Hertz theory (Fig. 12).

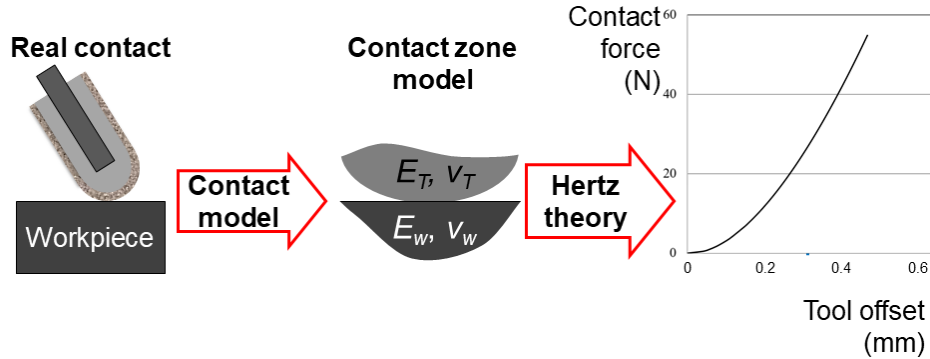


Fig. 12. Model of contact between the tool and the polished surface.

2.4.4) Computation of the tool/surface pressure distribution and repartition on the cutting grains

As proposed in Hertz theory, the contact ellipse between the tool and the workpiece is calculated using the contact force (F_m) determined in section 2.4.3 and the tool's mechanical properties (see section 3.1). The contact ellipse is defined by major and minor semi-axes noted a and b respectively. Finally, Hertzian pressure P_0 (Eq. 11) and pressure field between the tool and the surface (Eq. 12) are calculated. In eq. 12 X_i and Y_i represent the coordinates of the considered point in the local contact framework. This local contact framework is defined as follows: \mathbf{X} and \mathbf{Y} axes are respectively aligned with the minor and major semi-axis, the origin is centred on the contact area.

$$P_0 = \frac{1.5F_m}{\pi ab} \quad (11)$$

$$P_i = P_0 \sqrt{1 - \left(\frac{x_i^2}{b^2} + \frac{y_i^2}{a^2} \right)} \quad (12)$$

425 Fig. 13 illustrates the method to split the applied forces (see section 2.4.3) onto each active grain. The pressure field is used to define the proportion of F_m applied to each active grain. The pressure applied to each grain is derived from the position of this grain in the pressure field P_i (Eq. 12). Then, the proportion of F_m on each grain is computed into a proportional coefficient used to determine the effort applied on each grain (Eqs. 13-14).

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$$F_i = \frac{F_m \times P_i}{\sum_i P_i} \quad (13)$$

$$\sum_i F_i = F_m \quad (14)$$

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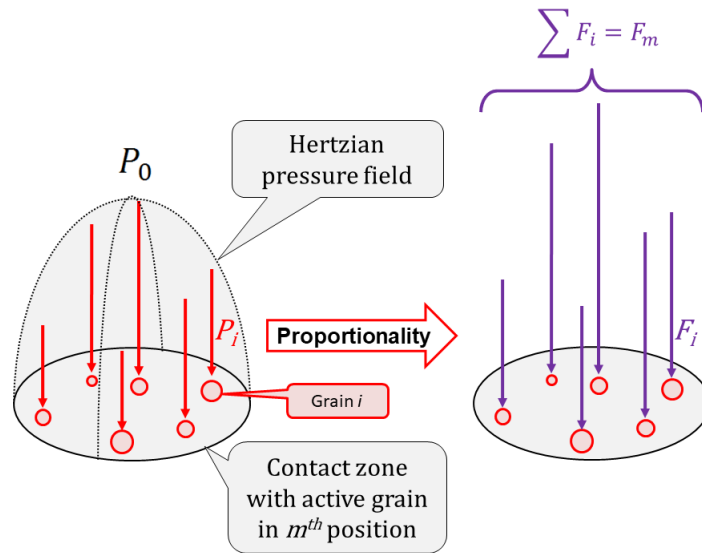


Fig. 13. Method to calculate the contact force on each active grain.

2.4.5) Surface cutting simulation

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Nguyen and butler (2005) simulated the interaction between the abrasive grain and the workpiece in grinding, considering the three cutting modes: cut, plough or rub. In this simulation the attack angle of grains is used to determine the adequate mode for each grain in the simulation. In a first approach, in AMPSP only the cutting mode is considered. In future work, the side ridges along the groove induced by rubbing mode, transferring material to neighbour cells of the discretised surface, could be taken into account. Thus, in this proposed method, the Brinell theory is used to consider the plastic deformation induced by the cutting phenomenon. The depth of cut of each grain inside the contact ellipse is obtained with that theory (Eq. 15). This depth varies depending on the force applied to each grain F_i (see section 2.4.4), its diameter D_i and Brinell hardness number HB. Fig. 14 illustrates the cutting simulation process. Thereafter, the depth of cut, h_i , of each active grain is subtracted from the tiles below the centre of each grain.

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$$h_i = \frac{F_i}{D_i \pi \text{HB}} \quad (15)$$

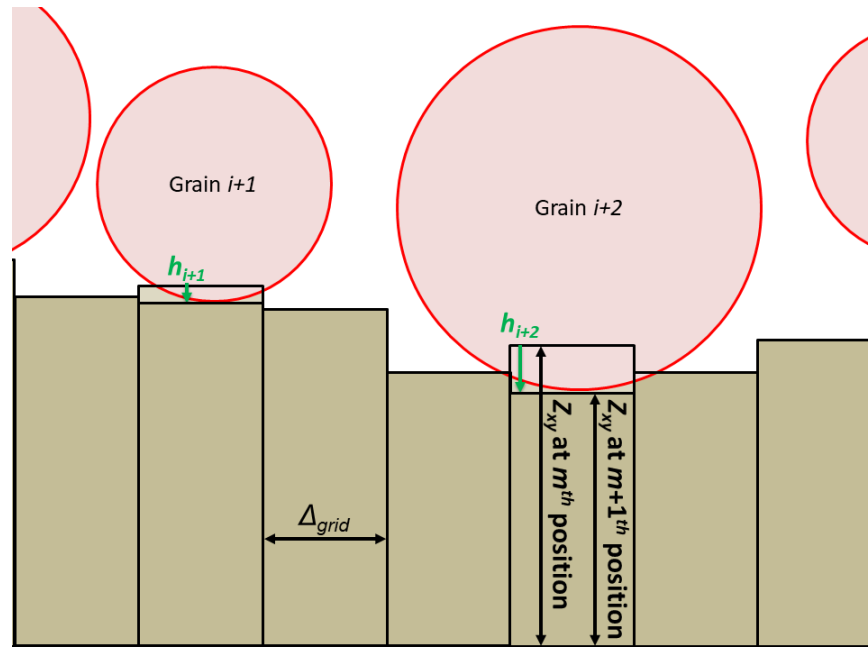


Fig. 14. Surface cutting model.

3) Experimental validation

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3.1) Experimental determination of method parameters

As presented in section 2.4.3, the contact between the tool and the polished workpiece is modelled using the Hertz theory. This theory requires the mechanical properties of both solids in contact: Young modulus and Poisson's ratio. For the workpiece, natural mechanical properties are used; for the tool, macroscopic mechanical properties are determined using an experiment. Thereby, in order to determine these two mechanical properties, an elementary sphere/plan experiment was conducted. The experiment consists in a plunge test without spindle rotation of the flexible tool on a metallic plane mounted on a dynamometer sensor (see Fig. 15). The tool is moved down in steps in the normal direction of the metallic plane for several steps. Between each step a timer is set without tool displacement to carry out force measurement. Thereafter, the experimental force/depth relation is established, and a Hertz model is best fitted. Thus, both optimised macroscopic mechanical parameters of the tool are determined.

In Fig. 15, the experiment was carried out with two different configurations:

- Fig. 15 a: a flexible wheel tool composed of abrasive grains embedded in a rubber matrix was set in a 3-axes configuration. Experiments were conducted with a grit number of #240 (average grain size of $46 \mu\text{m}$).

- Fig. 15 b: an abrasive cap composed of sandpaper with a resin bond and mounted on a rubber support, was used in a 5-axes configuration with an inclination of 50° . Experiments were conducted with a grit number of #60 (average grain size of $280 \mu\text{m}$).

Fig. 15 presents, for both tested tools, the force/depth curve obtained experimentally (points), and the Hertz best-fitted model (continuous line). Both experiments confirm that the Hertz theory is able to accurately model the complex contact between the tool and the polished workpiece within the depth range used in polishing (tool offset $< 0.4 \text{ mm}$).

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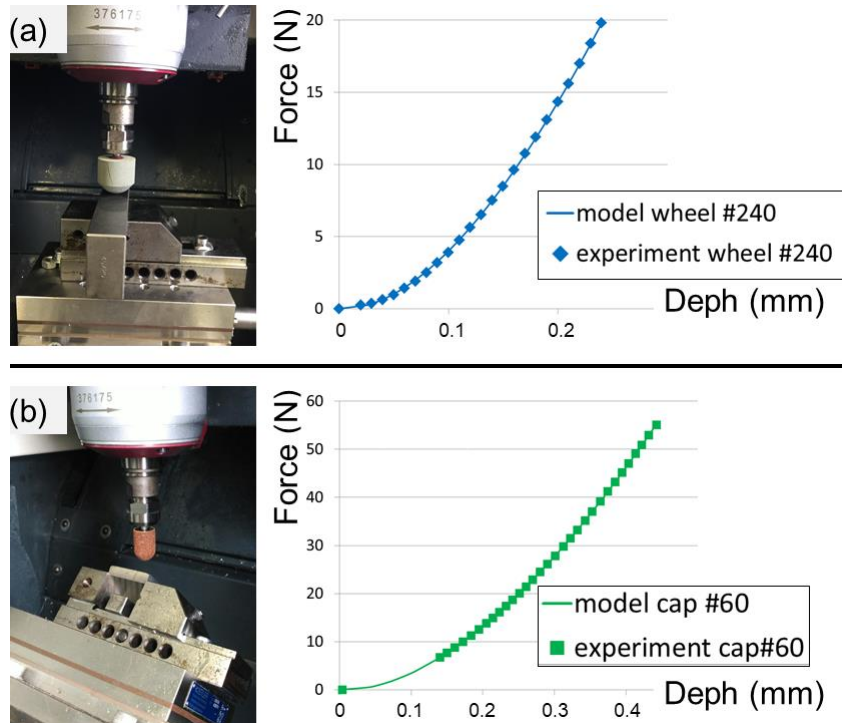


Fig. 15. Experimental determination of equivalent tool macroscopic mechanical parameters a) flexible wheel tool with a grit number of #240 in a 3-axes configuration
 490 b) abrasive cap with a grit number of #60 in a 5-axes configuration (inclination of 50°).

3.2) Validation of material removal rate: single polishing line test

To validate the material removal rate predicted by AMPSP a single polishing line test is carried out on a plane. This experiment consists in performing the polishing of a line on a plane (considered without defect). This experiment allows the depth of the realised groove to be measured accurately in relation to the reference plane surrounding it. 5 tests have been performed with different values of feed rate V_f , spindle rotation speed N and tilt angle θ_2 . Polishing parameters and experimental elementary lines are shown in Fig. 16. Lines are 40 mm long and the workpiece is made of stainless steel 17-4 ph (elastic limit around 1,100 Mpa). For all the experiments, the employed type of tool was an abrasive cap with a spherical head of 8.5 mm of radius and a grit number of #60. Experiments were carried out on a common 5-axes CNC milling machine. Used nominal tool offset was 0.3 mm. For simulation, the virtual polishing tool was generated with grains set to fully fit the surface of the tool in order to represent the sandpaper (number of active grain).

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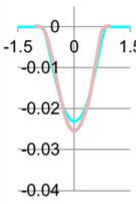
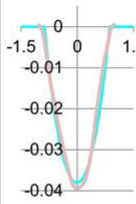
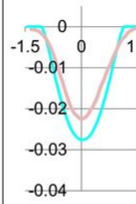
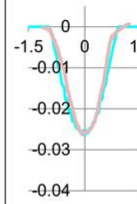
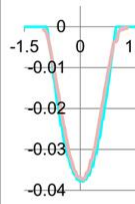
	#1	#2	#3	#4	#5
V_f (mm)	1,000	2,000	1,000	500	1,000
N (rpm)	12,000	12,000	6,000	12,000	12,000
θ_2 ($^\circ$)	45	45	20	20	20
Photo					
Average mean profile					
Abcissa (mm) Ordinate (mm)	-1.5 0 1.5 -0.01 -0.02 -0.03 -0.04	-1.5 0 1.5 -0.01 -0.02 -0.03 -0.04	-1.5 0 1.5 -0.01 -0.02 -0.03 -0.04	-1.5 0 1.5 -0.01 -0.02 -0.03 -0.04	-1.5 0 1.5 -0.01 -0.02 -0.03 -0.04
— AMPSP — Experiment					
Gap (%)	-3.14	1.23	33.32	10.17	9.42

Fig. 16. Five experimental line tests to validate the predicted material removal rate of AMPSP.

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Fig. 17 presents the measured surface topologies with experimental testing and the predicted surface topology by AMPSP for the line #1. Experimental surfaces have been characterised through optical measurements using a chromatic confocal sensor CMM. The resolution of this CMM is about 10 nm. The measurements were carried out on areas of 2.5 mm × 1 mm, centred on each line. The length of these areas was set perpendicularly to the direction of the lines and scanned in both directions with 4 μm intervals. Presented results highlight the quality of the predicted surface topologies. Furthermore, to evaluate the accuracy of the predicted material removal rate, a mean of the profiles, both measured and simulated, was computed along 1 mm of the groove. Both profiles are presented in Fig. 17 for the line #1 and in the average mean profile row of Fig. 16 for the 5 performed tests. The gap (%) row provides the percentage of gap between both lines. The result defines a qualitative projection of predicted material removal rate (gap < 35%).

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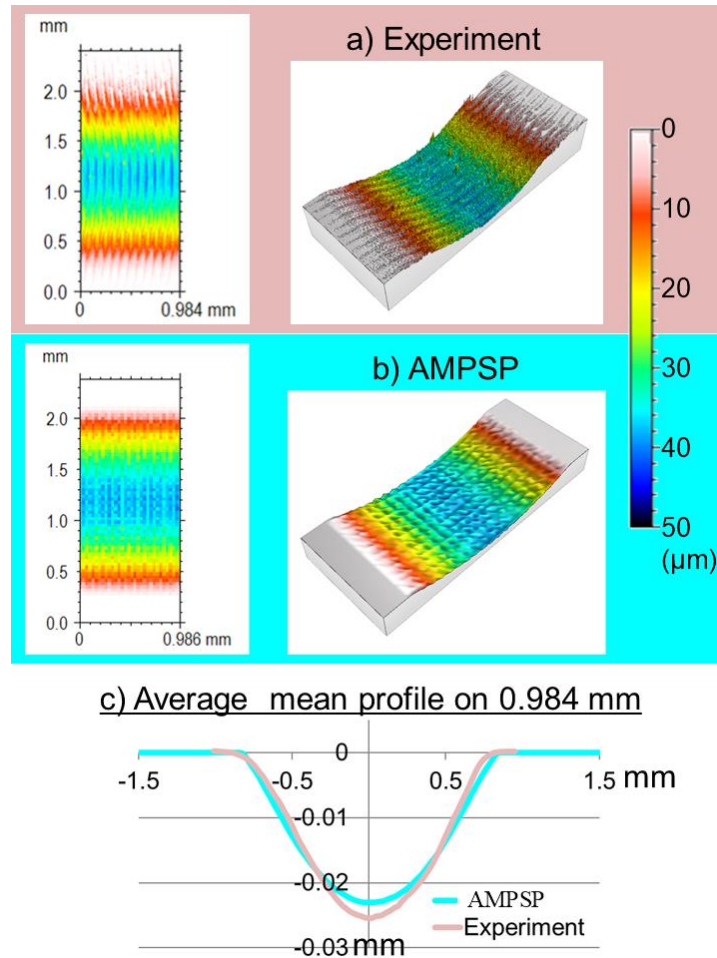


Fig. 17. a) Experimental #1 measure b) predicted #L1 measure c) mean profiles (predicted-blue line/ experimental-red line).

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3.3) Validation of obtained surface topology

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In order to validate the predicted topology of AMPSP, a surface polishing test was conducted. In this experiment, the surface of a workpiece made of stainless steel 17-4 ph and built by additive manufacturing was polished in the 5-axes CNC milling machine using a raster toolpath. Fig. 18 shows the workpiece before and after the polishing experiment. The polished part of this workpiece is composed of a succession of three planes of 10 mm length, linked by corner radius. The tool used was an abrasive cap with a spherical head of 8.5 mm radius and a grit number of #60 (average grain size of 280 μm). Polishing parameters were: a feed rate of 2000 mm/min, a spindle rotation speed of 12000 rpm, a tilt angle of 60°, a tool offset of 0.2 mm and a track spacing of 0.5 mm.

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The obtained surface was characterised through optical measurements using a chromatic confocal sensor CMM. As illustrated in Fig 19 and 20, the measured areas are three squares of 1 mm side located in the middle of the three polished planes. These three results are used to represent the experimental variability. Thereafter, a Gaussian filter with a cut-off of 0.25 mm was used to separate the waviness and the roughness signal of the polished surface (Fig. 19 and 20). Then, arithmetical mean height (S_a) was computed on each surface. In parallel, a simulation was carried out using AMPSP to represent the experimental polishing operation. The predicted surfaces undergo the same treatment as the experimental surfaces.

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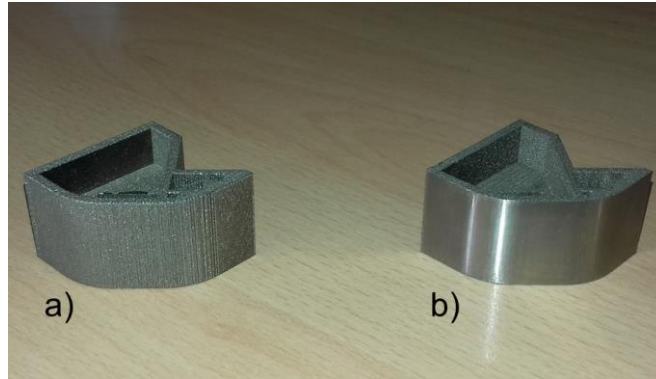


Fig. 18. Workpiece used to validate the obtained surface topology a) before polishing (rough of additive manufacturing) b) after polishing stage.

555 Waviness results (experimental and simulated) clearly highlight the successive parallel passes of the tool (from left to right in Fig. 19) separated by the values of lateral displacement between two passes (0.3 mm, in the height direction in Fig. 19). Experimental results present a small dispersion in surface topology, due to experimental variations (e.g., chatters or tool deflections). Apart from this mentioned variation, the predicted waviness is accurate in topology and values (gap between average values).

560 Roughness results (experimental and simulated) highlight scratches in the cutting speed direction (from top to bottom in Fig. 20). Furthermore, these results show a progressive reduction of the roughness as the polishing time passes (first machined surface, $S_a=0.53\mu\text{m}$, second $S_a=0.39\mu\text{m}$ and last $S_a=0.23\mu\text{m}$). This evolution may be explained by the fast tool wear of the abrasive caps. The tool wear directly modifies the grain distribution, as presented by Zhou and Xi (2002). The simulation performed by AMPSP (Fig. 20) was conducted with a digital “fresh tool” having a grain distribution as specified by the tool manufacturer. It seems logical that the performed simulation predicts a surface roughness close to the first polished surface ($S_a=0.53\mu\text{m}$), where the tool was the newest.

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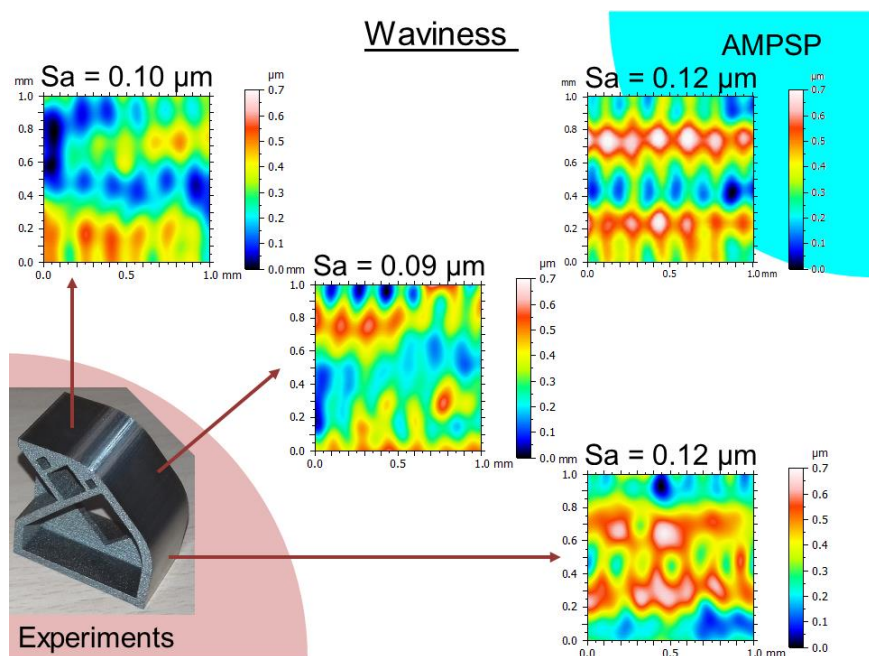


Fig. 19. Experimental and simulated surface waviness.

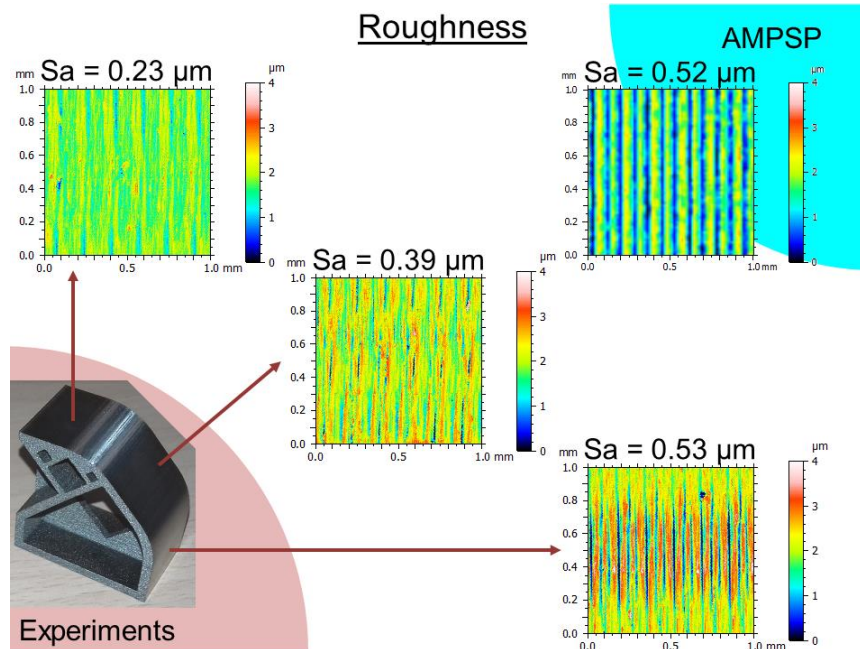


Fig. 20. Experimental and simulated surface roughness.

4). Conclusions

This work defines an Analytical-Method for Polishing-Surface Prediction (AMPSP) in order to control and predict results of polishing operations. The AMPSP is useful for 5-axes toolpaths and is a digital twin of polishing operation. It considers the cutting process of each grain of the polishing tool. In AMPSP, the tool is defined as a cloud of grains which may be distorted by the tool-surface contact, and the surface of the workpiece is discretised using a grid (which size is suitably selected). Local cutting phenomena are calculated with Brinell theory, and macroscopic contact between the tool and the surface is modelled with Hertz theory. The association of these two theories allows for a suitable multi-scale model of polishing phenomena. Experimental validation was conducted in three steps:

- Validation of the tool/surface contact model for two polishing tool-families.
- Validation of the material removal rate in five different polishing configurations (less than 35% of error).
- Validation of the predicted surface topology in waviness and roughness (less than a few percent).

One of the main limits of the AMPSP is the computation time (for example, the simulation of a 10 x 10 mm surface using a tool with an average grain of 280μm needs around 48h of computation on a common laptop). The time consumption cost of computation ought to be compared to the cost needed to obtain the same result experimentally. The ratio between both costs is in favour of the AMPSP. To conclude, AMPSP will enable engineers to predict the polished topology obtained with 5-axis toolpaths.

5) Acknowledgements

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