

MDPI

Article

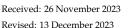
Evaluation of Machining Variables on Machinability of Nickel Alloy Inconel 718 Using Coated Carbide Tools

Muhammad Iftikhar Faraz 1,* and Jana Petru 20

- Department of Mechanical Engineering, College of Engineering, King Faisal University, Al-Ahsa 31982, Saudi Arabia
- Department of Machining, Assembly and Engineering Metrology, Mechanical Engineering Faculty, VŠB-Technical University of Ostrava, 17, Listopadu 2172/15, 708 00 Ostrava, Czech Republic
- * Correspondence: mfaraz@kfu.edu.sa

Abstract: The current work was undertaken with the research aim of experimental examination of tool wear, surface roughness and burr formation during the micro-milling of Inconel 718 using different coated tools. Inconel 718 is one of the most widely used materials for purpose-oriented utilization owing to its preferred mechanical and physical properties, including high strength and corrosion resistance. On the opposite end, the machining of Inconel 718 poses certain machinability challenges, which significantly elevates tool wear and subsequently surface roughness. Cutting speed, feed rate and depth of cut were selected as variable machining inputs. With reference to tool wear, all input variables were found to be significant, with tool coating having the highest contribution ratio of 36.19%. In case of surface roughness, cutting speed and tool coating were identified as effective input parameters with contribution ratios of 51.24% and 34.27%, respectively. Similarly, depth of cut proved to be an influential factor for burr height formation (in both up-milling and down-milling), whereas feed rate had the highest contribution ratios for burr width formation during up-milling and down-milling, i.e., 39.28% and 36.26%, respectively. Consequently, contour plots for output responses were drawn between significant parameters to analyze machinability. One of the vital research outcomes was the identification of a tool coating parameter that is significant for all four analyzed aspects of burr formation. In addition, regression equations were formulated for machining responses. The best- and worst-case scenarios for individual input parameters, as identified from main effects plots, were validated during confirmatory experimentation. Moreover, effects of input variables on output response were characterized using close-up imagery, and dominant wear mechanisms were also identified. The utility of the research is underlined by the optimization of the sustainability and productivity of the manufacturing process.

Keywords: Inconel 718; sustainable manufacturing; machinability; energy-efficient processes; coated tools; environmental footprint; process optimization for waste reduction; smart manufacturing; sustainability



Academic Editors: Ali Khalfallah

Accepted: 18 December 2023 Published: 20 December 2023

check for

updates

Citation: Faraz, M.I.; Petru, J. Evaluation of Machining Variables on

Machinability of Nickel Alloy Inconel

718 Using Coated Carbide Tools.

Machines **2024**, 12, 4. https://doi.org/10.3390/machines12010004



and Carlos Leitao

Copyright: © 2023 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https://creativecommons.org/licenses/by/4.0/).

1. Introduction

The rapid advancement of the technology of new production methods and materials has caused the swift progress of aviation sector in the last few decades. A prominent example is the redesigning of aircraft engines to be more lightweight as well as efficient with fuel expenses. One of the preferred materials worldwide, Inconel 718, has been extensively used because of its properties like corrosion resistance and high strength. Inconel 718 is extremely resistant to wear and corrosion owing to its composition of nickel, chromium and iron in addition to other elements. It is used to create blades, disks and other parts of aircraft engines [1–3]. Its corrosion resistance, high strength and high temperature creep resistance have also found utility in the manufacture of various components other than aviation engines. Various parts for gas turbines and rockets are also manufactured using Inconel 718. Electrochemical micromachining has also proved to be a productive

Machines **2024**, 12, 4 2 of 21

manufacturing process using Inconel 718. Additionally, Inconel 718 can be drilled using electrochemical machining for the manufacture of various micro-products. Nevertheless, the micromachining of Inconel 718 poses a challenge [4–7] due to its poor machinability. Resultantly, research is underway using various combinations of input variables to arrive at optimum machining conditions. Industrial component miniaturization with acceptable dimensional accuracy and a variety of functionalities is a widespread avenue of research. Using micromachining technology, small parts and components can be produced in large quantities. The technical definition of micromachining can be expressed in a variety of ways. This material removal technique produces tiny and accurate 3D objects with sizes between 1 nm and 0.99 nm [8–10]. Micro parts and components of all kinds have experienced a sharp rise in demand in recent years including for pharmaceutical micro pump delivery systems and inkjet print heads. The production of small parts necessitates precise tooling and highly reliable procedures with consistent repeatability.

Determining the rate of material removal in mechanical micromachining is a significant issue. The macro-machining industry has numerous physical and cutting mechanisms, for example, "size effect" phenomena. Due to their utilization in various industries, nickel alloys are one of the most researched materials available in the literature [11,12]. Numerous researchers have investigated various techniques for producing micro-components, including laser manufacturing, ion beam machining and ultrasonic photolithography [13– 16]. A variety of aids are used, such as different coatings, coolants, machining settings and laser-assisted machining, to pre-heat and soften the Inconel 718 workpiece [17–20]. Irfan et al. [21] investigated surface roughness and tool wear at 48 m/s cutting speed while micromachining Inconel 718. TiAlN + WC/C coatings and Diamond-Like Coatings (DLC) were found to exhibit lower tool wear and optimum built-up edge (BUE). The DLC-coated tool resulted in the lowest surface roughness after AlTiN- and TiAlN + WC/C-coated tools. According to Lu et al. [22], the input parameters of cutting speed and feed rate significantly influence surface roughness. An initial increase in surface roughness was accompanied by increases in cutting length. Aslantas et al. [23] found that tool coatings had an impact on cutting force, tool wear and machining quality while micro-milling Ti6Al4V alloy. Cutting force elevated with accelerated tool wear, reducing machining precision. Ozel et al. [24] found that the cBN-coated tool outperformed the uncoated one for micro-milling Ti6Al4V in terms of surface integrity and tool wear. Aramcharoen et al. [25] investigated various tool coatings during micro-milling tool steel. Coatings like TiN were found to be superior to TiAlN in terms of flank wear and edge chipping. In addition, it was noticed that TiAlN resulted in higher burr formation than the uncoated tool. It was demonstrated that cutting speed had the highest influence on surface roughness. In the micromachining of Inconel 718, minimum quality lubrication (MQL) has been found to be more sustainable and productive than dry machining, as concluded by Rahman et al. Due to the higher cutting zone temperature and increased area of tool/workpiece contact, tool wear under dry-cut operations is significantly greater than during wet-cut operations [26,27]. Researchers have also investigated input variables with the aim of improving the quality of the machined surface during micromachining. Attanasio et al. [28] investigated the influence of burr creation, tool wear and cutting forces on the final level of machining quality in terms of surface integrity. Zhanwen et al. [29] researched surface uniformity to enhance machining quality by examining input cutting variables including feed rate and spindle speed. Moreover, Aurich et al. [30] found that changing the spindle speed and tilt angle can reduce burr formation and improve surface roughness.

Several researchers have reportedly used a variety of tool coatings to analyze various machining responses during micro-milling techniques. However, tool wear has not been comprehensively analyzed using input cutting parameters in combination with tool coatings. The need for analysis of machinability, especially at conventional and transitional cutting speed ranges, has been identified as a vital research gap and forms the basis of the current work. The research motivation for the experimental investigation is also drawn from the fact that low-speed machining systems are more economical and widely available

Machines 2024, 12, 4 3 of 21

than high-speed machines. By investigating the effects of different machining parameters on surface roughness, burr formation, and tool wear, selected as output responses, the machining process can be made more efficient, productive and sustainable at the same time.

2. Research Motivation

Micromachining is used to make products whose precision is in the order of millimeters. At this scale, it is almost impossible to post-process the product in order to ensure quality and compatibility with the design parameters. An important challenge associated with micromachined parts includes deburring and surface improvement due to small features. Since deburring and post-processing are not viable options, identifying and selecting suitable parameters which reduce the intensity of burr formation as well as produce better surface quality are vital objectives in the case of micromachining. Similarly, tool wear is also a significant aspect as it is directly related to burr formation and surface integrity. Also, concerns related to sustainability require that tool wear should be minimized through careful selection of optimum parameters so that the wastage in terms of tool replacement can be minimized. These identified issues align well with the following sustainable development goals, as chalked out by the United Nations [31,32].

Goal 9: Industry, Innovation and Infrastructure emphasizes the need to foster sustainable industrialization and innovation. Achieving this goal requires a paradigm shift towards environmentally conscious manufacturing processes. Our research delves into the realm of sustainable machining, aiming to minimize the cost of tooling and post-processes. By seeking innovative ways to promote economic manufacturing, we contribute to more efficient and eco-friendly manufacturing processes.

Goal 12: Responsible Consumption and Production underscores the importance of responsible resource utilization and waste reduction. Sustainable machining of nickel alloy Inconel 718 is inherently aligned with this goal. Lowering tool wear not only reduces the carbon footprint of machining, by economical energy consumption, but also aligns with responsible resource consumption. Additionally, by enhancing the productivity of machining through optimized and superior surface finishes, we promote efficient use of resources.

3. Experimental Procedure

A comprehensive experimental design was formulated for the milling operation. Various aspects of the experimental methodology are discussed below.

3.1. Experimental Apparatus

A CNC milling machine, as shown in Figure 1, was used for the micro-milling testing of nickel-based super alloy Inconel 718. Initially, the surface of the workpiece was leveled using a carbide end mill with a 12 mm diameter. The surface was then used as a reference point throughout the design process. To ensure precise z-axis measurements, a tool presetter was employed. Table 1 contains adopted test parameters. Wedge-shaped tungsten carbide cutting tools with different coatings (nACo, AlTiN, TiSiN and uncoated), as shown in Figure 2, were utilized. These tools had a diameter of 0.06 inch (0.5 mm), whereas the average cutting edge radius of micro tools with nACo-, AlTiN- and TiSiN-coated cutting edges was 1.3 mm, 1.21 mm and 3.0 mm, respectively. The milling operation was carried out in a full-immersion machining environment.

The workpiece was prepared with dimensions of 146 mm, 10 mm and 22 mm using electrical discharge machining. To minimize tool wear and tear, experiments were carried out with a 10 mm slot in the cutting length. Slot spacing was kept at 2 mm. Figure 3 displays the schematics of the milling operation, where Figure 3a shows the top view of the workpiece with intended milling pathways, and Figure 3b depicts the side view of the milling slot. The material was initially ground and polished. The waterless itching treatment from Kalling was applied for 5 s. The average grain size was found to be 23.4 μ m using an Olympus DXS1000 digital microscope (Olympus Corporation, Tokyo, Japan) and

Machines **2024**, 12, 4 4 of 21

ASTM standard procedure. Vickers hardness of Inconel 718 was found to be 361 HV using a Vickers micro hardness tester.



Figure 1. CNC milling machine—model PARPAS PHS-680.

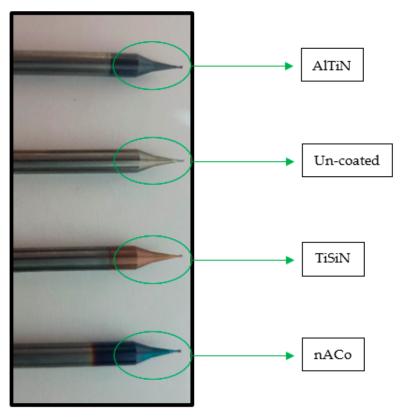


Figure 2. Tools used for experimentation.

Machines **2024**, 12, 4 5 of 21

Table 1. Experimental setup.

Milling Machine	CNC—PARPAS PHS-680
Workpiece material	Inconel 718
Number of flutes	2
Cutting length	10 mm

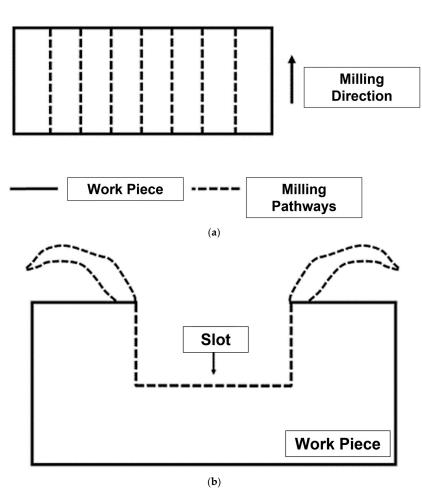


Figure 3. Workpiece: (a) top view with milling pathways indicated; (b) side view depicting milling slot (not to scale).

3.2. Design of Experiment

Input parameters of cutting speed (V_c), feed rate (f_z), depth of cut (a_p) and tool coating (t_c) were selected to analyze machining responses. These input parameters were employed due to their influence on surface roughness, tool wear and burr formation [33–36]. The ranges of input parameters, as shown in Table 2, were determined from the literature [37,38]. An L16 orthogonal array was formulated based on a Taguchi design of experimentation, as shown in Table 3. A Taguchi design of experiment is a robust substitute for full factorial experimentation. It produces valid and conclusive results with a lesser number of experimental runs. Various past researchers have used it to draw meaningful conclusions, as available in the literature. Each of the sixteen experiments was run three times to ensure repeatability [39]. A new insert was used for each of the 48 experimental sets. This was necessary to determine the specific tool wear corresponding to each set of machining combinations.

Machines **2024**, 12, 4 6 of 21

Table 2. Design of experiment.

Variables	Cutting Speed	Feed Rate	Depth of Cut	Coatings
Symbol	V_c	f_z	ap	t_c
Unit	m/min	μm/tooth	μm	-
Level 1	9	0.25	30	nACo
Level 2	14	0.50	50	AlTiN
Level 3	19	0.75	70	TiSiN
Level 4	24	1.00	90	uncoated

Table 3. Taguchi L16 array.

n.		Input Par	ameters	
Run	V _c (m/min)	f _z (μm/tooth)	a _p (μm)	t _c
1	9	0.25	30	nACo
2	9	0.5	50	AlTiN
3	9	0.75	70	TiSiN
4	9	1	90	uncoated
5	14	0.25	50	TiSiN
6	14	0.5	30	uncoated
7	14	0.75	90	nACo
8	14	1	70	AlTiN
9	19	0.25	70	uncoated
10	19	0.5	90	TiSiN
11	19	0.75	30	AlTiN
12	19	1	50	nACo
13	24	0.25	90	AlTiN
14	24	0.5	70	nACo
15	24	0.75	50	uncoated
16	24	1	30	TiSiN

3.3. Response Measurement

Tool wear, surface roughness and burr measurement were measured and analyzed during the conduct of the research. An Olympus DXS1000 digital microscope was used to quantify the tool wear response. One such insert is displayed in Figure 4, depicting tool wear. In addition, the Olympus DXS1000 digital microscope was utilized to examine the surface roughness (refer to Figure 5) owing to its capability to ascertain micro-surface roughness with high precision. Surface roughness for each machining condition was measured at 5 separate locations, and average value was determined to ensure accuracy. ISO 21920-2:2021 [40] was referred to for specification of surface roughness. With reference to burr analysis, there are various points on a machined surface where burrs can develop, including top, bottom, entrance and exit burrs, as shown in Figure 6. In the current study, the top and down burr height and width were the main areas of interest during burr analysis. Measurements of burr width and burr height were made, as depicted in Figure 7, using the DXS1000 digital microscope. Each machining combination was conducted three times to measure output response. The output responses were then used to analyze various aspects of the machinability of Inconel 718, including the effects of input parameters, as well as identification of influential machining parameters in terms of contribution ratio. The complete research methodology of the current work is displayed in Figure 8.

Machines **2024**, 12, 4 7 of 21

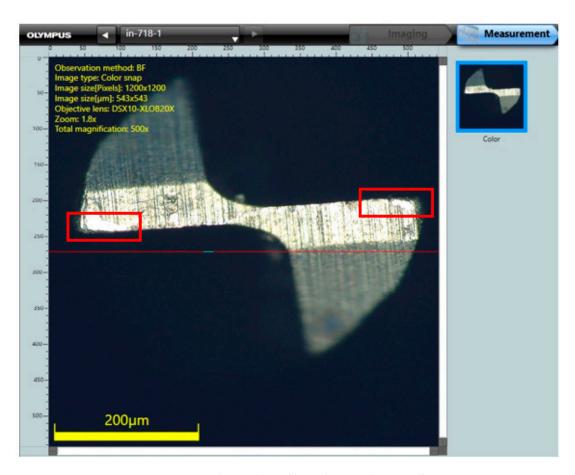


Figure 4. Cutting edges indicated by red rectangles on milling insert.



 $\label{lem:Figure 5.} \textbf{ Surface roughness measurement at yellow square}.$

Machines **2024**, *12*, 4 8 of 21

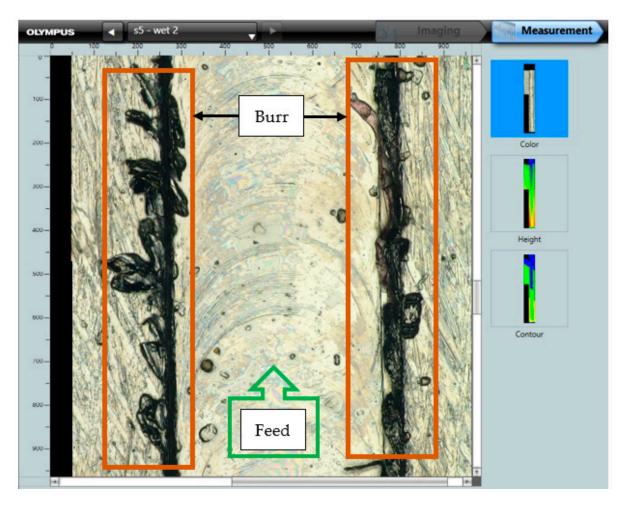


Figure 6. Burr generation during milling operation.

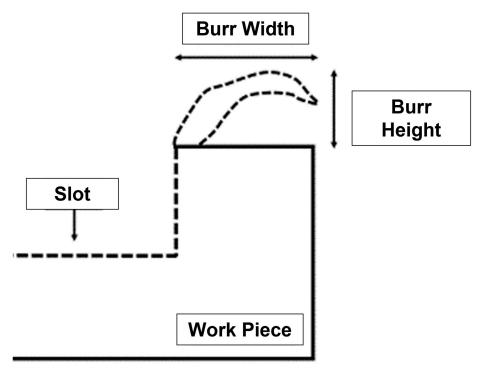
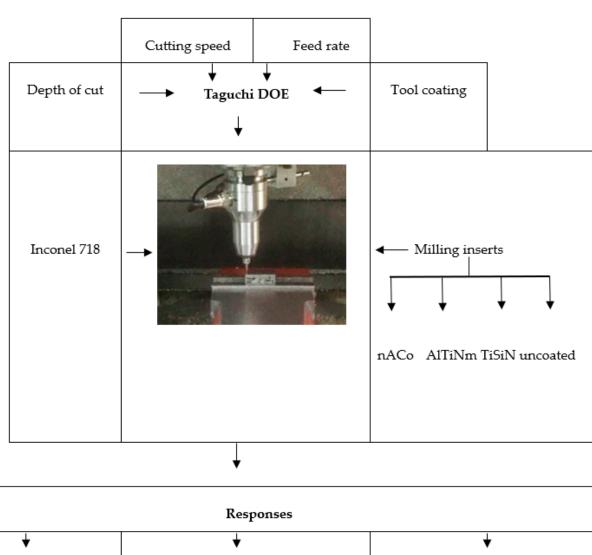


Figure 7. Measurement of burr width and burr height.

Machines **2024**, 12, 4 9 of 21



	Responses							
+	+	+						
Tool wear	Surface roughness	Burr formation						
+	+	+						
Main effects plot	Main effects plot	Main effects plot						
Analysis of variance	Analysis of variance	Analysis of variance						
Regression model analysis	Regression model analysis	Regression model analysis						
Contour plot	Contour plot	Contour plot						

Figure 8. Research methodology employed in this research.

Machines 2024, 12, 4 10 of 21

4. Results and Discussion

Experimental results for output responses including tool wear, surface roughness and burr formation were recorded working with a Taguchi design of experiment, as explained in Section 3.2. These results are displayed in Table 4. The output responses were selected keeping in view manufacturing system productivity. Tool wear is a vital response parameter indicative of manufacturing process economy. In addition, other significant responses including surface roughness (product quality) and energy consumption (process sustainability) depend upon tool wear. Tool wear has a negative impact on both the final product quality and the precision of the machining operation [41]. Keeping the above in view, tool wear was analyzed as an important factor in the machinability of Inconel 718. Surface roughness of machined parts is a significant phenomenon as it dictates product quality as well as product performance. The influence of surface roughness is more profound, especially in the micro-product industry, because of the need for precision and accuracy. Burr formation is a vital machining response owing to its impact on finished product quality, among other factors. It occurs on the edges of the machined surface, as already indicated in Section 3.3. Deburring a finished product is a challenging task at the micromachining level as it may change dimension distortion, in addition to causing other issues, including changes in elastic limit and residual stresses, as well as fatigue strength. Resultantly, it is imperative to keep the burr formation to a minimum. Burr analysis usually takes the burr width and burr height of both up-milling and down-milling into account. Experimental results obtained in the current research show that burr formation is more pronounced towards the down-milling side. The obvious reason for this occurrence is the lower comparative V_c towards the down-milling edge of the slot. Subsequent to the above discussion, it is inferred that the chosen output responses were selected keeping in view their importance in optimizing manufacturing output as well as their effects on other machining responses.

Table 4. Machining results for burr formation, surface roughness, tool wear.

				Output P	arameters		
F	(. l. D		Surface -		Burr Form	ation (µm)	
Experin	nental Run	Tool Wear (µm)	Roughness	Burr Height		Burr Width	
		(μπι)	(μm)	Left	Right	Left	Right
	1	100.136	0.069	102.128	262.403	338.413	510.167
	2	34.347	0.077	146.414	447.613	300.575	473.983
	3	46.885	0.095	251.747	219.736	505.049	403.683
	4	41.008	0.097	90.175	488.132	306.687	337.734
	5	52.952	0.084	51.642	451.017	268.437	515.621
	6	32.943	0.049	269.211	172.141	328.17	326.356
	7	42.439	0.073	124.613	183.593	373.504	362.552
	8	41.782	0.051	135.57	108.263	397.715	247.718
Run 1	9	37.150	0.065	141.336	256.026	285.573	434.421
	10	47.960	0.072	118.741	365.855	386.71	375.989
	11	55.085	0.048	108.777	385.27	260.425	314.183
	12	49.550	0.046	86.942	96.512	407.228	241.922
	13	29.146	0.071	66.358	428.854	293.971	437.448
	14	42.830	0.085	128.374	267.073	298.751	302.469
	15	42.704	0.106	102.557	439.501	368.26	277.863
	16	47.401	0.141	246.965	415.128	429.231	519.487

Machines **2024**, 12, 4 11 of 21

Table 4. Cont.

				Output P	arameters			
F	(.1 D		Surface -		Burr Formation (µm)			
Experin	nental Run	Tool Wear (µm)	Roughness	Burr I	Height	Burr	Width	
		(μπ)	(μm)	Left	Right	Left	Right	
	1	93.928	0.075	110.875	269.711	350.082	501.435	
	2	35.387	0.08	158.148	463.897	317.229	485.915	
	3	52.240	0.101	239.344	211.59	477.656	384.953	
	4	44.975	0.104	81.693	501.981	292.37	365.851	
	5	48.682	0.088	55.702	448.484	280.607	499.432	
	6	34.773	0.058	249.916	161.097	346.381	350.063	
	7	38.431	0.067	133.414	200.584	386.729	343.168	
	8	46.719	0.048	127.917	126.474	371.155	233.603	
Run 2	9	39.212	0.067	132.144	277.834	306.988	424.72	
	10	52.235	0.081	130.111	376.536	394.162	388.951	
	11	48.386	0.051	116.507	379.438	280.225	297.243	
	12	46.851	0.049	92.047	81.982	424.49	258.925	
	13	36.603	0.075	79.703	442.548	276.965	416.622	
	14	43.985	0.093	133.048	254.069	279.154	319.625	
	15	35.910	0.113	103.076	426.707	382.709	265.496	
	16	51.381	0.134	266.079	411.81	403.423	514.937	

In order to investigate the machinability of Inconel 718, an in-depth analysis of the experimental results achieved during milling operation was carried out. The parametric results were interpreted through MINITAB statistical software. Initially, main effects plots were drawn, which are informative representations of the input parameters on output responses. Then, analysis of variance was conducted to identify the contribution ratio of machining inputs. Regression analysis provided the equations for machining responses as an explicit function of input variables. Contour plots were also drawn to identify favorable combinations of machining inputs for higher output parameters. Individual analysis of machining responses is covered in succeeding sections.

4.1. Tool Wear Analysis

The main effects plot and analysis of variance for the experimental results for tool wear are given in Figure 9 and Table 5, respectively. The main effects plot shows the individual effect of input parameters on tool wear. The effect of V_c on tool wear is represented by an initially decreasing and then increasing trend. This is due to the shift of the machining range from conventional machining to the transition region. In terms of f_z , the tool wear decreased to a minimum at 0.5 f_z , which is, in fact, the tool edge radius. It is due to the higher elastic recovery that working at f_z equals the cutting tool radius during machining. A similar effect was observed for other workpiece materials by past researchers. The influence of these parameters was evident in the analysis of variance in terms of contribution ratio. The statistical result highlighted that t_c and a_p were most influential input parameters, with 36.19% and 25.51% contribution ratios. The uncoated micro tool caused the least wear among all tools, with the AlTiN coating being the second most effective tool. The nACo tool caused excessive tool wear. The reason for this occurrence is the highest thermal conductivity of the nACo-coated tool. A_p displayed a trend of reduced tool wear at higher values. On the other hand, tool wear initially decreased and then elevated with both V_c and

Machines **2024**, 12, 4 12 of 21

 f_z . Nevertheless, V_c and f_z proved to be moderate machining inputs with contribution ratios of 16.26% and 12.35%, respectively. Analysis of tool wear highlighted the effectiveness of the selected machining variables. Close observation of the main effects plot identified input parameters of 24 V_c , 0.50 fz $\mu m/t$ ooth, 90 μm ap as the optimum combination of input parameters for least tool wear when working with the uncoated micro tool. Similarly, the highest tool wear can be achieved at 9 m/min Vc, 0.25 fz μm/tooth, 30 μm ap using a nACocoated micro tool. These conditions were validated later in the conduct of the research. Regression analysis for tool wear yields four separate equations (Equations (1)–(4)). The reason for this occurrence is the categorical Input parameter of t_c. The trend of tool wear can also be equated with the positive intercept of the regression equations. The equation for the uncoated micro tool had a positive gain of 3%, 16% and 26% over AlTiN-, TiSiNand nACo-coated micro tools. In Figure 10, the contour plot for tool wear is given, which is a graphical representation of the output response. Here, points of favorable outcome can be ascertained. To ascertain the tool wear inflicted on the milling insert, both cutting edges are shown in a close-up image in Figure 11. The images highlight the tool wear which occurred during the cutting of Inconel 718 experiments. The brighter portion labeled on both images indicates the tool wear. The initial dominant tool wear mechanism is abrasive wear, which is mechanical in nature. Then, as the machining progresses, the workpiece material smears onto the cutting edge, resulting in a built-up edge which eventually breaks off, taking a portion of the insert. This occurs periodically when the insert comes into contact with the workpiece under the force of flowing chips. Resultantly, the cutting edge becomes dull, which increases the surface roughness on the newly machined workpiece surface. Surface roughness, which is a vital product quality index, is discussed in Section 4.2.

$$TW (nACo) = 90.2 - 0.798 V_c - 8.34 f_z - 0.2433 a_p$$
 (1)

$$TW (AlTiN) = 73.9 - 0.798 V_c - 8.34 f_z - 0.2433 a_p$$
 (2)

$$TW (TiSiN) = 82.9 - 0.798 V_c - 8.34 f_z - 0.2433 a_p$$
 (3)

$$TW (Uc) = 71.6 - 0.798 V_c - 8.34 f_z - 0.2433 a_p$$
 (4)

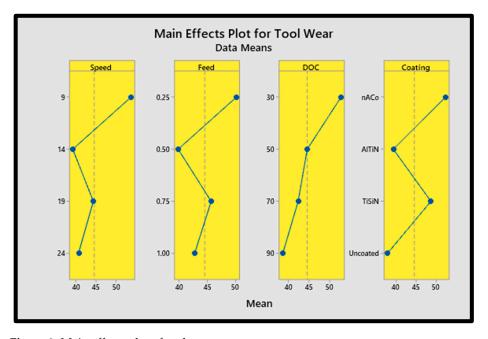


Figure 9. Main effects plot of tool wear.

Machines **2024**, 12, 4 13 of 21

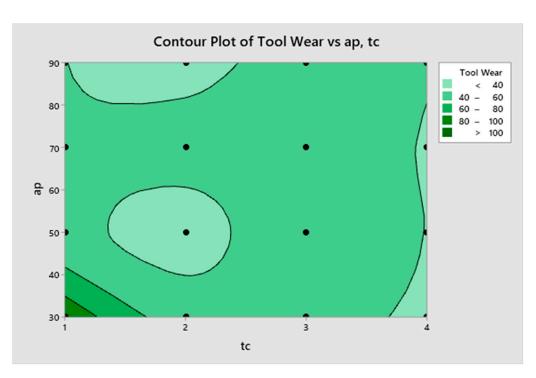


Figure 10. Contour plot for tool wear.



 $\label{lem:Figure 11.} \textbf{Close-up imagery of the milling insert highlighting tool wear.}$

Machines 2024, 12, 4 14 of 21

Source	DF	Seq SS	Contribution	Adj SS	Adj MS	F-Value	<i>p</i> -Value
V_c	3	1100.0	16.26%	1100.0	366.66	4.17	0.020
f_z	3	835.7	12.35%	835.7	278.57	3.17	0.048
a_p	3	1387.9	25.51%	1387.9	462.64	5.26	0.008
$t_{c}^{'}$	3	1772.0	36.19%	1772.0	590.68	6.72	0.003
Lack of Fit	3	1500.3	5.17%	1500.3	500.11		
Pure Error	16	169.8	4.51%	169.8	10.62		
Total	31	6765.8	100.00%				

Table 5. Analysis of variance for tool wear parameter.

4.2. Surface Roughness Analysis

The main effects plot and analysis of variance of input cutting parameters, along with different tool coatings on surface finish, are given in Figure 12 and Table 6, respectively. With respect to V_c, surface roughness tended to slightly decrease before elevating as V_c was increased from 9 to 24. The effect of t_c on surface roughness was also evident, with the AlTiN-coated tool yielding the best product quality. This is owing to the low thermal coefficient, which keeps the cutting zone temperature in check. This results in lower tool wear which, in turn, produces better surface quality. It can also be seen that surface integrity diminished with an increase in both f_z and a_p , although the effect was minimal and seems insignificant. A higher f_z facilitates a higher volume of machining. The surface quality is affected by tool wear brought on by a machining operation, as already mentioned above. One of the most important elements in surface finish is the tool's cutting edge radius [42]. It is preferable to micro-mill on the smallest chips in order to create good surface roughness because doing so minimizes surface distortion and therefore enhances surface clarity. The best and worst results in terms of surface roughness were predicted to occur at 19 m/min Vc, 0.25 fz µm/tooth, 70 µm ap with the AlTiN-coated tool and 24 m/min Vc, 1.00 fz μm/tooth, 50 μm ap with the TiSiN-coated tool, respectively. In order to quantify the influence of input variables on surface roughness, analysis of variance was conducted, with results displayed in Table 6. Fz and ap were found to be insignificant owing to their meager effect on surface roughness, whereas V_c and t_c were identified as the most effective members with contribution ratios of 51.24% and 34.27%, respectively. Another comparison of the different tool coatings used in this work was made through the analysis of regression equations, as given by Equations (5)–(8). The equation for the AlTiN-coated tool was the most efficient, with the least positive intercept of 0.0402, as it was based on a smaller-the-better model. In comparison, the AlTiN tool was 17%, 49% and 91% more effective than the nACo-coated, uncoated and TiSiN-coated tools, respectively. In order to find the optimum combinations of vital inputs, a contour plot was drawn. The result is displayed in Figure 13. Here, the brighter area of the graph represents better surface finish. The black circular points highlight the combination of various t_c values at certain V_c values. This is a substantial visual tool for optimizing the output response.

$$SR (nACo) = 0.0472 + 0.000802 V_c + 0.0143 f_z + 0.000004 a_p$$
 (5)

$$SR (AlTiN) = 0.0402 + 0.000802 V_c + 0.0143 f_z + 0.000004 a_p$$
 (6)

$$SR (TiSiN) = 0.0770 + 0.000802 V_c + 0.0143 f_z + 0.000004 a_p$$
 (7)

$$SR (Uc) = 0.0599 + 0.000802 V_c + 0.0143 f_z + 0.000004 a_p$$
 (8)

Machines **2024**, 12, 4 15 of 21

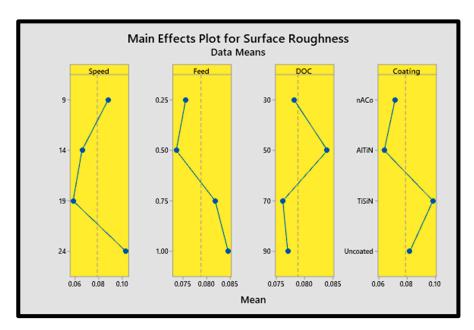


Figure 12. Main effects plot of surface roughness.

Table 6. ANOVA for surface roughness parameter.

Source	DF	Seq SS	Contribution	Adj SS	Adj MS	F-Value	<i>p</i> -Value
V _c	3	0.009413	51.24%	0.009413	0.003138	30.38	0.000
f_z	3	0.000586	3.19%	0.000586	0.000195	1.89	0.166
a_p	3	0.000113	0.62%	0.000113	0.000038	0.37	0.778
$t_{\rm c}^{'}$	3	0.006294	34.27%	0.006294	0.002098	20.32	0.000
Lack of Fit	3	0.001686	9.18%	0.001686	0.000562		
Pure Error	16	0.000276	1.51%	0.000276	0.000017		
Total	31	0.018368	100.00%				

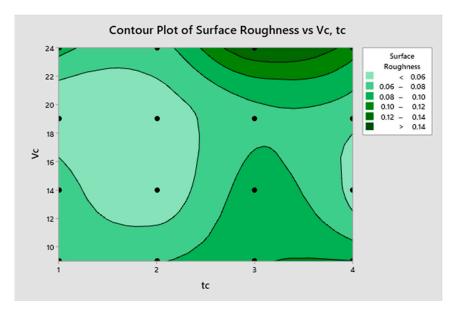


Figure 13. Contour plot for surface roughness.

4.3. Burr Formation Analysis

The effect of individual input parameters is displayed as the main effects plot given in Figure 14. One observation related to t_c is that the nACo-coated tool produced the

Machines **2024**, 12, 4 16 of 21

smallest burr in all cases, except in case of of burr width when down-milling. On the contrary, the TiSiN-coated tool yielded the highest recorded burr width and burr height in both up-milling and down-milling conditions. A probable reason for this is the highest coefficient of friction of the TiSiN-coated tool, resulting in intense shear deformation during milling. The cutting edge radius (mentioned in Section 3.1) is a significant parameter in comparison to f_z since, at f_z values higher than the cutting radius, lesser burrs were formed. Furthermore, it was found that increasing V_c resulted in a larger burr because V_c caused significant changes in cutting temperature [43]. A significant positive shift in burr formation was observed at 0.5 mm, which is the cutting edge radius of the milling tool. The exact effect of input variables was quantified using analysis of variance, given in Tables 7–10. A_p was identified as the significant input in the case of burn height for both up-milling and down-milling with contribution ratios of 45.12% and 23.25%. On the other hand, f_z was the dominant parameter for burr width, having contribution ratios of 39.28% and 36.26% for up-milling and down-milling, respectively. For burr height and burr width (up-milling), V_c proved to be insignificant, whereas a_p was found to be insignificant for burr width during both up-milling and down-milling machining. T_c was singled out as the input variable with significant influence on process output (burr formation) in all four cases analyzed. The optimum conditions for minimum burring for the four desired cases (burr height and burr width for both up-milling and down-milling) were noted for the conduct of subsequent validation experiments.

Table 7. Analysis of variance for burr height up-milling parameter.

Source	DF	Seq SS	Contribution	Adj SS	Adj MS	F-Value	<i>p</i> -Value
V _c	3	4931	4.07%	4931.1	1643.7	0.94	0.439
f_z	3	23,936	19.73%	23,935.9	7978.6	4.58	0.014
a_p	3	42,610	45.12%	42,610.0	14,203.3	8.15	0.001
$t_{c}^{'}$	3	16,744	13.80%	16,743.6	5581.2	3.20	0.047
Lack of Fit	3	32,188	16.53%	32,187.6	10,729.2	187.62	0.000
Pure Error	16	915	0.75%	915.0	57.2		
Total	31	121,324	100.00%				

Table 8. Analysis of variance for burr height down-milling parameter.

Source	DF	Seq SS	Contribution	Adj SS	Adj MS	F-Value	<i>p</i> -Value
V _c	3	121,906	23.30%	121,906	40,635.2	6.32	0.004
f_z	3	23,632	4.52%	23,632	7877.2	1.23	0.328
a_p	3	121,653	23.25%	121,653	40,551.1	6.31	0.004
$t_{c}^{'}$	3	133,904	35.59%	133,904	44,634.7	6.95	0.000
Lack of Fit	3	120,762	13.08%	120,762	40,254.1	478.58	0.000
Pure Error	16	1346	0.26%	1346	84.1		
Total	31	523,203	100.00%				

Table 9. Analysis of variance for burr width up-milling parameter.

Source	DF	Seq SS	Contribution	Adj SS	Adj MS	F-Value	<i>p</i> -Value
V_{c}	3	1981	1.61%	1981	660.5	0.25	0.863
f_z	3	36,084	39.28%	36,084	12,028.1	4.48	0.015
a_p	3	3470	2.82%	3470	1156.5	0.43	0.733
t_c	3	30,683	34.90%	30,683	10,227.7	3.81	0.027
Lack of Fit	3	48,272	19.17%	48,272	16,090.8	93.94	0.000
Pure Error	16	2741	2.22%	2741	171.3		
Total	31	123,232	100.00%				

Machines **2024**, 12, 4 17 of 21

Source	DF	Seq SS	Contribution	Adj SS	Adj MS	F-Value	<i>p-</i> Value
V _c	3	37,228	14.50%	37,228	12,409.3	4.71	0.013
f_z	3	93,120	36.26%	93,120	31,039.9	11.78	0.000
a _p	3	21,269	8.28%	21,269	7089.7	2.69	0.075
t_{c}^{r}	3	55,101	21.46%	55,101	18,366.8	6.97	0.002
Lack of Fit	3	47,813	10.62%	47,813	15,937.7	113.36	0.000
Pure Error	16	2249	8.88%	2249	140.6		
Total	31	256,780	100.00%				

Table 10. Analysis of variance for burr width down-milling parameter.

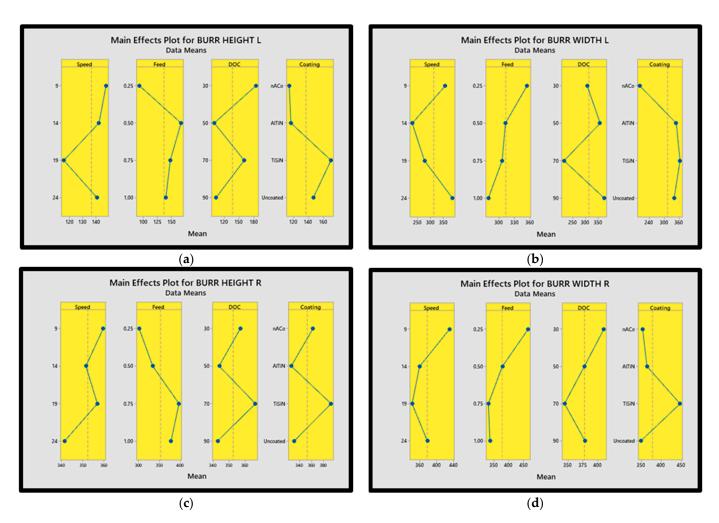


Figure 14. Main effects plot for burr formation: (a) burr width up-milling; (b) burr height up-milling; (c) burr width down-milling; (d) burr height down-milling.

5. Validation Tests

The current research focuses on the analysis of output response with tool coatings as the input variable in addition to other fundamental machining parameters. Such a methodology initially requires the identification of significant input variables in terms of their contribution ratio followed by selection of their discrete values which will contribute towards optimization. In this research, tool wear, surface roughness and burr formation were selected, based on the smaller-is-better model. Table 11 displays the desired values of input parameters as identified through the main effects plot in Section 4. Resultantly, validation experiments under these conditions were performed to confirm the authenticity of the obtained results. These results are given in Table 12. Here, 3 out of the 12 machining parameter combinations were already experimented on during the initial de-

Machines 2024, 12, 4 18 of 21

sign of the experiment (L16 orthogonal array). The rest of the conditions were run three times each. It was observed that all best and worst output response conditions yielded corroborative results.

Table 11. Machining condition sets for individual best and worst responses.

Responses			Input Parameters			
		Condition	V _c (m/min)	f _z (μm/tooth)	a _p (μm)	t _c
Tool Wear (μm)		Best	14	0.50	90	uncoated
		Worst	9	0.25	30	nACo
Surface Roughness (μm)		Best	19	0.25	70	AlTiN
		Worst	24	1.00	50	TiSiN
Burr Width (μm)	Left	Best	24	0.25	90	AlTiN
		Worst	9	0.75	70	TiSiN
	Right	Best	19	0.75	70	uncoated
		Worst	9	0.25	30	TiSiN
Burr Height (μm)	Left	Best	19	0.25	50	nACo
		Worst	9	0.50	30	TiSiN
	Right	Best	14	1.00	70	nACo
		Worst	24	0.25	90	TiSiN

Table 12. Results of validation tests.

Responses		Condition	Validation Test	Initial Run	Difference
Tool Wear (μm)		Best 27.956 μm		29.146 μm	4.0%
		Worst	Already examined	100.136 μm	-
Surface Roughness (μm)		Best 0.044 μm		0.046 μm	4.3%
		Worst	0.144 μm	0.141 μm	2.1%
Burr Width (µm)	Left	Best	Already examined	260.425 μm	-
		Worst	Already examined	505.049 μm	-
	Right	Best	221.042 μm	233.603 μm	5.3%
		Worst	527.401 μm	519.487 μm	1.5%
Burr Height (µm)	Left	Best	48.972 μm	51.642 μm	5.1%
		Worst	274.075 μm	269.211 μm	1.8%
	Right	Best	77.416 μm	81.982 μm	5.6%
		Worst	533.092 μm	501.981 μm	6.2%

6. Conclusions

In the current experimental research, Inconel 718 was micromachined using end mills with different coatings to analyze the effects of input machining parameters on tool wear, surface roughness and burr formation. The following conclusions were reached during the conduct of this work:

- 1. Tool wear was found to be significantly affected by cutting speed, feed rate, depth of cut and tool coating, with tool coating bearing the highest contribution ratio of 36.19%:
- 2. Abrasion and chipping were found to be the dominant tool wear mechanisms;

Machines 2024, 12, 4 19 of 21

3. Cutting speed and tool coating had contribution ratios of 51.24% and 34.27%, respectively, on surface roughness response;

- 4. Analysis of regression equations for surface roughness determined positive gains of 17%, 49% and 91% using the AlTiN-coated tool over the nACo-coated, uncoated and TiSiN-coated tools;
- 5. Burr formation analysis identified depth of cut as an influential parameter for burr height formation for both up-milling (39.28%) and down-milling (36.26%);
- 6. Cutting tool edge radius is a vital input parameter affecting burr formation;
- Tool coating input was singled out as the only parameter significantly affecting all machining responses;
- 8. Validation of confirmatory experiments endorsed the accuracy of the experimental procedure by improving the output responses.

Author Contributions: Conceptualization, M.I.F.; Data curation, M.I.F.; Formal analysis, M.I.F. and J.P.; Investigation, M.I.F.; Methodology, J.P.; Resources, J.P.; Software, J.P.; Supervision, J.P.; Validation, M.I.F.; Visualization, J.P.; Writing—original draft, M.I.F.; Writing—review and editing, J.P. All authors have read and agreed to the published version of the manuscript.

Funding: The work was funded through the Department of Machining, Assembly and Engineering Metrology, Mechanical Engineering Faculty, VŠB-Technical University of Ostrava, 17. Listopadu 2172/15, 708 00 Ostrava, Czech Republic. This work was supported by the Deanship of Scientific Research, Vice Presidency for Graduate Studies and Scientific Research, King Faisal University, Saudi Arabia (Grant no. 5335).

Data Availability Statement: Data of the research are being used for further extended research and can be made available in due course.

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

References

- 1. Jain, V.K.; Patel, D.S.; Ramkumar, J.; Bhattacharyya, B.; Doloi, B.; Sarkar, B.R.; Ranjan, P.; Sarath Sankar, E.S.; Jayal, A.D. Micro-machining: An overview (Part II). *J. Micromanuf.* **2022**, *5*, 46–73. [CrossRef]
- 2. Chen, N.; Li, H.N.; Wu, J.; Li, Z.; Li, L.; Liu, G.; He, N. Advances in micro milling: From tool fabrication to process outcomes. *Int. J. Mach. Tools Manuf.* **2021**, *160*, 103670. [CrossRef]
- 3. Corbett, J.; McKeown, R.A.; Peggs, G.N.; Whatmore, R. Nanotechnology: International developments and emerging products. *CIRP Ann.* **2000**, *49*, 523–545. [CrossRef]
- 4. Kiswanto, G.; Azmi, M.; Mandala, A.; Ko, T.J. The Effect of Machining Parameters to the Surface Roughness in Low Speed Machining Micro-milling Inconel 718. In Proceedings of the IOP Conference Series: Materials Science and Engineering, Incheon, Republic of Korea, 19–22 August 2019; Volume 654.
- 5. Schuster, R.; Kirchner, V.; Allongue, P.; Ertl, G. Electrochemical micromachining. Science 2000, 289, 98–101. [CrossRef] [PubMed]
- 6. Kock, M.; Kirchner, V.; Schuster, R. Electrochemical micromachining with ultrashort voltage pulses-a versatile method with lithographical precision. *Electrochim. Acta* **2003**, *48*, 3213–3219. [CrossRef]
- 7. Kim, B.H.; Ryu, S.H.; Choi, D.K.; Chu, C.N. Micro electrochemical milling. *J. Micromechanics Microengineering* **2005**, *15*, 124–129. [CrossRef]
- 8. Attanasio, A. Tool run-out measurement in micro milling. *Micromachines* **2017**, 8, 221. [CrossRef] [PubMed]
- 9. Liu, Y.; Zhu, D.; Zhu, L. Micro electrochemical milling of complex structures by using in situ fabricated cylindrical electrode. *Int. J. Adv. Manuf. Technol.* **2012**, *60*, 977–984. [CrossRef]
- 10. Xu, L.; Zhao, C. Nanometer-scale accuracy electrochemical micromachining with adjustable inductance. *Electrochim. Acta* **2017**, 248, 75–78. [CrossRef]
- 11. Bissacco, G.; Hansen, H.N.; De Chiffre, L. Size effects on surface generation in micro milling of hardened tool steel. *CIRP Ann.* **2006**, *55*, 593–596. [CrossRef]
- 12. Mian, A.J.; Driver, N.; Mativenga, P.T. Identification of factors that dominate size effect in micro-machining. *Int. J. Mach. Tools Manuf.* **2011**, *51*, 383–394. [CrossRef]
- 13. Ling, S.; Li, M.; Liu, Y.; Wang, K.; Jiang, Y. Improving machining localization and surface roughness in wire electrochemical micromachining using a rotating ultrasonic helix electrode. *Micromachines* **2020**, *11*, 698. [CrossRef] [PubMed]
- 14. Allegri, G.; Colpani, A.; Ginestra, P.S.; Attanasio, A. An experimental study on micro-milling of a medical grade Co-Cr-Mo alloy produced by selective laser melting. *Materials* **2019**, *12*, 2208. [CrossRef] [PubMed]

Machines **2024**, 12, 4 20 of 21

15. Wu, M.; Saxena, K.K.; Guo, Z.; Qian, J.; Reynaerts, D. Fast fabrication of complex surficial micro-features using sequential lithography and jet electrochemical machining. *Micromachines* **2020**, *11*, 948. [CrossRef] [PubMed]

- 16. Marrocco, V.; Modica, F.; Bellantone, V.; Medri, V.; Fassi, I. Pulse-type influence on the micro-edm milling machinability of Si₃N₄—Tin workpieces. *Micromachines* **2020**, *11*, 932. [CrossRef] [PubMed]
- 17. Rahman, M.A.; Ali, M.Y.; Rosli, A.R.S.; Banu, A. Process Capability of High Speed Micro End-Milling of Inconel 718 with Minimum Quantity Lubrication. In Proceedings of the IOP Conference Series: Materials Science and Engineering, Kuala Lumpur, Malaysia, 25–27 July 2016; Volume 184.
- 18. Markopoulos, A.P.; Karkalos, N.E.; Mia, M.; Pimenov, D.Y.; Gupta, M.K.; Hegab, H.; Khanna, N.; Aizebeoje Balogun, V.; Sharma, S. Sustainability assessment, investigations, and modelling of slot milling characteristics in eco-benign machining of hardened steel. *Metals* 2020, 10, 1650. [CrossRef]
- 19. Tadavani, S.A.; Razavi, R.S.; Vafaei, R. Pulsed laser-assisted machining of Inconel 718 superalloy. *Opt. Laser Technol.* **2017**, *87*, 72–78. [CrossRef]
- 20. Reddy, M.M.; Nie, V.Y.C. Evaluation of surface roughness and tool wear in high speed machining of Inconel 718. In *Materials Science Forum*; Trans Tech Publications: Stafa-Zurich, Switzerland, 2018; Volume 943, pp. 66–71.
- 21. Ucun, I.; Aslantas, K.; Bedir, F. An experimental investigation of the effect of coating material on tool wear in micro milling of Inconel 718 super alloy. *Wear* **2013**, 300, 8–19. [CrossRef]
- 22. Lu, X.; Jia, Z.; Wang, H.; Si, L.; Wang, X. Surface roughness prediction model of micro-milling Inconel 718 with consideration of tool wear. *Int. J. Nanomanuf.* **2016**, 12, 93–108. [CrossRef]
- 23. Aslantas, K.; Hopa, H.E.; Percin, M.; Ucun, I.; Çicek, A. Cutting performance of nano-crystalline diamond (NCD) coating in micro-milling of Ti6Al4V alloy. *Precis. Eng.* **2016**, *45*, 55–66. [CrossRef]
- 24. Özel, T.; Thepsonthi, T.; Ulutan, D.; Kaftanolu, B. Experiments and finite element simulations on micro-milling of Ti-6Al-4V alloy with uncoated and cBN coated micro-tools. CIRP Ann. 2011, 60, 85–88. [CrossRef]
- 25. Aramcharoen, A.; Mativenga, P.T.; Yang, S.; Cooke, K.E.; Teer, D.G. Evaluation and selection of hard coatings for micro milling of hardened tool steel. *Int. J. Mach. Tools Manuf.* **2008**, *48*, 1578–1584. [CrossRef]
- 26. Devillez, A.; Le Coz, G.; Dominiak, S.; Dudzinski, D. Dry machining of Inconel 718, workpiece surface integrity. *J. Mater. Process. Technol.* 2011, 211, 1590–1598. [CrossRef]
- 27. Tansel, I.N.; Arkan, T.T.; Bao, W.Y.; Mahendrakar, N.; Shisler, B.; Smith, D.; McCool, M. Tool wear estimation in micro-machining. *Int. J. Mach. Tools Manuf.* 2000, 40, 609–620. [CrossRef]
- 28. Attanasio, A.; Gelfi, M.; Pola, A.; Ceretti, E.; Giardini, C. Influence of material microstructures in micromilling of Ti6Al4V alloy. *Materials* **2013**, *6*, 4268–4283. [CrossRef] [PubMed]
- 29. Sun, Z.; To, S. Effect of machining parameters and toolwear on surface uniformity in micro-milling. *Micromachines* **2018**, *9*, 268. [CrossRef] [PubMed]
- 30. Aurich, C.; Bohley, M.; Reichenbach, I.G.; Kirsch, B. Surface quality in micro milling: Influences of spindle and cutting parameters. *CIRP Ann.* **2017**, *66*, 101–104. [CrossRef]
- 31. Pradhan, P.; Costa, L.; Rybski, D.; Lucht, W.; Kropp, J.P. A Systematic Study of Sustainable Development Goal (SDG) Interactions. *Earth's Future* **2017**, *5*, 1169–1179. [CrossRef]
- 32. IEA. World Energy Outlook 2018: Highlights; International Energy Agency (IEA): Paris, France, 2018; pp. 1–661.
- 33. Kumar, R.; Bilga, P.S.; Singh, S. Multi objective optimization using different methods of assigning weights to energy consumption responses, surface roughness and material removal rate during rough turning operation. *J. Clean. Prod.* **2017**, *164*, 45–57. [CrossRef]
- 34. Thakur, D.G.; Ramamoorthy, B.; Vijayaraghavan, L. Machinability investigation of Inconel 718 in high-speed turning. *Int. J. Adv. Manuf. Technol.* **2009**, 45, 421–429. [CrossRef]
- 35. Hughes, I.; Sharman, A.R.C.; Ridgway, K. The effect of cutting tool material and edge geometry on tool life and workpiece surface integrity. *Proc. Inst. Mech. Eng. Part B J. Eng. Manuf.* **2006**, 220, 93–107. [CrossRef]
- 36. Barry, J.; Byrne, G.; Lennon, D. Observations on chip formation and acoustic emission in machining Ti-6Al-4V alloy. *Int. J. Mach. Tools Manuf.* **2001**, *41*, 1055–1070. [CrossRef]
- 37. Liu, R.; Mittal, S. Single-step superfinish hard machining: Feasibility and feasible cutting conditions. *Robot. Comput. Integr. Manuf.* 1996, 12, 15–27. [CrossRef]
- 38. Zhou, M.; Chen, Y.; Zhang, G. Force prediction and cutting-parameter optimization in micro-milling Al7075-T6 based on response surface method. *Micromachines* **2020**, *11*, 766. [CrossRef] [PubMed]
- 39. Karna, S.K.; Sahai, R. An Overview on Taguchi Method. Int. J. Eng. Math. Sci. 2016, 1, 11–18.
- 40. *ISO 21920-2:2021*; Geometrical Product Specifications (Gps)—Surface Texture: Profile—Part 2: Terms, Definitions. ISO: Geneva, Switzerland, 2021.
- 41. Dadgari, A.; Huo, D.; Swailes, D. Investigation on tool wear and tool life prediction in micro-milling of Ti-6Al-4V. *Nanotechnol. Precis. Eng.* **2018**, *1*, 218–225. [CrossRef]

Machines **2024**, 12, 4 21 of 21

42. Lu, X.; Jia, Z.; Wang, H.; Hu, X.; Li, G.; Si, L. Measurement-based modelling of cutting forces in micro-milling of Inconel 718. *Int. J. Nanomanuf.* **2017**, *13*, 1–11. [CrossRef]

43. Muhammad, A.; Gupta, M.K.; Mikołajczyk, T.; Pimenov, D.Y.; Giasin, K. Effect of Tool Coating and Cutting Parameters on Surface Roughness and Burr Formation during Micromilling of Inconel 718. *Metals* **2021**, *11*, 167. [CrossRef]

Disclaimer/Publisher's Note: The statements, opinions and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions or products referred to in the content.