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## EXPERIMENTAL INVESTIGATION OF SURFACE QUALITY IN PLA 3D PRINTED COMPONENT USING A DOE APPROACH

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***Abstract:** Surface roughness is a critical quality parameter in 3D-printed components, influencing mechanical properties, aesthetic appeal, and functional performance. This study investigates the influence of key process parameters on the surface roughness of Polylactic Acid 3D-printed parts using a Design of Experiments approach. A structured experimental methodology was developed to analyze the effects of the default thickness, number of walls and infill density on surface quality, quantified through Ra measurements. The experiments were designed using a Taguchi orthogonal array to optimize parameter combinations while minimizing experimental runs. An ANOVA analysis and a regression model was performed to determine the significance of each factor. The study provides an optimized parameter set for improving the surface roughness of 3D printed parts.*

***Key words:** PLA, 3D printing, surface roughness, Design of Experiments, Taguchi method, ANOVA*

### 1. INTRODUCTION

Additive manufacturing (AM) enables users to create complex structures, with minimal material, time and cost. 3D printed parts require no molds or dies, being accessible to be used in a variety of industries [1]. In this context, quality parameters of 3D printed parts have been optimized to generate consistent and competitive products. Several researchers studied the influence of 3D printing parameters on the average surface roughness of 3D printed parts using Fused Deposition Modelling (FDM). Researchers demonstrated that having a low temperature bed, high printing speed, a medium infill, low layer thickness and a low number of shells are recommended to reduce surface roughness (Ra), when using ABS [2]. Another study demonstrated that when using PLA, the lowest values for Ra were obtained with a low layer height and a high speed [3]. Surface roughness is also influenced by the measuring direction. If a wall has a higher tilt angle, the value of Ra also has higher registered values [4]. To analyze the effects of three printing parameters the current study proposed a Taguchi orthogonal array for 24 PLA samples. The

selected printing parameters considered important for the study were infill, default thickness and wall line count. Samples were measured using a digital roughness tester and weighed using a four decimal electronic scale. An Analysis of Variance(ANOVA) was used to determine which factor had the most impact on the measured surface roughness, by comparing the mean values of the three groups.

### 2. METHODS

To determine the surface quality in a PLA 3D printed component an experimental plan based on a Taguchi orthogonal array was used, allowing different parameter combination comparisons while minimizing the number of experimental runs. For mass and roughness, an ANOVA method was applied to determine each of the parameters' significance and their interactions. The Taguchi method was chosen due to its significant advantages in reducing the number of experiments, thus saving time and resources, while identifying critical factors, improving product performance and enabling data-driven optimization. In the implementation of the Design of Experiments (DOE) methodology, the experimental process was

structured into four distinct and methodologically defined stages.

The *first stage* involves defining the factors and establishing the levels. Three factors were defined to be analyzed and optimized, selected from the responses obtained through the systemic analysis, as follows: Infill Density (%); Wall Line Count (number of perimeters/walls of the part); Wall Thickness (the Default Thickness measured in mm).

Each of the three factors will be tested at three levels (Table 1), starting from a Taguchi L9 (3<sup>3</sup>) orthogonal array to optimize the number of experiments. An extension of the experiment is proposed for a more robust analysis, reducing the risk of statistical errors when applying ANOVA. It was decided to replicate the factors at the intermediate levels, resulting in an L12 matrix (Table 2).

Table 1

Levels of the three factors selected for analysis				
No.	Parameter	Minimum	Average	Maximum
1	Infill	15%	40%	100%
2	Wall Line Count	1	2	3
3	Default Thickness	0.45 mm	0.67 mm	1 mm

Table 2

No.	Variables			Variable Codes		
	Infill	Wall Line Count	Default Thickness	A	B	C
E1	40%	2	0.67 mm	0	0	0
E2	15%	1	0.45 mm	-1	-1	-1
E3	15%	1	1 mm	-1	-1	+1
E4	15%	3	0.45 mm	-1	+1	-1
E5	15%	3	1 mm	-1	+1	+1
E6	40%	2	0.67 mm	0	0	0
E7	100%	3	1 mm	+1	+1	+1
E8	100%	3	0.45 mm	+1	+1	-1
E9	100%	1	1 mm	+1	-1	+1
E10	100%	1	0.45 mm	+1	-1	-1
E11	40%	2	0.67 mm	0	0	0
E12	40%	2	0.67 mm	0	0	0

This array will make sure that each factor will be equally tested.

In the *second phase* of the experimental procedure, a selection of 24 samples, in accordance with the Taguchi array, was

additively manufactured using FDM. The material chosen for fabrication was Winkle PLA filament, utilized in two color variants (Glacier White and Jet Black) sourced from the same manufacturer. This approach allows for the investigation of whether the color pigment present in the filament has any measurable influence on the experimental outcomes, while maintaining consistency in material composition and production origin. Each of the 24 specimens was individually fabricated on the 3D printer’s build platform (Fig.1) to eliminate potential sources of variability associated with part positioning, such as non-uniform thermal distribution, localized cooling rates, or bed-leveling inconsistencies. This controlled printing strategy was employed to ensure the repeatability and reliability of the experimental results by minimizing process-induced deviations.

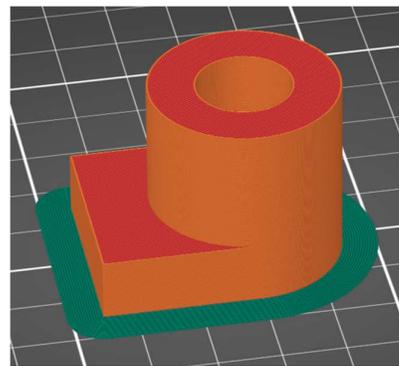


Fig. 1. The position of the sample on the printer bed

For the fabrication of the 3D printed samples a Prusa MK4S was used loaded with PLA material. The printing parameters for part fabrication were set as given in Table 2, for every combination. An outer brim was also added and lower the first layer speed to 5 mm/s in order to ensure printing adhesion.

In the *third phase* of the experimental protocol, quantitative assessments were conducted to determine the mass and surface roughness of the fabricated specimens. Each of the 24 samples, comprising 12 white PLA and 12 black PLA specimens, was weighed using a Pioneer Plus PA323 precision electronic scale to ensure accurate mass measurements. Surface roughness evaluations were performed employing a contact INSIZE roughness tester to

measure the arithmetic average roughness (Ra) parameter. This dual-measurement approach facilitated a comparative analysis of the influence exerted by color pigmentation on both the mass and surface texture characteristics of PLA-based 3D-printed components.

The *fourth stage* involves applying ANOVA to examine the statistical influence of each studied parameter.

### 3. RESULTS AND DISCUSSIONS

#### 3.1 Mass measurement and analysis

The mass of each specimen was measured using a Pioneer Plus precision electronic scale with a resolution of four decimal places, ensuring accurate and repeatable readings for subsequent statistical analysis. The complete dataset for the white and black PLA samples is presented in Table 3 and Table 4, respectively. Each sample corresponds to a unique combination of three key printing parameters: infill density, wall line count, and default wall thickness.

Table 3

No.	Sample code	Variables			Mass [g]
		Infill	Wall Line Count	Default Thickness	
1	E1_A	40%	2	0.67 mm	15.1899
2	E2_A	15%	1	0.45 mm	7.7154
3	E3_A	15%	1	1 mm	10.5110
4	E4_A	15%	3	0.45 mm	11.0470
5	E5_A	15%	3	1 mm	18.5983
6	E6_A	40%	2	0.67 mm	15.2743
7	E7_A	100%	3	1 mm	23.9801
8	E8_A	100%	3	0.45 mm	23.6209
9	E9_A	100%	1	1 mm	23.7957
10	E10_A	100%	1	0.45 mm	23.6798
11	E11_A	40%	2	0.67 mm	15.1818
12	E12_A	40%	2	0.67 mm	15.2156

To evaluate the influence of the selected process parameters on sample weight, a one-way single factor ANOVA was performed separately for white and black PLA specimens.

The three independent variables (Infill, Wall Line Count, and Default Wall Thickness) were treated as categorical factors, while sample

weight was considered the dependent variable. The ANOVA results are summarized in Table 5.

Table 4

No.	Sample code	Variables			Mass [g]
		Infill	Wall Line Count	Default Thickness	
1	E1_N	40%	2	0.67	15.5496
2	E2_N	15%	1	0.45 mm	7.8454
3	E3_N	15%	1	1 mm	10.7222
4	E4_N	15%	3	0.45 mm	11.3403
5	E5_N	15%	3	1 mm	19.0465
6	E6_N	40%	2	0.67 mm	15.5266
7	E7_N	100%	3	1 mm	24.3688
8	E8_N	100%	3	0.45 mm	24.0156
9	E9_N	100%	1	1 mm	24.2803
10	E10_N	100%	1	0.45 mm	24.0848
11	E11_N	40%	2	0.67 mm	15.5397
12	E12_N	40%	2	0.67 mm	15.5332

Table 5

Parameters	Infill	Wall Line Count	Default Thickness
F-stat White	20.552	0.487	0.456
p-value White	0.0004	0.629	0.647
Interpretation White	☑ Significant	△ Almost significant	△ Almost significant
F-stat Black	19.915	0.489	0.459
p-value Black	0.0004	0.628	0.645
Interpretation Black	☑ Significant	△ Almost significant	△ Almost significant

The statistical analysis reveals that, for both white and black PLA materials, the Infill parameter exhibits a statistically significant influence on the final weight of the printed samples (p-value < 0.001), as indicated by the high F-statistic values. Conversely, the Wall Line Count and Default Wall Thickness parameters show no statistically significant effect, with p-values well above the commonly accepted threshold of 0.05. These results are consistent across both color variants, suggesting that infill percentage predominantly governs the mass of FDM-printed parts, while modifications in wall count or thickness have a comparatively negligible impact under the conditions tested.

The consistency between the two datasets confirms that the influence of filament color (i.e., pigment variation) does not introduce any systematic bias in weight outcomes, provided

that all other printing parameters are held constant.

**3.2. Roughness measurement and analysis**

Surface roughness measurements were performed using an INSIZE surface roughness tester. For each specimen, measurements were taken at seven distinct locations to ensure a representative assessment of the overall surface quality. This distribution of measurement points was designed to provide a comprehensive representation of the surface texture, accounting for possible anisotropies introduced during the layer-by-layer deposition process. The seven selected points were:

- ✓ On the outer cylindrical surface along two diametrically opposed generators: *GS1* – left generator in the lower area of the specimen; *GS2* – left generator in the upper area of the specimen; *GDI* – right generator in the lower area of the specimen; *GD2* – right generator in the upper area of the specimen.
- ✓ On the inner cylindrical surface, along three generators positioned at 0°, 60°, and -60°: *IS0* – generator at 0°; *IS60* – generator at 60°; *IS-60* – generator at -60°.

Figure 2 illustrates a representative measurement area selected on each specimen’s surface for Ra evaluation, emphasizing the consistent methodology applied across all samples to minimize spatial variability in surface texture data. The measurement results for the Ra surface roughness parameter [ $\mu\text{m}$ ] are presented in Tables 6 and 7 below. The INSIZE roughness tester generates a comprehensive analysis report, capturing values for 19 distinct surface texture parameters, among which Ra was selected as the primary indicator for comparative evaluation in this study.

To statistically evaluate the influence of the selected process parameters on the surface roughness (Ra) values obtained from the seven measurement zones, a one-way Analysis of Variance (ANOVA) was conducted using the Anova: Single Factor tool available in Microsoft Excel. The analysis was performed independently for each of the three variables (Infill, Wall Line Count, and Default Thickness) with separate evaluations for the white and black PLA samples.

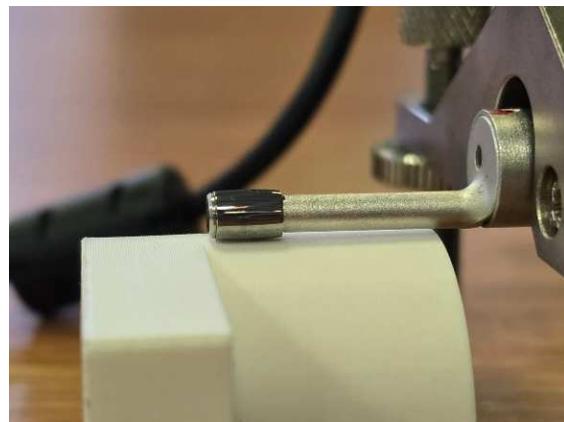
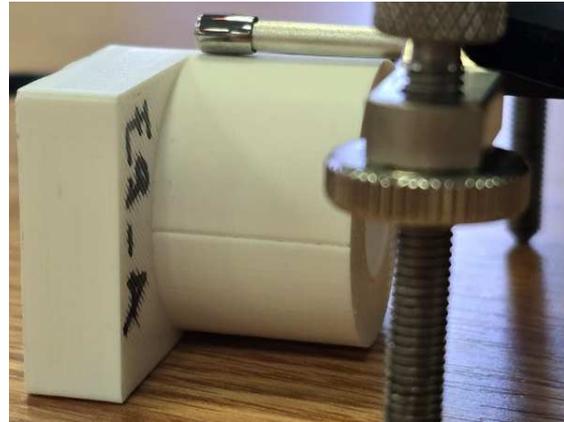


Fig. 2. One of the areas of sample measurement

Table 6

**Roughness values for white samples**

Ra [ $\mu\text{m}$ ]								
No.	Sample code	GS1	GS2	GDI	GD2	IS0	IS60	IS-60
1	E1_A	12.197	11.835	11.556	11.375	13.023	11.978	12.96
2	E2_A	11.524	11.914	10.786	11.413	11.834	12.773	13.292
3	E3_A	12.561	12.874	12.803	12.550	13.120	13.092	12.999
4	E4_A	10.785	11.463	10.978	12.094	12.576	12.644	13.113
5	E5_A	13.017	12.132	11.452	11.811	12.559	12.609	12.658
6	E6_A	12.157	11.854	13.681	11.919	12.162	12.71	11.299
7	E7_A	13.367	12.114	12.515	12.61	13.505	13.271	13.751
8	E8_A	12.547	11.315	12.483	12.479	12.612	13.228	13.251
9	E9_A	12.318	11.621	12.732	12.835	13.329	13.362	13.726
10	E10_A	11.2	11.709	11.487	11.819	13.186	12.598	13.923
11	E11_A	12.199	11.821	12.663	13.745	13.037	12.117	11.602
12	E12_A	11.841	13.632	12.423	12.503	12.404	13.676	13.413

The response values of the analyzed factors were divided into three groups (according to the variation levels – Table 1), each with 4 observations (according to the Taguchi L12 matrix – Table 2): *For Infill*: Group 1 – 15%, Group 2 – 40%, Group 3 – 100%; *For Wall Line Count*: Group 1 – 1, Group 2 – 2, Group 3 – 3; *For Default Thickness*: Group 1 – 0.45 mm, Group 2 – 0.67 mm, Group 3 – 1 mm.

Table 7  
Roughness values for black samples

Ra $\mu\text{m}$ ]								
No.	Sample code	GS1	GS2	GD1	GD2	ISO	IS60	IS-60
1	E1_N	11.35	12.217	11.637	11.493	11.282	13.136	12.061
2	E2_N	11.002	11.257	11.98	12.138	12.992	12.746	13.615
3	E3_N	11.973	12.329	12.019	12.014	12.986	12.316	13.713
4	E4_N	11.513	11.267	11.57	12.141	12.808	12.405	12.162
5	E5_N	13.746	11.471	11.136	10.473	12.244	12.739	12.617
6	E6_N	12.379	12.229	11.612	12.822	12.686	11.525	12.399
7	E7_N	13.013	11.889	11.835	11.679	12.737	12.681	12.042
8	E8_N	12.109	11.596	11.791	11.423	12.492	13.064	12.63
9	E9_N	11.771	12.347	11.813	11.349	12.714	13.344	13.420
10	E10_N	11.215	10.842	11.954	11.955	12.236	12.736	13.785
11	E11_N	11.541	11.502	11.637	12.061	12.381	12.75	11.981
12	E12_N	11.701	11.331	11.391	12.178	11.634	11.967	12.687

The resulting  $F$ -statistics and  $p$ -values for each surface measurement zone are presented in Tables 8, 9, and 10. These tables provide a comparative view of the statistical significance for each parameter across all sample groups and measurement zones. Statistical significance was determined based on a conventional threshold of  $p < 0.05$ . Values below this threshold are marked as significant, those between 0.05 and 0.10 as almost significant, and values above 0.10 as not significant.

ANOVA results for infill

Measure zone	F-stat White	p-value White	Interpretation White	F-stat Black	p-value Black	Interpretation Black
GS1	0.252	0.782	✗ Not significant	0.165	0.849	✗ Not significant
GS2	0.881	0.447	✗ Not significant	0.199	0.822	✗ Not significant
GD1	1.971	0.194	✗ Not significant	1.247	0.332	✗ Not significant
GD2	0.546	0.596	✗ Not significant	0.958	0.419	✗ Not significant
ISO	2.160	0.171	✗ Not significant	3.085	0.095	△ Almost significant
IS60	0.997	0.406	✗ Not significant	1.707	0.235	✗ Not significant
IS-60	4.488	0.044	☑ Significant	1.587	0.256	✗ Not significant

Based on the obtained ANOVA results, *Infill density* appears to have a limited global impact on  $Ra$  values. *Wall Line Count* and *Default Thickness* exhibit localized, material-dependent effects on surface roughness, particularly in

angled or top-layer measurement zones. These findings highlight the importance of parameter tuning not only for mechanical performance but also for surface quality optimization.

ANOVA results for Wall Line Count

Measure zone	F-stat White	p-value White	Interpretation White	F-stat Black	p-value Black	Interpretation Black
GS1	0.486	0.630	✗ Not significant	2.919	0.105	✗ Not significant
GS2	0.636	0.551	✗ Not significant	0.240	0.791	✗ Not significant
GD1	0.799	0.478	✗ Not significant	4.291	0.049	☑ Significant
GD2	0.101	0.904	✗ Not significant	1.678	0.240	✗ Not significant
ISO	0.161	0.853	✗ Not significant	2.928	0.104	✗ Not significant
IS60	0.509	0.617	✗ Not significant	0.867	0.452	✗ Not significant
IS-60	3.090	0.095	△ Almost significant	30.791	0	☑ Significant

ANOVA results for Default Thickness

Measure Zone	F-stat White	p-value White	Interpretation White	F-stat Black	p-value Black	Interpretation Black
GS1	6.262	0.019	☑ Significant	3.452	0.077	△ Almost significant
GS2	1.438	0.287	✗ Not significant	3.921	0.059	△ Almost significant
GD1	2.588	0.129	✗ Not significant	0.971	0.415	✗ Not significant
GD2	0.618	0.560	✗ Not significant	2.150	0.172	✗ Not significant
ISO	1.683	0.239	✗ Not significant	2.734	0.118	✗ Not significant
IS60	0.821	0.470	✗ Not significant	0.853	0.457	✗ Not significant
IS-60	2.841	0.110	✗ Not significant	1.608	0.252	✗ Not significant

#### 4. CONCLUSION

This study has demonstrated that among the key FDM process parameters investigated *Infill density* had the most consistent and statistically significant impact on sample weight, while its influence on  $Ra$  was generally limited and localized. Surface roughness analysis showed that *Wall line count* and *Default thickness* exhibited zone-specific effects, particularly in internal and angled areas, with notable variation between white and black PLA samples. These findings emphasize the importance of a localized and material-sensitive approach. By employing a structured DOE methodology and ANOVA, the study provides evidence-based insights into the complex interaction of FDM process

parameters. The localized influence of surface-related parameters such as wall line count and default thickness, particularly in internal and angled zones, highlights the directional sensitivity of surface finish in FDM-printed parts. The consistency of results across both filament colors confirms that pigment differences have minimal impact under controlled conditions. These findings support the need for zone-specific process calibration and encourage future investigations that integrate multi-criteria optimization and post-processing strategies for improved surface quality. The results contribute to the formulation of optimized parameter sets for enhancing the surface finish of PLA components.

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### Investigație experimentală a calității suprafeței la componente imprimate 3D din PLA utilizând DoE

Rugozitatea suprafeței reprezintă un parametru de calitate în cazul componentelor realizate prin imprimare 3D, influențând proprietățile mecanice, aspectul estetic și performanța funcțională. Acest studiu investighează influența principalilor parametri de proces asupra rugozității suprafeței pieselor imprimate 3D din PLA, utilizând o abordare bazată pe DOE. S-a dezvoltat o metodologie experimentală pentru a analiza efectele celor trei parametri analizați asupra calității suprafeței, cuantificată prin măsurători Ra. Experimentele au fost proiectate utilizând o matrice ortogonală Taguchi, cu scopul de a optimiza combinațiile de parametri și de a reduce numărul total de teste. Analiza datelor a fost realizată prin ANOVA și modelare de regresie, pentru a determina semnificația fiecărui factor și a interacțiunilor dintre aceștia. Studiul propune un set de parametri optimizați pentru îmbunătățirea rugozității, oferind informații pentru producători și cercetători interesați de creșterea calității suprafețelor pieselor realizate prin FDM.

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