Grinding Wheel Surface Texture Characterization Using Scale Sensitive Fractal Analysis.

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Abstract

The objective of this project is to characterize grinding wheel surface texture using conventional parameters and scale sensitive fractal analysis, for the purpose of differentiation. Combining this knowledge with grinding performance data will allow wheels to be designed and/or dressed to custom specifications. Replicas were made of grinding wheel surfaces and measured. Conventional and fractal parameters were calculated using software, and F-tests were performed to differentiate surface texture based on these parameters. The fractal method was found to differentiate surface textures better.

1. Introduction

1.1 Objectives

The aim of this research is to measure the surface texture of grinding wheels for the purpose of differentiating between wheel composition (abrasive, bond) and degree of wear (dressed, ground).

1.2 Rationale

A grinding wheel's surface texture has a strong influence upon its grinding performance. This is clearly evidenced by the increase in grinding forces, power consumption, and cutting zone temperatures as a wheel wears [Butler and Blunt, 2002]. If the surface texture of a grinding wheel could be measured and quantified, then it could be compared to known grinding performance data for similar texture. Grinding wheels could also be designed and/or dressed to take advantage of better textures leading to wheels that stay sharper longer.

Sharper wheels generally have high material removal rates and low grinding forces and power consumption. High material removal rates shorten cycle times, which allow more parts to be made in less time. Low grinding forces and power consumption reduces wear on grinding machinery and improves the surface finish of the work piece. Due to these properties it stands to reason that producing grinding wheels that remain sharp longer will save consumers time and money.

Understanding the connection between surface textures and grinding performance also benefits quality control. Currently wheels are tested by performing a grinding operation and examining the performance data, which generally consists of material removal rate, grinding force, power consumption, specific energy, and grinding ratio. This process is time consuming, destroys the product, and is an indirect method of characterizing the wheel.

Measuring the surface texture, however, would be faster and automatable, be completely non-contact, and give a direct characterization of the wheel. The wheel surface could easily be compared to an industry standard and assigned a value of how well it fits the standard, similar to the ratings of electrical resistors. This would all work together to reduce the time and money spent testing wheels, reduce the time to market, and allow for a guarantee of wheel performance.

1.3 State of the Art

Figure 1 shows a table outlining the sources described below, and the methods used in their research.

Names:	Measurement Technique:	Parameter Examined:
Zhou and Xi (2002)	Contact Stylus	Active cutting edges vs. Grinding power
Blunt and Ebdon	Contact Stylus	Static Cutting Points
Butler, Blunt, See, Webster,	Contact Stylus	Summit density and curvature
and Stout		

Figure 1: Outline of State of the Art references

Zhou and Xi (2002) developed a new analytical method for predicting the surface roughness of a grinded work piece for a variety of different grinding conditions. This was accomplished by applying the stochastic distribution model of grain protrusion heights to kinematic analyses. The surface texture of the grinding wheel was measured using a contact stylus, and the coinciding points on the trajectories of multiple grains were sorted consecutively from highest to lowest. This model of the grinding wheel surface texture was regressed with grinding performance data and the number of active cutting edges to predict the surface roughness of a work piece after grinding. They found that this model of the surface texture more accurately predicted the surface roughness of a work piece than previous models. This is important to our research because it is an attempt to relate the surface texture of a grinding wheel to aspects of grinding performance.

Blunt and Ebdon (1995) characterized the surface topography of grinding wheels in terms of static cutting points and static cutting grains using three-dimensional contact profilometry techniques. The advantage of having a three-dimensional plot over a twodimensional profile is discussed. Static cutting points and static cutting grains were quantified by producing highly magnified stereographic images of the grinding surface to

produce contour maps and cutting edges and grains were counted by hand. Blunt and Ebdon also discuss sampling strategies for the characterization of grinding wheel topographies and outline optimum sampling spacing criteria.

Butler, Blunt, See, Webster, and Stout (2002) examined the topographical change occurring on a conventional aluminum oxide wheel as it machined steel work pieces. The same three-dimensional contact profilometry techniques suggested by Blunt and Ebdon (1995) were employed. The team investigated how the grinding force, summit density, and summit curvature changed as a function of stock removal. They found that the grinding force had a brief first phase where the force steeply climbs to a maximum level, and then a prolonged phase where the force tapers off from the maximum value. The change between these two phases was found to occur when the force reached a point that deflects the machine to the point where the measured depth of cut is equal to the true depth of cut.

1.4 Approach

In this work six different aluminum carbide inside diameter grinding wheels were examined after dressing and after grinding. The surface texture of the grinding wheels was characterized by making replicas of the grinding surface and measuring them using a non-contact optical profilometry technique. The conventional surface texture parameters, a complete list of which can be found in figure 7, are calculated using Mountains software. The fractal properties of relative area and average texture depth of the surface textures are calculated using the surface metrology and fractal analysis software package

SFRAX. Scale based F-tests are performed to find the scales, if any, at which fractal properties become differentiable.

2. Methods

The methods used to obtain and analyze the data necessary to meet the objectives of this work are outline in the following sections. The results of the analyses are explained in full detail in the results section of this report. The subsequent flow chart shows the sequence of necessary steps to achieve the objectives, and is followed by a series of sections with a detailed description of each step.



Figure 2: Flow chart of sequence of events

2.1 Grinding

Saint Gobain Abrasives performed dressing and grinding on a Bryant OD/ID grinder. A sample of 6 inside diameter grinding wheels was tested. The wheels had

dimensions 1.15"x 0.39" x 1.0" and a grit number of 100, and were composed of an assortment of aluminum oxide abrasives and vitrified bond types shown in the following table:

Name	Wheel Code	Grade	Structure	Bond No	Grain No
Blue 1	428.1.3	L	10	Bond2	Grain6
Blue 2	927-4	L	10	Bond2	Grain2
Blue 3	927-5	L	7	Bond3	Grain3
White 1	428.3.3	L	10	Bond2	Grain5
White 2	927-2	L	10	Bond1	Grain1
White 3	928-4.1	L	10	Bond2	Grain4

Figure 3: Table of grinding wheels examined

The wheels were 14 mm wide and had an initial diameter of approximately 30 mm. The wheels were a brought up to 32,000 rpm, and cooled using Trim e-210 coolant diluted with 5% distilled water. The wheels were dressed using a CDP diamond roll dresser with a speed of 5,200 rpm and a dress lead of 32 mm. The wheels were used to grind a work piece 6.35 mm wide, an initial diameter of 33 mm, and a Rockwell C hardness of 61 Rc at an infeed rate of 0.06477 mm/s. Plunge grinding was conducted in climb mode and each wheel completed twenty 1.6 second runs for a cumulative grinding time of 32 seconds. Grinding power, as well as normal and tangential force, was measured using a Norton Field Instrumentation System (FIS), and recorded to an Excel spreadsheet. The change in work piece diameter was measured periodically using a bore gauge, and was used in conjunction with grinding power as well as normal and tangential force to calculate grinding performance parameters, specifically: material removal rate, specific energy, and grinding ratio. Surface roughness and waviness of the work piece were also measured periodically to further characterize grinding performance.

2.2 Replication

Measurements of the surface textures were made from replicas of the grinding wheels surfaces. Replication is the process of producing a mirror image of a surface by applying an impressionable material to a surface, allowing it to set and then separating it from the surface. The downside of using replicas is that they are not a direct measurement of the grind wheel surface, which means that surface texture information is either not captured by the replica material or information is sheared from the replica material when removed from the surface. The latter of which is illustrated in the following image:



Figure 4: Replica material left on grinding surface

Attempts to measure the grinding surface of grinding wheels directly using optical techniques proved unsuccessful due to inherent properties of bonded abrasives. The fact that some grains are more transparent than others makes it difficult for the light source to

find the top of the grain. High reflectivity of the grain also poses a problem by saturating the measurement equipment with light. Also, measurement equipment cannot easily transition from focusing on bond to focusing on grain. Using replicas removes all of these problems by providing an opaque, non-reflective image of the grinding surface that is made out of consistent material.

Replicas were prepared using Coltène President, a polyvinylsiloxane base and catalyst system generally used to make dental molds, which is commercially available from Coltène-Whaledent. The area on the wheel for replication was prepared by holding the wheel steady in a custom made fixture and applying a size 10 washer to the top of the wheel using scotch tape. A small amount of base material and catalyst, less than a gram of each, were mixed together at a one to one ratio. The replica material was then placed in the center of the washer and held under a 200g weight for a total time of 10 minutes.

In order to determine the quality of the replicas taken a number of tests were performed. The first test was performed as the replica was removed from the wheel. The ease with which the replica was removed from the grinding surface, the amount of material left behind on the grinding surface, and the apparent texture transferred to the replica material were noted. If any of these seemed to be irregular that replica was discarded and another was taken.

The second test was performed after all replicas had been made. Each wheel was placed under a high-resolution camera and highly magnified. The wheels were then visually inspected for replica material left behind on the grinding surface. If this amount was small then it was an indication that the texture transferred to the replica material was not greatly altered when the replica was removed. High-resolution images were also

taken of the replicas, and were visually compared to the images taken of the grinding wheel surface. If the two images resembled each other it was an indication that the replica closely approximated the grinding surface.

2.3 Measurement

Initially a Micromeasure made by Microphotonics, located at Saint Gobain Abrasives in Worcester MA, was used in an attempt to measure the surface texture of the grinding wheels directly. This method was abandoned for two reasons. The primary reason was that after completing a test to measure the noise recorded by the system it was deemed too high to obtain an accurate measurement at the desired scale of 10 microns. The results of this noise test are discussed in detail in the results section. The second reason was that the intensity of the light generated by the xenon bulb in the equipment was too great to measure the reflective grinding surface. For these reasons the measurements were made using replicas, and measured using the UBM scanning laser microscope located at Worcester Polytechnic Institute in Worcester MA.

Replicas were measured using a UBM scanning laser microscope, located in the surface metrology lab at Worcester Polytechnic Institute. The UBM uses a Keyence LC-2210 confocal point sensor, which reflects a laser light source off of a surface and through a detector pinhole to determine height information. The laser beam is shown through an objective lens that rapidly oscillates on a vertical axis. When the surface being measured crosses the focus of the lens the light intensity reaches its maximum value. Conversely, when the distance between the surface and the lens is greater than or less than the radius of curvature of the lens the reflected light reaching the pinhole is faint

and is not detected. Therefore a height measurement is only recorded when the maximum intensity of light goes through the pinhole. This process is illustrated in the following picture:



Image from http://www.solarius-inc.com/html/confocal.html

Figure 5: Measurement principal of a confocal point sensor

A test was performed on the UBM equipment to measure the internal and external noise experienced by the system while a measurement is being made. Because the measurements being made are on the order of microns, very small ambient vibrations and vibrations from the equipments motors could skew the final outcome of the measurements. Noise testing was performed by attached a 90 degree bracket to the height sensor in a manner that would allow a stationary point on the bracket to be measured. Arbitrary parameters were entered in the UBM software, and the equipment was left to run for several minutes. Therefore, any height data recorded by the UBM during this procedure was purely from noise affecting the system. This data was then

analyzed and the amount of noise was found to be negligible. The results of the noise test are discussed in further detail in the results section.

Measurement of the replicas was performed in batches. Each batch was first arranged in a systematic pattern that would easily allow files to be associated with their correct surface texture afterwards. Once on the table of the replicas were held in place using magnets. Then a check was made to confirm that the replica was situated near the center of the range of the height sensor by moving the height sensor and noting its uppermost and lowermost limits. Once the replicas were in the range of the sensor, the table was moved to align the upper left hand corner of each measurement area with the laser and the UBM software recorded the positions. The UBM was then able to begin measuring the replicas. The following table shows the parameters used for measurement:

	Parameter	Value
Area	Length Ground/Dressed	2.54 mm/2.0 mm
	Width Ground/Dressed	2.54 mm/2.0 mm
	Vertical Range	30 mm
	Step Size	10 µm
Light	Wavelength	780 nm
Source	Spot Diameter min/max	70-90 µm
	Pulse Width	12.5 µm
	Power	3 mW
Data	Sampling Rate	40 KHz
Acquisition	Response Frequency	16 KHz
	Response Time	100 µs
	Averaging	128 pts
	Measurement Rate	100 pixel/s
	Table Speed	1 mm/s

Figure 6: Table of Measurement Parameters

When the UBM finished measuring, the data files were renamed to correspond with their respective replica and saved in a .UB3 file format. The data was then manipulated by leveling it using the UBM software. This was done by performing a linear regression to remove any inherent slope or form from the measurement.

The data was then transferred into MountainsMap software to calculate the conventional parameters, which will be discussed in detail later in the methods. After the conventional parameters were calculated the format of the files was changed to .SUR to be compatible with the software used to calculate the fractal properties of the surface, which will be discussed later in the methods. The values obtained for the conventional parameters are discussed in detail in the results section.

To ensure that the UBM was in fact generating a clear portrayal of the textures captured by the replicas a test was performed to judge the equipments ability to reproduce a result. This was done simply by measuring a batch of replicas using the method previously outlined and then, without moving or reorienting any of the replicas, remeasuring the batch under the same conditions. Each surface was then compared to its counterpart from the other trial by using a height-height diagram. A height-height diagram works by plotting the height of every x-y position of one surface versus another surface. The closer the surfaces are the being the same the closer the points on the plot should align along the line y = x. The results of this test are discussed in detail in the results section.

2.5 Characterization

The surface files were characterized using two different methods. The following displays the list of conventional surface parameters that were calculated for every surface file using MountainsMap software:

Symbol	Description	Definition (ASME B46)
Sa	Average Roughness	$S_{a} = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} z(x_{k}, y_{l}) - \mu ^{\dagger}$
Sq	Root Mean Squared (RMS) Roughness	$S_{q} = \sqrt{\frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} \left[z(x_{k}, y_{l}) - \mu \right]^{2}}$
Sp	Maximum Peak Height	$S_p = Z_{max}^{\dagger}$
Sv	Maximum Valley Depth	$S_p = Z_{min}^{\dagger}$
St	Maximum Peak to Valley Distance	$S_t = \mathbf{Z}_{max} - \mathbf{Z}_{min}^{\dagger}$
Ssk	Surface Skewness	$S_{sk} = \frac{1}{MNS_q^3} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} \left[(z(x_k, y_l) - \mu)^3 \right]^{k}$
Sku	Surface Kurtosis	$S_{ku} = \frac{1}{MNS_q^4} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} \left[z(x_k, y_l) - \mu \right]^4^{\dagger}$
Sz	Ten Point Height	$S_{z} = \frac{\sum_{i=1}^{5} (z_{pi} - \mu) + \sum_{i=1}^{5} (z_{vi} - \mu)}{5}^{\dagger}$
Ра	Unfiltered Average Roughness	$Pa = \frac{1}{lb} \int_{lb} z(x) dx^*$
Pq	Unfiltered RMS Roughness	$Pq = \sqrt{\frac{1}{lb}\int_{lb}^{l} z^2(x) dx} $

† - Equation from http://www.imagemet.com/WebHelp/spip.htm#roughness_parameters.htm

* - Equation from http://www.digitalsurf.fr/en/guideparam2D.htm

Figure 7: Table of Conventional Parameters

The second method used to characterize the surface textures was scale sensitive fractal analysis. This method can be used to analyze linear profiles, surface area, and surface depth and volume, denoted length-scale, area-scale, and filling scale respectively. Length-scale can be described by asking the question, how long is the coast of California? One might first draw a line from the northern most point to the southern most point tip of the coastline, but this only accounts for one scale of observation. By breaking that line into progressively smaller segments of equal length more and more detail of the bays and coves come into view. As illustrated in the following figure:



From Recent Developments in Surface Metrology Using Fractal Analysis by Christopher A. Brown, Slide 17.

Figure 8: Length-Scale Illustration

Similarly, area-scale analysis is used to find the area of surface at progressively smaller scales. The apparent area is calculated by covering the surface with a patchwork of triangular tiles with progressively smaller areas, which is illustrated in the following figure:



From Recent Developments in Surface Metrology Using Fractal Analysis by Christopher A. Brown, Slide 94.

Figure 9: Area-Scale Illustration

The relative area at a particular scale is then calculated as the measured area (area of a single tile multiplied by the total number of tiles) divided by the nominal area (area of the x-y plane). Filling-scale analysis is performed in a similar fashion by replacing the triangular patches with rectangular prisms and adding their volumes together.

Fractal analysis of the surface textures was performed using SFRAX software. Once a surface file was loaded into the software alterations were performed on the data. First the data had to be inverted to account for the negative image of the grinding wheel surface produced by replication. Second the data was distorted by a factor of 9.95 to correct a calibration error inherent to the UBM measurement equipment.

Area-Scale analysis was performed to calculate the relative area of the surface texture of each replica measurement across a range of scales, using the four corners full overlap method. The area-scale curves produced were then grouped together by the wheel used to produce the replicas. As was stated previously 6 wheels of distinct bondgrain composition were measured 12 times each, 6 on the ground region and 6 on the dressed region, producing 72 unique area-scale curves.

Filling-Scale analysis was performed to calculate to average texture depth of the surface texture of each replica measurement across a range of scales, using the Volume Absolute analysis method. The filling-scale curves produced were then grouped together by the wheel used to produce the replicas. As was stated previously 6 wheels of distinct bond-grain composition were measured 12 times each, 6 on the ground region and 6 on the dressed region, producing 72 unique filling-scale curves.

2.5 Differentiation

Differentiation of the surface textures was done by performing F-tests. An F-test is a statistical method for comparing the difference in the standard deviation of 2 sets of data. Differentiation was performed for every conventional parameter listed in section 2.5, as well as for relative area and average texture depth.

For each parameter a matrix was filed out which compared the ground and dressed region of each wheel against one another systematically, as shown in the following figure:

B1D	B1D		_					B = Bl	ue Whe	el		
B2D		B2D		_				W = W	hite W	heel		
B3D			B3D		_			1,2,3 = Wheel Number				
B1U				B1U		_		D = D	ressed	Region		
B2U					B2U		_	U = G	round F	Region		
B3U						B3U		_				
W1D							W1D		_			
W2D								W2D		_		
W3D									W3D		_	
W1U										W1U		_
W2U											W2U	
W3U												W3U

Figure 10: Differentiation Matrix

Differentiation of the conventional parameters was performed using the FTEST function in Microsoft Excel. The function returns a value from 0 to 1, which is highest to lowest level of differentiation respectively. Any value returned below 0.5 was interpreted to be differentiable with 95% confidence.

Differentiation of the fractal parameters was performed using the F-test function in SFRAX at a 95% confidence level. The function would return a plot of mean square ratio as a function of scale, which would display graphically at what scales the two surfaces are differentiable. Any F-test displaying differentiability at any scale or range of scales finer than the smooth-rough crossover was interpreted as being differentiable with 95% confidence.

3. Results

3.1 Grinding

After grinding was performed on each of these wheels there were clear differences on the surface of the wheel. The width of the work piece was smaller than the width of the grinding wheel, so the dressed section was able to be visually distinguished from the ground section. The following picture shows the differences between the ground and dressed regions of the wheel.



Figure 11: Regions of the grinding surface

The grinding performance of the wheel is generally gauged by plotting the cumulative material removed as a function of the power consumed by the grinder. The greater the slope of the resulting line the more material it can remove while consuming less power. Below are the performance graphs for each of the wheels examined.



Figure 12: Performance graphs

These graphs clearly indicate that there is a large amount of variance in the performance of these wheels, which means there should be noticeable differences in each of the wheels surface textures.

The performance graphs were all measured in metric units. The Power was measured in W/mm and Cumulative Material Removed measured in mm³/mm. Some points were interpreted as outliers if there deviation from the average was more than twice the standard deviation and were removed from the graphs. These points are most likely measurement artifacts as the result of equipment error. These points generally occurred at the very beginning or very end of the tests. The data used to create these plots along with other performance data generated from the tests can be found in the appendix.

3.2 Replication

Below are highly magnified images of a grinding wheel surface and its corresponding replica surface.



Figure 13: Replica quality inspection

The wheels were then visually inspected for replica material left behind on the grinding surface. As can be seen in Figure 13 the amount of replica material left on the grinding surface is minimal, which is a good indication that the texture transferred to the replica material was not greatly altered when the replica was removed.

A visual inspection of the similarity in surface texture was also performed. It can be seen in Figure 13 that the texture transferred to the replica material closely approximates the texture of the grinding surface. Since the two images resemble each other it is a good indication that the replica is an accurate approximation the grinding surface.

3.3 Measurement

Figure 14 shows a comparison of the noise inherent to the Micromeasure located at Saint Gobain Abrasives and the UBM located at Worcester Polytechnic Institute.



Figure 14: Noise Comparison

The noise tests run on the Micromeasure and UBM scanning laser microscope were analyzed in SFRAX. Figure 14 is the result of area-scale analysis that was run on the surface files created by the noise tests, and shows the relative area as a function of scale. Ideally these plots should show a horizontal line at a relative area of 1.00 at all scales, which would be a perfectly flat surface. From these plots it is clearly shown that at the scale of $200\mu m^2$ the Micromeasure is measuring a high level of noise whereas the UBM is not. The Micromeasure measures a high level of noise because it set directly on a table with no means of damping ambient vibrations. The following pictures show three-dimensional representations to further characterize the noise collected during the noise tests.



Figure 15: Three-dimensional representations of noise test results

Figure 16 shows the height-height plot used to determine the repeatability of the measurements using the UBM scanning laser microscope.



Figure 16: Height-Height plot of repeatability tests

While measuring the replicas, it was important to know that the results could be repeated, which was accomplished by measuring the same surface twice and creating a height-height diagram. A height-height diagram works by plotting the height of every x-y position of one surface versus another surface. The closer the surfaces are to being the same the closer the points on the plot should align along the line y = x. Since the data in Figure 16 follows the form of the equation y = x it shows that the UBM is accurately measuring the surface of the replicas. Because the points on this graph are not random, this test also shows a low level of noise in the UBM equipment.

3.4 Characterization

Figure 17 displays the calculated average and standard deviation of every conventional parameter listed in Figure 7 broken down by wheel and region.

	P1	П	B3	חי	Ba	חי	P1		P2		B3	11	
Symbol		о е п		e n		e n		0 6 D		.U e n		ю е п	
Symbol	Avy	3. D.	Avg	3. D.	Avg	3. D.	Avg	3. D.	Avg	3. D.	Avy	3. D.	
Sa	2.22	0.18	1.60	0.12	1.19	0.18	2.48	0.39	2.05	0.18	1.94	0.27	
Sq	2.81	0.22	2.01	0.14	1.52	0.22	3.16	0.56	2.63	0.21	2.41	0.33	
Sp	7.25	1.00	5.77	0.62	4.38	0.38	8.40	1.13	6.95	0.93	6.38	0.49	
Sv	12.13	1.32	7.75	1.05	6.48	1.42	13.87	2.95	12.25	2.26	9.82	2.52	
St	19.38	1.83	13.52	1.38	10.87	1.71	22.27	3.29	19.20	2.18	16.20	2.91	
Ssk	-0.58	0.21	-0.41	0.16	-0.48	0.27	-0.57	0.37	-0.68	0.34	-0.45	0.23	
Sku	3.49	0.20	3.12	0.33	3.51	0.68	3.68	0.86	3.84	0.90	3.11	0.45	
Sz	14.73	1.16	10.44	1.64	8.36	0.86	15.43	2.00	13.35	2.07	11.38	2.12	
Ра	1.98	0.09	1.45	0.10	1.06	0.15	2.01	0.46	1.70	0.26	1.49	0.14	
Pq	2.45	0.11	1.80	0.12	1.33	0.17	2.48	0.58	2.13	0.32	1.86	0.18	
											14/0		
	VV'	D	VV2	20	VV3	W3D		W1U		W2U		W3U	
Symbol	Avg	S. D.	Avg	S. D.	Avg	S. D.	Avg	S. D.	Avg	S. D.	Avg	S. D.	
Sa	2.17	0.25	1.36	0.25	1.49	0.14	2.58	0.15	1.82	0.37	1.97	0.17	
Sq	2.74	0.32	1.72	0.31	1.88	0.15	3.24	0.20	2.28	0.46	2.44	0.22	
Sp	7.48	0.84	5.32	0.66	6.28	1.11	8.55	0.55	6.27	0.71	6.60	0.88	
							4 5 00	2 5 2	0.02	2/3	9.80	1.60	
Sv	12.62	3.53	6.90	1.49	7.43	1.16	15.00	3.5Z	9.03	2.40			
Sv St	12.62 20.10	3.53 4.08	6.90 12.22	1.49 1.63	7.43 13.62	1.16 1.47	15.00 23.55	3.52 3.85	9.03 15.30	2.40	16.40	2.06	
Sv <mark>St</mark> Ssk	12.62 20.10 -0.57	3.53 4.08 0.14	6.90 12.22 -0.42	1.49 <mark>1.63</mark> 0.19	7.43 13.62 -0.32	1.16 1.47 0.24	15.00 23.55 -0.59	3.85 0.24	9.03 15.30 -0.39	2.43 2.80 0.34	<mark>16.40</mark> -0.45	<mark>2.06</mark> 0.21	
Sv St Ssk Sku	12.62 20.10 -0.57 3.55	3.534.080.140.46	6.90 12.22 -0.42 3.23	1.49 1.63 0.19 0.24	7.43 13.62 -0.32 3.31	1.16 1.47 0.24 0.53	15.00 23.55 -0.59 3.59	3.85 0.24 0.79	9.03 15.30 -0.39 3.11	2.43 2.80 0.34 0.15	16.40 -0.45 2.96	2.06 0.21 0.33	
Sv St Ssk Sku Sz	12.62 20.10 -0.57 3.55 13.69	 3.53 4.08 0.14 0.46 2.56 	6.90 12.22 -0.42 3.23 9.31	 1.49 1.63 0.19 0.24 1.00 	7.43 13.62 -0.32 3.31 9.55	1.16 1.47 0.24 0.53 1.00	15.00 23.55 -0.59 3.59 17.48	 3.82 3.85 0.24 0.79 1.88 	9.03 15.30 -0.39 3.11 10.58	2.80 2.80 0.34 0.15 1.88	16.40 -0.45 2.96 11.17	2.06 0.21 0.33 0.58	
Sv St Ssk Sku Sz Pa	12.62 20.10 -0.57 3.55 13.69 1.93	 3.53 4.08 0.14 0.46 2.56 0.22 	6.90 12.22 -0.42 3.23 9.31 1.24	1.49 1.63 0.19 0.24 1.00 0.24	7.43 13.62 -0.32 3.31 9.55 1.34	1.16 1.47 0.24 0.53 1.00 0.09	15.00 23.55 -0.59 3.59 17.48 2.08	 3.82 3.85 0.24 0.79 1.88 0.11 	9.03 15.30 -0.39 3.11 10.58 1.36	2.43 2.80 0.34 0.15 1.88 0.13	16.40 -0.45 2.96 11.17 1.47	2.06 0.21 0.33 0.58 0.15	

Figure 17: Table of conventional parameter results

These results are used for the purpose of differentiating surface textures, the result of which will be discussed later in the differentiation section of the results. Figure 18 is an example of the area-scale curves of two distinct regions plotted on the same graph. Each region is designated by a different color red or black.



Figure 18: Comparison of area-scale curves

This graph shows that there is a clear difference in the relative area of the two regions. This implies that the regions will be differentiable. The complete set of areascale comparisons can be found in the appendix.

Along with area-scale analysis, filling scale analysis was also run on the measurements. Figure 19 shows a graph comparing the filling-scale curves of two distinct regions.



Figure 19: Comparison of filling-scale curves

This plot shows less distinction between the two data sets, which indicates that the regions will be harder to differentiate based on filling-scale analysis. The complete set of filling-scale comparisons can be found in the appendix.

3.5 Differentiation

The following figures (20-29) show the results of differentiation by conventional

parameters.

Sa								5%	Diffe	erent	iabil	ity	
B1D	B1D											_	
B2D	0.339	B2D											
B3D	0.919	0.391	B3D		_								
B1U	0.132	0.020	0.111	B1U									
B2U	0.947	0.372	0.973	0.117	B2U								
B3U	0.406	0.085	0.353	0.473	0.370	B3U							
W1D	0.525	0.122	0.463	0.361	0.484	0.840	W1D						
W2D	0.514	0.118	0.452	0.370	0.473	0.854	0.986	W2D					
W3D	0.554	0.708	0.623	0.044	0.599	0.165	0.228	0.222	W3D				
W1U	0.620	0.637	0.692	0.053	0.667	0.193	0.266	0.258	0.922	W1U			
W2U	0.151	0.024	0.127	0.938	0.135	0.521	0.402	0.412	0.051	0.061	W2U		
W3U	0.859	0.432	0.939	0.096	0.912	0.317	0.419	0.409	0.677	0.749	0.11	1 W3U	l
			Figu	re 20: \$	Sa diffe	rentiat	tion ma	trix					
60			8**					00/		1		• .	
Sq		1	8**					9%	Diffe	erent	iabil	ity	
Sq B1D	B1D							9%	Diffe	erent	iabil	ity	
Sq B1D B2D	B1D 0.340	B2D						9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D	B1D 0.340 0.992	B2D 0.335	B3D		1			9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D B1U	B1D 0.340 0.992 0.067	B2D 0.335 0.009	B3D 0.069	B1U				9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D B1U B2U	B1D 0.340 0.992 0.067 0.862	B2D 0.335 0.009 0.431	B3D 0.069 0.854	B1U 0.048	B2U			9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D B1U B2U B3U	B1D 0.340 0.992 0.067 0.862 0.410	B2D 0.335 0.009 0.431 0.086	B3D 0.069 0.854 0.416	B1U 0.048 0.280	B2U 0.322	B3U		9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D B1U B2U B3U W1D	B1D 0.340 0.992 0.067 0.862 0.410 0.433	B2D 0.335 0.009 0.431 0.086 0.093	B3D 0.069 0.854 0.416 0.439	B1U 0.048 0.280 0.263	B2U 0.322 0.342	<mark>B3U</mark> 0.967	W1D	9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D B1U B2U B3U W1D W2D	B1D 0.340 0.992 0.067 0.862 0.410 0.433 0.498	B2D 0.335 0.009 0.431 0.086 0.093 0.113	B3D 0.069 0.854 0.416 0.439 0.504	B1U 0.048 0.280 0.263 0.222	B2U 0.322 0.342 0.397	<mark>B3U</mark> 0.967 0.881	W1D 0.913	9%	Diffe	erent	iabil	ity	
Sq B1D B2D B3D B1U B2U B3U W1D W2D W3D	B1D 0.340 0.992 0.067 0.862 0.410 0.433 0.498 0.371	B2D 0.335 0.009 0.431 0.086 0.093 0.113 0.950	B3D 0.069 0.854 0.416 0.439 0.504 0.366	B1U 0.048 0.280 0.263 0.222 0.011	B2U 0.322 0.342 0.397 0.467	<mark>B3U</mark> 0.967 0.881 0.097	W1D 0.913 0.104	9% W2D 0.127	Diffe W3D	erent	iabil	ity	
Sq B1D B2D B3D B1U B2U B3U W1D W2D W3D W1U	B1D 0.340 0.992 0.067 0.862 0.410 0.433 0.498 0.371 0.851	B2D 0.335 0.009 0.431 0.086 0.093 0.113 0.950 0.440	B3D 0.069 0.854 0.416 0.439 0.504 0.366 0.843	B1U 0.048 0.280 0.263 0.222 0.011 0.047	B2U 0.322 0.342 0.397 0.467 0.988	B3U 0.967 0.881 0.097 0.315	W1D 0.913 0.104 0.335	9% <u>0.127</u> 0.389	W3D 0.476	erent W1U	iabil	ity	
Sq B1D B2D B3D B1U B2U B3U W1D W2D W3D W1U W2U	B1D 0.340 0.992 0.067 0.862 0.410 0.433 0.498 0.371 0.851 0.143	B2D 0.335 0.009 0.431 0.086 0.093 0.113 0.950 0.440 0.023	B3D 0.069 0.854 0.416 0.439 0.504 0.366 0.843 0.146	B1U 0.048 0.280 0.263 0.222 0.011 0.047 0.677	B2U 0.322 0.342 0.397 0.467 0.988 0.106	B3U 0.967 0.881 0.097 0.315 0.497	W1D 0.913 0.104 0.335 0.472	9% <u>W2D</u> 0.127 0.389 0.409	W3D 0.476 0.026	w1U 0.103	iabil W2U	ity	
Sq B1D B2D B3D B1U B2U B3U W1D W2D W3D W1U W2U W3U	B1D 0.340 0.992 0.067 0.862 0.410 0.433 0.498 0.371 0.851 0.143 0.996	B2D 0.335 0.009 0.431 0.086 0.093 0.113 0.950 0.440 0.023 0.343	B3D 0.069 0.854 0.416 0.439 0.504 0.366 0.843 0.146 0.987	B1U 0.048 0.280 0.263 0.222 0.011 0.047 0.677 0.067	B2U 0.322 0.342 0.397 0.467 0.988 0.106 0.867	B3U 0.967 0.881 0.097 0.315 0.497 0.407	W1D 0.913 0.104 0.335 0.472 0.430	9% 0.127 0.389 0.409 0.494	W3D 0.476 0.026 0.374	W1U 0.103 0.855	iabil W2U	ity 2W3U	

Sn									-			
эр		1						3%	Diffe	erent	iabil	ity
B1D	B1D		I									
B2D	0.315	B2D		1								
B3D	0.055	0.318	B3D									
B1U	0.781	0.205	0.032	B1U								
B2U	0.879	0.390	0.074	0.668	B2U		1					
B3U	0.148	0.634	0.592	0.090	0.190	B3U						
W1D	0.723	0.507	0.107	0.529	0.839	0.262	W1D		1			
W2D	0.380	0.893	0.260	0.253	0.465	0.543	0.595	W2D				
W3D	0.821	0.223	0.036	0.958	0.706	0.099	0.563	0.275	W3D		1	
W1U	0.224	0.824	0.433	0.142	0.283	0.799	0.379	0.721	0.155	W1U		
W2U	0.484	0.751	0.195	0.333	0.582	0.431	0.727	0.854	0.359	0.590	W2U	
W3U	0.791	0.453	0.091	0.588	0.910	0.228	0.929	0.536	0.624	0.334	0.661	W3U
			Figu	re 22: S	Sp diffe	erentia	tion ma	ıtrix				
Sv			9					9%	Diffe	erent	iabil	ity
B1D	B1D											-
B2D	0.626	B2D										
B3D	0.876	0.522	B3D									
B1U	0.102	0.040	0.134	B1U								
B2U	0.263	0.118	0.331	0.572	B2U							
B3U	0.183	0.077	0.235	0.736	0.818	B3U						
W1D	0.050	0.019	0.068	0.707	0.352	0.479	W1D					
W2D	0.797	0.460	0.919	0.160	0.382	0.274	0.082	W2D				
W3D	0.787	0.827	0.671	0.062	0.171	0.115	0.030	0.600	W3D		_	
W1U	0.051	0.019	0.068	0.709	0.353	0.480	0.998	0.083	0.030	W1U		
W2U	0.208	0.090	0.265	0.678	0.880	0.938	0.433	0.308	0.132	0.434	W2U	
W3U	0.684	0.376	0.801	0.205	0.467	0.342	0.108	0.880	0.501	0.109	0.382	W3U
			Figu	re 23: S	Sv diffe	erentia	tion ma	trix				
St			0					5%	% Dif	fere	ntiab	ilitv
B1D	B1D											J
B2D	0.551	B2D										
B3D	0.884	0.651	B3D									
B1U	0.223	0.079	0.176	B1U								
B2U	0.705	0.335	0.602	0.390	B2U							
B3U	0.331	0.127	0.267	0.794	0.545	B3U]					
W1D	0.102	0.032	0.078	0.647	0.196	0.474	W1D					
W2D	0.806	0.724	0.921	0.149	0.535	0.229	0.065	W2D				
W3D	0.643	0.893	0.749	0.101	0.404	0.160	0.042	0.826	W3D			
W1U	0.128	0.042	0.099	0.739	0.239	0.554	0.899	0.082	0.054	W1U		
W2U	0.371	0.146	0.301	0.732	0.599	0.935	0.427	0.260	0.183	0.502	W2U	
W3U	0.799	0.398	0.689	0.328	0.902	0.467	0.160	0.618	0.475	0.197	0.518	W3U
			Figu	re 24: \$	St diffe	rentia	tion ma	trix				

Ssk		0% Differentiability
B1D	B1D	
B2D	0.570 B2D	
B3D	0.593 0.278 B3D	
B1U	0.258 0.099 0.539 B1U	
B2U	0.335 0.135 0.658 0.861 B2U	
B3U	0.836 0.441 0.743 0.351 0.444 <mark>B3</mark>	30
W1D	0.393 0.770 0.175 0.057 0.080 0.	293 W1D
W2D	0.832 0.720 0.458 0.184 0.244 0.	.675 0.517 W2D
W3D	0.798 0.413 0.779 0.375 0.473 0.	.961 0.272 0.641 W3D
W1U	0.785 0.404 0.792 0.384 0.483 0.	.948 0.265 0.629 0.986 W1U
W2U	0.334 0.135 0.657 0.863 0.998 0.	.443 0.080 0.243 0.471 0.482 W2U
W3U	0.942 0.620 0.545 0.231 0.301 0.	.779 0.433 0.889 0.743 0.730 0.300 W3U
	Figure 25: Ssk differe	entiation matrix
Sku		26% Differentiability
B1D	B1D	· · ·
B2D	0.311 B2D	
B3D	0.019 0.135 B3D	
B1U	0.006 0.053 0.612 B1U	
B2U	0.005 0.045 0.552 0.929 B2U	
B3U	0.104 0.507 0.382 0.177 0.152 83	3U
W1D	0.097 0.484 0.403 0.188 0.162 0.	.970W1D
W2D	0.716 0.508 0.039 0.014 0.011 0.	.195 0.183 W2D
W3D	0.052 0.306 0.610 0.316 0.277 0.	.710 0.738 0.103 W3D
W1U	0.009 0.074 0.739 0.861 0.792 0.	.234 0.249 0.020 0.403 W1U
W2U	0.528 0.111 0.005 0.002 0.001 0.	.031 0.029 0.325 0.014 0.002 W2U
W3U	0.291 0.964 0.146 0.058 0.049 0.	.536 0.511 0.480 0.327 0.080 0.102 W3U
	Figure 26: Sku differe	entiation matrix
Sz		12% Differentiability
B1D	B1D	y
B2D	0.462 B2D	
B3D	0.534 0.184 B3D	
B1U	0.255 0.674 0.088 B1U	
B2U	0.229 0.625 0.078 0.946 B2U	
B3U	0.210 0.586 0.070 0.901 0.955 B3	3U
W1D	0.106 0.351 0.032 0.602 0.649 0.	.690 <mark>W1D</mark>
W2D	0.749 0.297 0.761 0.151 0.135 0.	122 0.059 W2D
W3D	0.753 0.299 0.756 0.153 0.136 0.	123 0.059 0.995 W3D
W1U	0.309 0.770 0.112 0.896 0.843 0.	799 0.516 0.188 0.190 W1U
W2U	0.310 0.771 0.112 0.895 0.842 0.	.798 0.515 0.189 0.190 0.999 W2U
W3U	0.151 0.038 0.395 0.016 0.014 0.	012 0.005 0.254 0.252 0.021 0.021 W3U
	Figure 27: Sz differe	ntiation matrix

Pa (AVG)								15%	% Dif	ferer	ntiabil	ity
B1D	B1D											
B2D	0.853	B2D										
B3D	0.317	0.411	B3D									
B1U	0.003	0.004	0.025	B1U								
B2U	0.037	0.053	0.231	0.239	B2U							
B3U	0.368	0.471	0.917	0.020	0.196	B3U						
W1D	0.072	0.101	0.382	0.137	0.734	0.330	W1D					
W2D	0.055	0.078	0.313	0.174	0.843	0.267	0.887	W2D		_		
W3D	0.974	0.827	0.302	0.003	0.035	0.352	0.067	0.051	W3D			
W1U	0.647	0.784	0.579	0.008	0.089	0.652	0.163	0.128	0.624	W1U		
W2U	0.441	0.556	0.811	0.015	0.157	0.893	0.270	0.217	0.423	0.751	W2U	
W3U	0.281	0.368	0.935	0.030	0.263	0.852	0.426	0.351	0.268	0.525	0.748 <mark>V</mark>	V3U
			Figur	- 28∙ P	a differ	entiati	on mat	rix				
			rigui		a uniter	unuau	on mai	1 1/5				
Pq (AVG)			rigui	. 20. 1	i unici	ciitiati	on mut	18%	6 Dif	ferer	ntiabil	ity
Pq (AVG) ^{B1D}	B1D		rigui	. 20. 1	<u>a unici</u>	ciitiati	<u>on mut</u>	18%	6 Dif	ferer	ntiabil	ity
Pq (AVG) B1D B2D	B1D 0.883	B2D	Figur		<u>a unici</u>	Circut	<u>on mut</u>	18%	⁄6 Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D	B1D 0.883 0.388	<mark>B2D</mark> 0.472	B3D		<u>unici</u>	Cittati	<u>on mut</u>	18%	6 Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U	B1D 0.883 0.388 0.003	B2D 0.472 0.004	B3D 0.018	B1U		Cittati	<u>on mut</u>	18%	6 Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U	B1D 0.883 0.388 0.003 0.039	B2D 0.472 0.004 0.052	B3D 0.018 0.193	B1U 0.218	B2U			18%	6 Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U	B1D 0.883 0.388 0.003 0.039 0.328	B2D 0.472 0.004 0.052 0.403	B3D 0.018 0.193 0.904	B1U 0.218 0.023	B2U 0.234	B3U		18%	6 Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U W1D	B1D 0.883 0.388 0.003 0.039 0.328 0.069	B2D 0.472 0.004 0.052 0.403 0.090	B3D 0.018 0.193 0.904 0.301	B1U 0.218 0.023 0.135	B2U 0.234 0.774	B3U 0.358	W1D	18%	% Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U W1D W2D	B1D 0.883 0.388 0.003 0.039 0.328 0.069 0.066	B2D 0.472 0.004 0.052 0.403 0.090 0.087	B3D 0.018 0.193 0.904 0.301 0.293	B1U 0.218 0.023 0.135 0.140	B2U 0.234 0.774 0.788	B3U 0.358 0.349	W1D 0.985	18%	% Dif	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U W1D W2D W3D	B1D 0.883 0.388 0.003 0.039 0.328 0.069 0.066 0.788	B2D 0.472 0.004 0.052 0.403 0.090 0.087 0.678	B3D 0.018 0.193 0.904 0.293 0.263	B1U 0.218 0.023 0.135 0.140 0.001	B2U 0.234 0.774 0.788 0.023	B3U 0.358 0.349 0.218	W1D 0.985 0.041	18% W2D 0.039	% Dif W3D	ferei	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U W1D W2D W3D W1U	B1D 0.883 0.388 0.003 0.039 0.328 0.069 0.066 0.788 0.398	B2D 0.472 0.004 0.052 0.403 0.090 0.087 0.678 0.482	B3D 0.018 0.193 0.904 0.301 0.293 0.263 0.986	B1U 0.218 0.023 0.135 0.140 0.001 0.001	B2U 0.234 0.774 0.788 0.023 0.187	B3U 0.358 0.349 0.218 0.890	W1D 0.985 0.041 0.294	W2D 0.039 0.285	% Dif W3D 0.270	ferer W1U	ntiabil	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U W1D W2D W3D W1U W2U	B1D 0.883 0.388 0.003 0.039 0.328 0.069 0.066 0.788 0.398 0.398	B2D 0.472 0.004 0.052 0.403 0.090 0.087 0.678 0.482 0.655	B3D 0.018 0.193 0.904 0.301 0.293 0.263 0.986 0.781	B1U 0.218 0.023 0.135 0.140 0.001 0.017 0.010	B2U 0.234 0.774 0.788 0.023 0.187 0.120	B3U 0.358 0.349 0.218 0.890 0.691	W1D 0.985 0.041 0.294 0.196	W2D 0.039 0.285 0.190	% Dif W3D 0.270 0.393	fere <u>W1U</u> 0.795	ntiabil W2U	ity
Pq (AVG) B1D B2D B3D B1U B2U B3U W1D W2D W3D W1U W2U W3U	B1D 0.883 0.388 0.003 0.039 0.328 0.069 0.066 0.788 0.398 0.398 0.355	B2D 0.472 0.004 0.052 0.403 0.090 0.087 0.678 0.482 0.482 0.655 0.434	B3D 0.018 0.193 0.904 0.301 0.293 0.263 0.986 0.781 0.948	B1U 0.218 0.023 0.135 0.140 0.001 0.017 0.010 0.020	B2U 0.234 0.774 0.788 0.023 0.187 0.120 0.214	B3U 0.358 0.349 0.218 0.890 0.691 0.955	W1D 0.985 0.041 0.294 0.196 0.331	W2D 0.039 0.285 0.190 0.322	% Dif W3D 0.270 0.393 0.238	feren W1U 0.795 0.935	W2U 0.732	ity V3U

The differentiation was performed using the FTEST function in Microsoft Excel. The function returned a value from 1 to 0, which can be seen in each cell of the matrices. A cell containing a number less than or equal to 0.05 was interpreted to be differentiable and is highlighted in green. The percent of surface textures that were differentiable is also shown in the upper right-hand corner of each figure. From these matrices it can be determined that kurtosis is the best conventional parameter to use for differentiation while skewedness is the worst. Combining all the differentiable results from all the conventional parameter matrices produces the matrix in Figure 30.



This figure shows that even though each conventional parameter can only differentiate a small amount of the surface textures there is not much overlap in the textures they differentiate.

Figure 31 shows examples of F-tests performed on two sets of area-scale curves showing the mean square ratio of both as a function of scale.



Figure 31: F-test results of area-scale comparisons

Any point of the graph that lies above the horizontal bar is differentiable with 95% confidence. A differentiation matrix was filled out that shows the result of each F-test comparing all regions of all wheels, and can be seen in Figure 32.

Differentiable at:												
B1D	B1D		_	Some Sca								
B2D	<10 ⁵	B2D		No Scales								
B3D	<10 ⁵	<10 ⁵	B3D									
B1U	<10 ⁵		10 ⁴ -10 ⁵	B1U		_						
B2U	10 ³ , 10 ⁵	2*10 ³ -10 ⁵	<10 ⁵		B2U		_					
B3U	<10 ⁵		<10 ⁵		2*10 ⁴	B3U		_				
<mark>W1D</mark>		<10 ⁵	<10 ⁵	10 ² -2*10 ³ , 2*10 ⁴	10 ⁵	<10 ⁵	W1D					
W2D	<10 ⁵	<10 ⁵	<10 ⁵		<10 ⁵	<10 ⁵	<10 ⁵	W2D				
W3D	<10 ⁵	<10 ⁵	10 ⁴ -10 ⁵		<10 ⁵	<10 ⁵	<10 ⁵	2*10 ² -10 ³ ,10 ⁵	W3D			
<mark>W1U</mark>		<10 ⁵	<10 ⁵	<10 ⁵	<10 ⁵	<10 ⁵	10 ⁵	<10 ⁵	<10 ⁵	W1U		
W2U	<10 ⁵	10 ⁵	10 ³ -10 ⁵		10 ³ -10 ⁵	3*10 ⁴ -10 ⁵	<10 ⁵	10 ⁵	10 ⁴ -10 ⁵	<10 ⁵	W2U	
W3U	<10 ⁵	10 ² , 10 ⁵	<10 ⁵		10 ² -10 ⁵	5	<10 ⁵	<10 ⁵	<10 ⁵	<10 ⁵	10 ⁵	W3U
Figure 32: Relative area differentiation matrix												

Figure 32 clearly shows that relative area is a much better way to differentiate surface textures than the best conventional parameter. Green cells represent comparisons in which the F-test produced results that were differentiable at all scales below the smooth rough crossover at $10^5 \,\mu\text{m}^2$, while dark green cells represent comparisons where the F-test produced results that were differentiable only at a particular scale or range of scales that is contained within the cell.

Figure 33 shows examples of F-tests performed on two sets of filling-scale curves showing the mean square ratio of both as a function of scale.



Figure 33: F-test results of a filling-scale comparison

The height of the bar denoting 95% confidence in differentiability indicates that it

is much harder to differentiate surface textures using filling-scale analysis. Figure 34

shows the complete differentiation matrix for filling-scale analysis.

47% Differentiability										Differentiable at:			
B1D	B1D									All Scales			
B2D	<10^2	B2D		_						Son	ne Sca	lles	
B3D	<10^2		B3D		_					No	Scale	es	
B1U				B1U		_							
B2U	<10^2	<10^2	<10^2		B2U		_						
B3U		<10^2	<10^2			B3U		_					
W1D		<10^2	<10^2				W1D		_				
W2D	<10^2				<10^2	<10^2	<10^2	W2D		_			
W3D	<10^2				<10^2	<10^2	<10^2		W3D		_		
W1U	<10^2	<10^2	<10^2		<10^2	<10^2		<10^2	<10^2	W1U		_	
W2U	<20									<10^2	W2U		
W3U	<10^2		<10^2					<10^2	<10^2	<10^2		W3U	
Figure 34: Average texture depth differentiation matrix													
This matrix shows that average texture depth does not do as good a job as relative area at differentiating the surface textures. One advantage that it does have is that textures are either differentiable or they are not. The results do not change with scale like they do with the relative area F-tests. This can be advantageous in the sense that the wheels do not have to be measured at a specific scale to be differentiated using this method.

Figure 35 shows the combined differentiability of relative area and average texture depth.



This figure shows that there is some improvement in the success rate of differentiation using fractal techniques. However, many of the surface textures differentiable by average texture depth are already differentiable by relative area. Only 2 surface comparisons were added by superimposing the two matrices.

4. Conclusions

- 1. Because the performance of the grinding wheels decay in a linear fashion, sharpness of the surface texture must decay in a similar fashion.
- The type of inside diameter grinding wheel used in this research cannot be measured directly using a chromatic white light profilometer or scanning laser microscope.
- 3. The replica material can be used to replicate surfaces at a scale of 10 microns, and can successfully capture features of the surface texture at this scale.
- 4. Fractal techniques, especially relative-area, can be used to differentiate surfaces with a much higher success rate than conventional parameters.
- 5. The difference in differentiability by relative area/average texture depth at different scales could be due to features on the surface that are prominent at those scales and account for the difference in grinding performance.

5. Discussion

Measuring a grinding wheels performance by plotting the cumulative material removed as a function of power shows a high linear correlation. This implies that the surface texture of a grinding wheel must also wear in a linear fashion, and some link must exist between the starting texture and ending texture of the wheel. In order to find this link relative area and average texture depth must be correlated with the grinding performance at differentiable scales.

Originally the approach to measuring the surface textures of the grinding wheels was to use a chromatic white light profilometer. Aside from technical difficulties and scheduling conflicts with the equipment, this equipment could not be used due to problems encountered in measuring grinding surfaces directly, which are attributed to inherent properties of bonded abrasives. One of these properties is the reflective nature of the grains, which would reflect enough light back into the equipment to saturate the reading even at the lowest intensity levels. Also, the transparency of some grains makes it difficult for the light source to find the actual surface. The grinding surfaces could not be measured directly on the UBM scanning laser microscope either due to its inability to adjust to the changes in focus caused by the light moving between bond and grain.

When it was realized that the grinding wheel surfaces were not able to be measured directly using a chromatic white light profilometer or scanning laser microscope, it was decided that replication of the surface textures of the grinding wheels was the next best option, but it was uncertain whether or not the material used to make replicas was going to be able to capture enough detail of the surface at very small scales.

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It was determined retrospectively that the replicas were providing information at a scale of 10 microns by examining the relative area plots. If there were a decrease in the amount of new area covered by triangular patches as the tiling routine moved to a sequentially smaller scale, then there would be a noticeable decrease in the slope of relative-area curve at the scale this would begin to occur. Because the slopes of the relative area curves all remain increasing to the finest measurable scale it can be inferred that detail of the surface textures at these scales was captured by the replica material.

The best differentiation using conventional parameters was found using kurtosis, which produced a 26% success rate. This is in stark contrast to average texture depth, which had a 47% success rate, and relative area, which produced an 83% success rate. Even superimposing every conventional parameter matrix still only produced a 59% success rate. This can be attributed to the fact that each conventional parameter only examines one specific detail of a surface. This is like trying to differentiate waves by looking at their amplitude. Although many waves can have similar peak heights their wavelengths could vary by miles, yet by only examining the amplitude would be considered non-differentiable. Fractal techniques, however, examine all aspects of the surface texture at the same time, and consequently do not produce such high levels of erroneous results.

Another added benefit to using fractal analysis to differentiate surface texture is that the differentiation is broken down by scale. In the case of conventional parameter differentiation, surface textures are either differentiable of they are not. This is not the case with fractal parameters. Some surfaces are only differentiable at a certain scale or a range of scales. This shows that there is a specific feature or group of features at those

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scales that is very different between the two surface textures. Surface features that exist at these highly differentiable scales are most likely responsible for differences in grinding performance.

6. References

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7. Appendices

Appendix A

Relative-area plots






































































Appendix B Average texture-depth plots






































































Appendix C F-test Plots

B1DVSB1U



B1DVSB2U



BIDVSB2D



B1DVSB3D



B1DVSB3U







B1DVSW1U



B1DVSW2D



B1DVSW2U



B1DVSW3D



B1DVSW3U



B1UVSB2U



B1UVSB3U



B1UVSW1D



B1UVSW1U



B1UVSW2D



B1UVSW2U



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B3DVSB1U







B3DVSB3U



B3DVSWID







B3DVSW2D







B3DVSW3D



B3DVSW3U



B3UVSW1D



B3UVSW1U



B3UVSW2D







B3UVSW3D







W1DVSW1U



W1DVSW2D



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W1DVSW3U



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