Laser-based assessment of road aggregate particle shape and texture properties with the aim of deriving comparative models

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ABSTRACT

Research was undertaken using an innovative three dimensional (3D) laser scanning

tool to study the shape and texture characteristics of road aggregate particles.

Aggregate materials used for road construction, including G1 crushed rocks of

different geological origins, recycled aggregate, and alluvial gravel (not used as

aggregate) were used for this study. Representative samples were scanned using the

laser system to collect 3D aggregate data for analyses and subsequently, develop

comparative models.

The objective was to arrange the aggregate particles in a sequence based on their

surface texture. Two models were proposed and key aspects evaluated against each

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other. Ultimately, one model was selected that may be improved and used for further research. The study found that while it is possible to use the 3D aggregate scan data to produce comparative models, distinguishing between particle shape and texture proved a daunting task and that particle elongation must be considered as a major influencing factor.

INTRODUCTION

The study forms part of a PhD research (Breytenbach, 2013), which is aimed at deriving a tabulated range of particles that could graphically or numerically represent a range of aggregate shape and texture properties. The proposed model, if validated, should be able to arrange particles in a certain sequence based on the texture properties. The method would lend scientific substantiation to particle shape characterisation instead of visual observations, which have previously (to a larger extent) used to quantify aggregate particle shape and texture.

An innovative study using three dimensional laser scanning technology in the application of road materials is under development at CSIR. The laser scanning method is applied to road construction aggregate using different approaches, attempting to study aggregate particle properties in great detail and for different applications. Emphasis has thus far been placed on particle shape and surface properties (Anochie-Boateng *et al.*, 2010) and the flakiness index (Anochie-Boateng *et al.*, 2011a) amongst others.

In this paper, the laser scanning method was applied in an attempt to refine the description of particle textures and shapes through two comparative proposed models considering that these models seek to sort aggregate particle data obtained from coarse/angular to smooth/rounded particles.

EXISTING METHODS

A number of advanced methods have been applied to study particle shape properties of aggregates. Rao *et al.* (2002) attempted to quantify the angularity of particles and ultimately showed that their approach (using image analysis) could distinguish between rounded gravel and crushed stone. Fletcher *et al.* (2002) discuss an aggregate imaging system (AIMS), using a back-lit system. It is clear that this method is able to determine the angularity of coarse and fine aggregate, as well as relative dimensions, thereby enabling the identification of flat or elongated particles. Other back-lit systems have also been used (Fernlund, 2005; Ken *et al.*, 2009), while Descantes *et al.* (2006) and Bouquety *et al.* (2006) used videographers in combination with a backlit system.

Kim *et al.* (2003) analysed materials using a laser-based approach. The digital image technology (DIT) calculates the volume of a particle, as well as the minimum square aperture through which a particle could fit.

However, the approaches discussed above are based on imaging techniques, which could at best capture two dimensional (2D) physical attributes of the aggregate

particles. In reality, aggregate particles are three dimensional; accordingly any improved techniques or methods should be able to address the physical properties in three dimensions.

The approach used in this research is based on 3D scanning techniques which were recently introduced by the CSIR (Anochie-Boateng *et al.*, 2010; Anochie-Boateng *et al.*, 2011a; Anochie-Boateng *et al.*, 2011b). Details will not be repeated in this paper.

METHODOLOGY

Scanning Procedure

Scanning of aggregate particles is done according to a protocol developed specifically for the laser scanner at the CSIR (Anochie-Boateng and Komba, 2010). The approach adopted consists of four steps which can be summarised as follows:

- 1. Scanning: The first step entails physical processing of the particle by means of scanning. During the first phase scanning four faces of a particle, which is placed on a rotating table in the scanner, are scanned. The particle is then rotated in order to scan the remaining two faces which were not in line of sight during the first scan phase (i.e. top and bottom).
- 2. Alignment: During this step the data obtained from the two scanning phases are orientated by the user. In essence this step simply rotates and moves the two scan dataset into the correct orientation relative to each other.
- 3. Combining: Software is used to combine the correctly orientated data.

4. Merging: The combined data is merged into a single object and saved as a single particle presented in three dimensions. Any refinement (e.g. removal of small imperfections or overlapping areas etc.) is also executed during this step.

The scanning density used by the scanning equipment can be varied according to the user's requirements. The density used affects the resolution of the final result and also the amount of time required to scan the particle. A higher scan resolution results in longer scan time. The maximum scan resolution of the equipment is 0.1mm. The time required to scan a particle is also affected by the size of a particle.

Particle Selection

The aggregate particles and materials used during this study is part of other on-going projects running congruently at the CSIR. The sourced aggregates used in this study included the following geological materials:

- Quartzite (G1 aggregate from stockpile) Magaliesberg Formation, Pretoria
 Group, collected in Pretoria, Gauteng
- Granite (G1 aggregate from stockpile) Johannesburg Dome, collected in Midrand, Gauteng
- Tillite (G1 aggregate from stockpile) Dwyka Group, Karoo Supergroup,
 collected in Verulam, Kwazulu-Natal
- Hornfels (G1 aggregate from stockpile) Tygerberg Formation, Malmesbury
 Group, collected in Durbanville, Western Cape

- Dolerite (G1 aggregate from stockpile) Karoo Supergroup, collected in Trichardt, Mpumalanga
- Recycled aggregate National Asphalt plant in Durban, Kwazulu-Natal
- Alluvial gravel Quaternary aged surface deposit sampled from the river bed
 of the Molopo River, some 120km west of Mafikeng, North West. This
 material is not an aggregate source

Each material was screened (using a sieve stack) into different particle sizes, for the purposes of scanning. The bulk samples were screened into the +26.5 mm, +19.0 mm, +13.2 mm, +9.5 mm, +6.7 mm and +4.75 mm constituents. Particles of 37.5 mm size and larger were not considered as they fall outside the COLTO (1998) specification for G1 crushed rock aggregate. Particles smaller than 4.75 mm were not scanned due to practical limitations of the scan device.

Once bulk samples were graded into its fractions, 30 particles were selected from each size constituent of each material. Table 1 summarises the number of particles used. In a number of materials the 26.5mm size fraction had some shortages, as the materials were shared with other research projects. After selection, the particles were cleaned and oven-dried (at 105°C for 24 hours) and left to cool before being scanned.

[Insert Table 1 here]

Data Capture

Scan data are captured in a spread sheet format and was entered manually into Microsoft Excel®. The final result was six spread sheets (i.e. 26.5 mm, 19.0 mm, 13.2 mm, 9.5 mm, 6.7 mm and 4.75 mm) which contained data of all material types scanned.

Parameters that were recorded in the dataset include the material type, the number of the sample particles and the three primary dimensions (i.e. length, width and depth/height). The volume (mm³) and total surface area (mm²) of each particle was recorded directly from the scan results. The scan data and/or spread sheets could not be included in this article due to its volume (i.e. 1149 data sets).

DISCUSSION

Two working concept models were developed using the scan data, one of which was ultimately abandoned as the model was heavily affected by factoring in the so-called elongation value (the ratio between the minimum and maximum dimensions of a particle). The second model proved to be superior to the first in that it discerned between differences in particle textures, as opposed to being excessively affected by the particle shape. The second model was further refined to develop a reference system for each particle size to enable comparison of particle textures.

Data Manipulation

The data arrangement described above was done specifically with the aim of calculating another parameter to be used in analysis. In spread sheet form the parameter was labelled "Elongation Value". The flakiness index has previously been assessed using this same laser scanning system by Anochie-Boateng *et al.* (2011) but the approach adopted here (i.e. using the elongation value) simplified data analysis considerably.

In order to make provision for the effects of elongated or flattened particle shapes, data were divided into subgroups to assess whether the elongation/flatness would affect the derived model(s). It was therefore necessary to derive a descriptor for its assessment. Particle data were divided into "regular" and "elongated" subsets, based on the ratio between the maximum and minimum dimensions of a particle. An "elongation value" was calculated as follows:

$$EV = \frac{Length}{Depth} \tag{1}$$

where

EV = Elongation Value

Length = the particle's maximum dimension

Depth = the particle's minimum dimension

Particles with an EV > 2 were considered to be "elongated" while particles with EV < 2 were considered to be "regular". Spread sheet data were sorted based on the calculated elongation values and subdivided accordingly. Once data sorting had been completed, limited descriptive statistics and histogram analyses were done to verify data properties and identify potential shortages in data ranges.

Data were processed for all particle sizes and subsequently two comparative models, one for elongated particles and one for regular particles, and were produced for each size fraction.

Proposed Model One

The first model considers three parameters, namely particle volume, particle surface area and the calculated elongation value. The *model value* is calculated as follows:

$$Model\ value = \frac{V}{A} \times EV \tag{2}$$

where

 $V = \text{particle volume (mm}^3)$

 $A = \text{particle surface area (mm}^2)$

EV = elongation value

Histogram analyses for the regular and elongated data sets are illustrated in Figure 1 and Figure 2, respectively. From the elongated particles histogram (Figure 2) it is clear that there are no data for the model value range between 6.0 and 7.0, which produced gaps in the output chart for the model. This underlines the effects of factoring in the elongation value and the limitations it imposes on the derived models.

[Insert Figure 1 here]

[Insert Figure 2 here]

After revision of the histogram data, model values were sorted in ascending order and model value ranges were calculated. Ten random values within the data range were selected and the closest corresponding particle was entered into a table. Table 2 shows the models selected for elongated particles, while Table 3 shows the models selected for regular particles.

[Insert Table 2 here]

[Insert Table 3 here]

The tabular results reflect the findings of the histogram analyses where gaps were found in the data range proposed for the elongated particles. The effects of factoring the elongation value into the model also introduced a bias towards elongated particles in both datasets. Subsequently the results show little refinement in data, particularly for the elongated particles. Considering that the aim of the model was to sort the particle data from coarse/angular to smooth/rounded particles, it is clear that this model was only moderately successful for regular particles and largely unsuccessful for elongated particles.

Proposed Model Two

The results from the first model led to the second model, where the elongation value is no longer factored into the model value:

$$Model\ value = \frac{V}{A} \tag{3}$$

V, and A are defined in Equation 2.

The subsequent histogram analyses show an improvement in model data presentation.

While the histogram for regular particles (Figure 3) shows little improvement over the

same dataset for model one, the histogram for elongated particles (Figure 4) no longer

has the previous gap in data for the second model.

[Insert Figure 3 here]

[Insert Figure 4 here]

The calculated model values were again sorted in ascending order. Ten equal

increments were identified between the minimum and maximum data values. The data

sets were used to identify the nearest corresponding value and the applicable particle

was again entered into a table. The results are summarised in Table 4 while Table 5

and Table 6 show the results for elongated and regular particles, respectively.

[Insert Table 4 here]

[Insert Table 5 here]

[Insert Table 6 here]

In this instance the results showed significant refinement based on the second model

data. The refinement is perhaps more clearly observed in the elongated particle results

than in the regular particle results. Model two shows good potential to arrange

particles from rough/angular to smooth/rounded, although there is still room for

improvement.

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Model Comparison

By comparing the results from the two proposed models visually, it is clear that the second model refines and arranges data more effectively from rough or angular particles to smooth or rounded particles. The limited calculated model data in model one may be due to factoring the elongation value into the calculations.

CONCLUSION

The research was undertaken to determine if a three dimensional laser-based scanning method can be applied to the modelling of shape and texture of road aggregate particles in an attempt to arrange particles based on their surface texture. The laser technique applied was relatively successful to produce comparative models that can be used to describe or classify aggregate particle shape and surface texture properties. Scan data obtained from scanning of various aggregate materials were used to develop and compare two derived models.

The results indicate that the proposed model 2 shows potential to better describe the shape and surface texture properties of the aggregates when compared to model 1 although further development is required. It was found that factoring an elongation value (*EV*) into the proposed model 1 had negative effect. The data with the elongation value were heavily biased towards the inclusion of either elongated or flat particles. Generally, none of the proposed models could effectively discern between particle roughness and angularity as the two parameters appear to be strongly codependant.

RECOMMENDATIONS

It is suggested that the approach used in model two be considered for further analyses and investigations of particle properties. Additional scan data obtained from mixed geological sources will also assist in the further refinement of the models. Further studies may have to investigate how elongation factor can be combined with surface area or volume separately as alternatives.

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 Table 1 Summary of Scanned Particles

	Quartzite	Granite	Tillite	Hornfels	Recycled	Alluvial	Dolerite	Total
					Aggregate	Gravel		
26.5mm	-	22	30	18	-	-	30	100
19.0mm	30	30	30	30	30	30	30	210
13.2mm	30	30	30	30	30	30	30	210
9.5mm	30	30	30	30	30	30	30	210
6.7mm	30	30	30	30	30	30	30	210
4.75mm	30	30	30	30	30	30	29	209
Total	150	172	180	168	150	150	179	1149

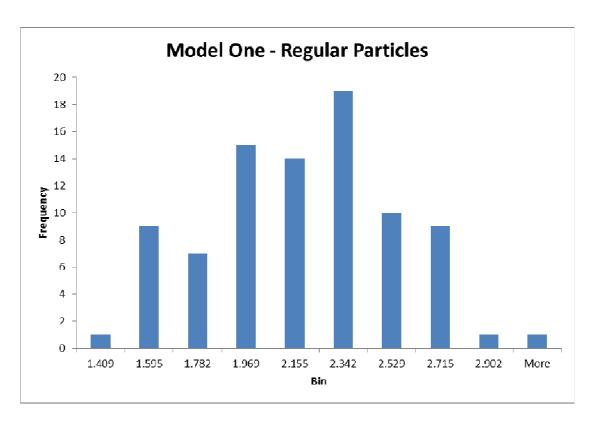


Figure 1 Histogram for model one output data - regular particles

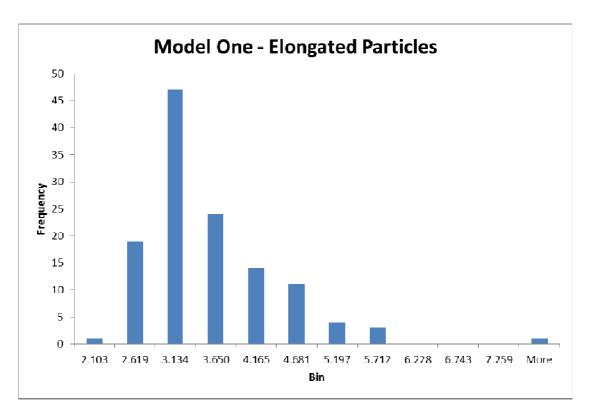
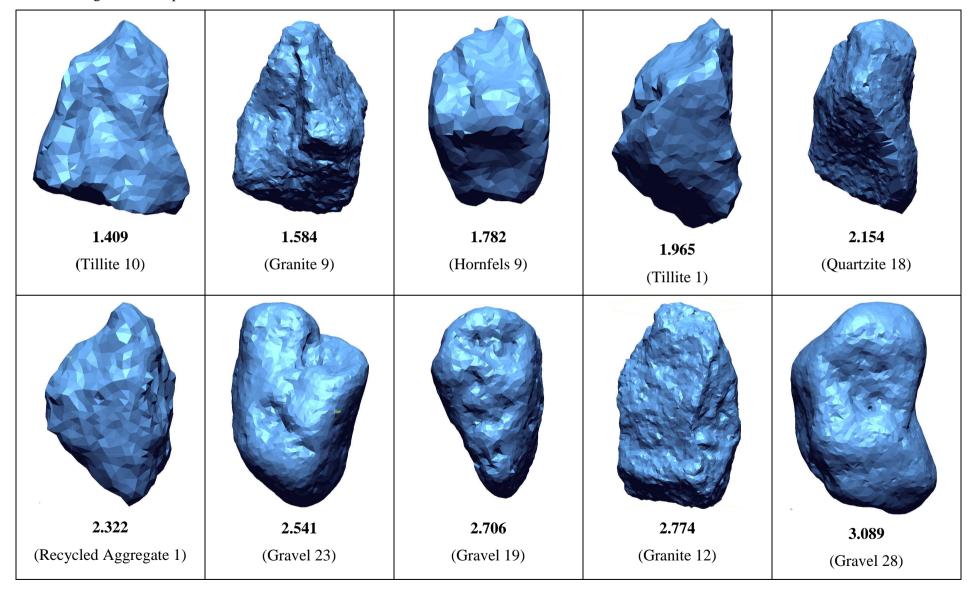


Figure 2 Histogram for model one output data - elongated particles

Table 2 Elongated 6.7mm particles – model one

2.103	2.736	3.345	4.000	4.630
(Hornfels 16)	(Dolerite 11)	(Gravel 21)	(Hornfels 24)	(Granite 28)
5.433 (Quartzite 29)	5.600 (Quartzite 28)	(Insufficient data)	(Insufficient data)	7.775 (Tillite 28)

Table 3 Regular 6.7mm particles – model one



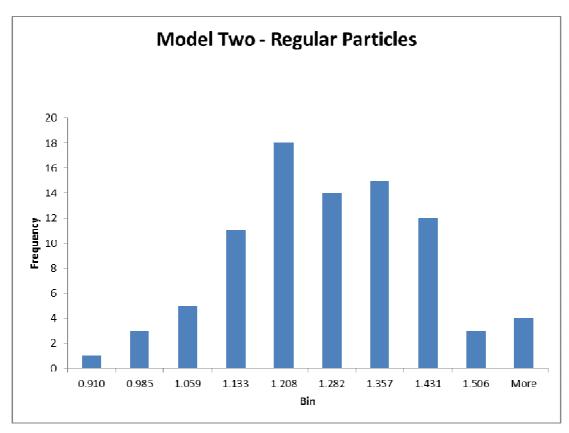


Figure 3 Histogram for model two output data - regular particles

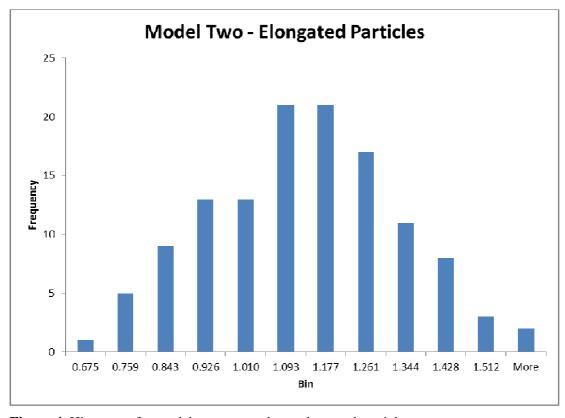
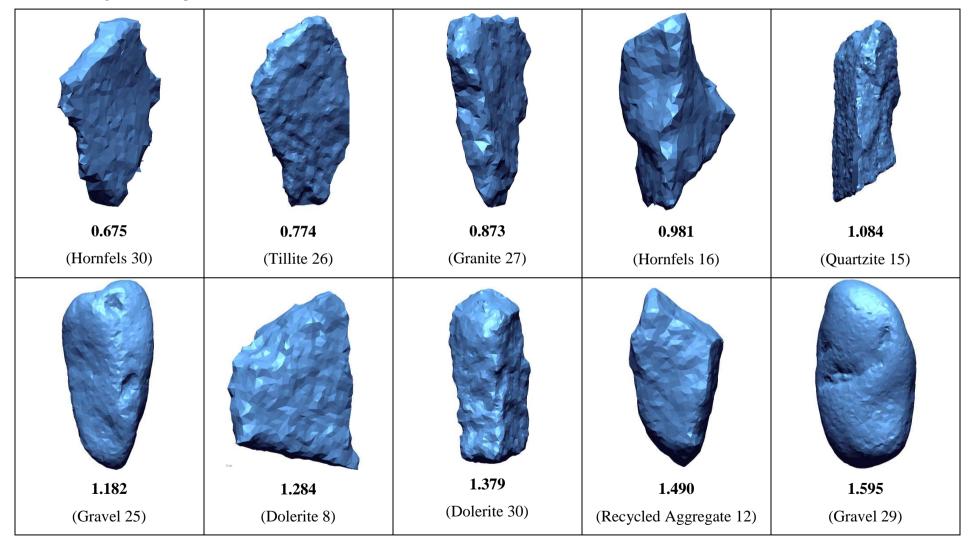


Figure 4 Histogram for model two output data - elongated particles

Table 4 Particle selection for 6.7mm – model two

	REGULAR PARTICLES	5		
Minimum V/A		0.910		
Maximum V/A	1.580			
Range	0.670			
Increments Calculated	0.074			
Increment Value	Nearest Match	Sample/Particle Number		
0.910	0.910	Hornfels 10		
0.985	0.981	Hornfels 7		
1.059	1.062	Tillite 9		
1.134	1.139	Tillite 15		
1.208	1.208	Quartzite 2		
1.282	1.285	Granite 6		
1.357	1.354	Gravel 4		
1.431	1.439	Gravel 19		
1.506	1.513	Gravel 5		
1.580	1.580	Gravel 22		
E	LONGATED PARTICLE	ES		
Minimum V/A		0.675		
Maximum V/A	1.595			
Range		0.920		
Increments Calculated		0.102		
Increment Value	Nearest Match	Sample/Particle Number		
0.675	0.675	Hornfels 30		
0.778	0.774	Tillite 26		
0.880	0.873	Granite 27		
0.982	0.981	Hornfels 16		
1.084	1.084	Quartzite 15		
1.186	1.182	Gravel 25		
1.289	1.284	Dolerite 8		
1.391	1.379	Dolerite 30		
1.493	1.490	Recycled Aggregate 12		
1.595	1.595	Gravel 29		

Table 5 Elongated 6.7mm particles – model two



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Table 6 Regular 6.7mm particles – model two

