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Rock Fragmentation Analysis using WIPFRAG

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Abstract: Size of rock fragments subsequent to blasting has direct impact on cost of transportation and processing. The cost increases with the increasing size of the fragments. This necessitates quick and accurate measurements of size distribution of fragmented rocks to decide further course of action for optimizing the cost of entire operation. There are many such measurement methods available and used by industry/researchers but most of the methods are time consuming and not precise. In such scenario WipFrag came as an automated image based granulometry system that uses digital image analysis of rock photographs and video tape images to determine grain size distributions. In this project, images of rock pile samples will be captured at different angles using camera and analysis of cumulative size distribution and optimum rock fragmentation will be carried out using WipFrag.

Keywords: Rock fragmentation, blasting, size distribution, WipFrag, image analysis.

I. INTRODUCTION

Fragmentation is a process utilized for breaking solid in situ rock masses into smaller particles. The fragmentation of rock is usually performed by blasting or drilling. Various parameters have to be considered while blasting to minimize the effects of the process. The Wipfrag software is used for measuring fragmentation. It uses the digital images of the blasted site to analyze the fragmentation. The fragmentation sizing system, also known as WipFrag, is used in mines and materials handling industries for the evaluation of the efficiency of comminution processes. The images taken by a digital camera or camcorder are transferred to the WipFrag system. WipFrag is a software that automatically captures and displays images from a muck pile. The images are then analyzed using an automatic netting system.

The detailed methodology of fragmentation analysis with a WipFrag system is discussed in further sections

II. METHODOLOGY

Qualitative visual observation is a field approach that uses a visual interpreter. It is widely used for initial observation. Due to the increasing popularity of digital image processing systems, the sizing of materials is becoming more prevalent. Numerous image processing programs are available to assist in this task.



Fig 1. Wipfrag Main Screen

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2.1. Image Processing

Image processing is used to transform the image rock fragments (Figure 3.4) into a binary image consisting of a net of block outlines

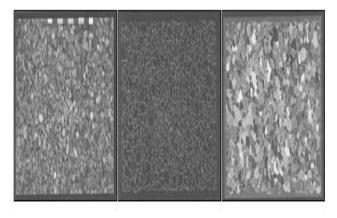


Figure 2.1. Images of pea gravel Figure 2.2. Net of rock edges Figure 2.3. Identified rock fragments

2.2. Fragment Delineation and Editing

The generation of binary images from acquired images is done by Wipfrag. It automatically identifies and quantifies the particles in the images.

Editing the images is necessary to enhance their delineation and some manual editing is also required.

2.3. Block Identification

The delineation of blocks in WipFrag involves the identification of block edges. This is done in a two-stage process.

The first stage uses various image processing techniques to image smooth and dark areas. The operators are used to detect the faint shadows between the blocks and provide clean images.

The second stage uses various reconstruction techniques to identify blocks that are only partially outlined in the first stage.

2.4. Edge Detection Variables (EDV)

Edge detection variable (EDV) are used to improve the efficiency of the image processing

stages. The user can either select one of nine preset combinations or modify the individual variables.

A representation of Wipfrag software is shown in figure

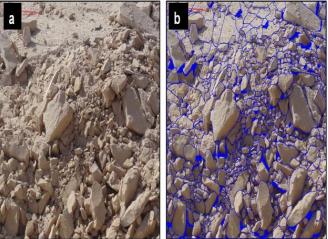


Fig 3. (a) Field Image for calculation of size distribution and (b) Delineated image.

2.5. Editing to improve the fidelity of the net

When the net is not as accurate as intended, manual editing is often necessary. This method involves removing false edges and polylines, and then drawing missing edges.

2.6. Reconstruction from 2-D to 3-D

The first step in this process is to divide the 2-D distribution into 40 bin sizes. The first step in this process is to divide the 2-D distribution into 40 bin sizes.

III. RESULTS AND ANALYSIS

The cumulative size distribution of rock piles is obtained from the multiple image analysis technique. This method is used for optimal rock fragmentation.Validation of results are represented below [Swanirman Sunirmit Publications of Research - Special Issue ICRTTEAS April 2022] [2022-23]

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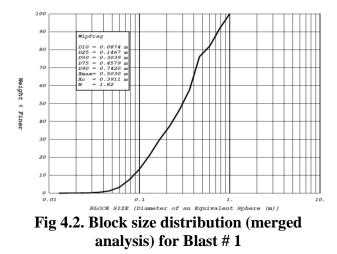
3.1. Assessment of Fragmentation for 102 mm Diameter



Frame 1Frame 2Frame 3Fig 4.1. Original gray scale images for Blast #1

Table 1. Fragment size distribution for blast #1 for different frames

Photo ID	Mean fragment size, (m)	Uniformity Index, (n)	Characteristic Size, m (Xc)	No of Blocks	Maximum Fragment Size, (m)
Frame 1	0.445	2.6	0.4162	265	1.000
Frame 2	0.266	1.82	0.2148	575	0.774
Frame 3	0.444	2.01	0.4409	120	0.774





Frame 1Frame 2Frame 3Fig 5.1. Original gray scale images for blast #2

Table 2. Fragment size distribution for blast #2 for different frames

Photo ID	Mean fragment size, (m)	Uniformity Index, (n)	Characteristic Size, m (Xc)	No of Blocks	Maximum Fragment Size, (m)
Frame 1	0.254	2.50	0.2321	423	0.599
Frame 2	0.288	2.85	0.2739	412	0.774
Frame 3	0.401	2.42	0.4271	207	0.774

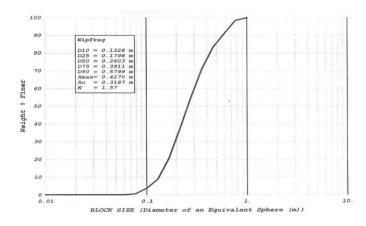


Fig 5.2. Block size distribution (merged analysis) for Blast # 2

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Table 3. Merged Analysis of Size Distributionfor Blasts 1 & 2 (102mm)

Photo ID	Mean fragment size, (m)	Uniformity Index, (n)	Characteristic Size, m (Xc)	No of Blocks	Maximum Fragment Size, m
Blast 1	0.396	1.82	0.3911	962	1.0
Blast 2	0.306	2.19	0.2865	1044	0.8

Table4. MergedAnalysisofPassingPercentage for Blasts 1 & 2(102mm)

Blast #	Sieve Size(Mm), Percentage Of Passing								
	1000	1000 800 500 300 150 125 100							
Blast 1	100.0%	96.5%	77.0%	49.50%	25.8%	19.8%	13.3%		
Blast 2	100.0%	95.2%	87.5%	65.4%	27.6%	21.8%	14.7%		

3.2. Assessment of Fragmentation for 165 mm Diameter

The size distribution of the blast is obtained by merging individual frames into a single result. For each blast, an object of known length is taken.(1m \times 1m)





Frame 1

Frame 2





Frame 3

Frame 4

Fig 6.1. Original Gray Scale Images for Blast # 3

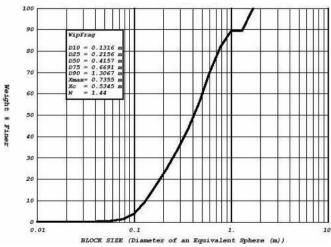


Fig 6.2. Fragment size distribution Curve forBlast #3

Table 5. Fragment size distribution for blast #3 for different frames

Photo ID	Mean fragment size, (m)	Uniformity Index, (n)	Characteristic Size, m (Xc)	No of Blocks	Maximum Fragment Size, m
Frame 1	0.406	1.48	0.4770	568	0.774
Frame 2	0.259	2.90	0.3103	335	0.464
Frame 3	0.428	1.72	0.4303	700	1.000
Frame 4	0.964	1.52	0.9323	203	1.668



Frame-1

Frame-2



Frame-3 Frame-4 Fig 7.1. Original gray scale for Blast # 4

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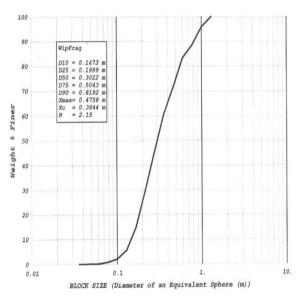


Fig 7.2. Fragment size distribution curve forBlast # 4

Table 6. Fragment size distribution for blast #4 for different frames

Photo ID	Mean fragment size, (m)	Uniformity Index, (n)	Characteristic Size, m (Xc)	No of Blocks	Maximum Fragment Size, m
Frame 1	0.570	2.04	0.5706	207	1.000
Frame 2	0.576	1.99	0.5739	417	1.292
Frame 3	0.613	2.51	0.6630	186	1.000
Frame 4	0.405	2.23	0.3985	286	1.000

Table 7. Merged Analysis Of Fragment SizeDistribution For Blasts 3 & 4 For 165mm

Blast #	Mean fragment size, (m)	Uniformity Index, (n)	Characteristic Size, m (Xc)	No of Blocks	Maximum Fragment Size, (m)
Blast 3	0.584	1.44	0.53	1767	1.668
Blast 4	0.539	1.34	0.54	1130	1.292

Table 8. Summary Of Merged Analysis OfPassing Percentage For 165mm.

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	SIEVE SIZE(mm), percentage of passing							
Blast #	1000	1000 800 500 300 150 125 100						
Blast 3	92.00%	83.20%	61.30%	33.20%	11.90%	8.90%	6.90%	
Blast 4	96.80%	96.80% 82.50% 58.40% 30.60% 12.40% 9.60% 6.10%						

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The results obtained from the individual analysis of the rock pile samples cannot be treated as perfect because the digital images used for analysis cannot reveal the conditions of fragmentation behind the muck pile surface. Hence, it becomes necessary to obtain an average result of the analysis carried out with various samples. For this purpose merging of the individual results is done. The results thus obtained would be precise enough to predict the optimum blast parameters. The results obtained from multiple image analysis are shown in Table 9 below

Table 9. Mean Passing Percentage(%) of All 4Blasts

SIEVE SIZE(mm), Passing Percentage(%)								
1000mm 800mm 500mm 300mm 150mm 125mm 100mm								
97.2%	89.35%	71.05%	44.67%	19.42%	15.02%	10.25%		

IV. CONCLUSION

The WipFrag is efficient fragmentation analysis software which uses photos to analyze the rock fragments. It is a direct method of fragmentation assessment as compared to the other methods such as the shovel loading rate method, explosive consumption in secondary blasting method or lab sampling method.

Optimum size distribution of the samples are analyzed with multiple image analysis of Wipfrag software and found the passing percentage for 1000mm- 97.2%, 800mm- 89.35%, 500mm-

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71.05%, 300mm- 44.67%, 150mm- 19.42%, 125mm-15.02% and 100mm- 10.25%.

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