THESIS FOR THE DEGREE OF DOCTOR OF PHILOSOPHY

DEM Modelling and Simulation of Cone Crushers and High Pressure Grinding Rolls

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CHALMERS UNIVERSITY OF TECHNOLOGY
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DEM Modelling and Simulation of Cone Crushers and High Pressure Grinding Rolls
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Cover:
Illustration of DEM-particles being crushed

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“Everything develops through the mechanism of iteration”
Abstract

The comminution of rock and ore materials consumes ~1.5-1.8 % of the total energy production in mining intensive countries (Tromans, 2008). Several research findings show that there are ways of utilizing compressive breakage modes that are more energy efficient compared to conventional grinding circuits based on large inefficient tumbling mills. Circuits using Cone Crushers and High Pressure Grinding Rolls (HPGR) have proven to be more energy efficient. These comminution devices have during the last two decades been implemented for hard rock materials. These machines are hence suitable subjects for further performance improvement and optimization.

In this thesis a simulation platform, based on the Discrete Element Method (DEM), for simulation of compressive breakage machines such as Cone Crushers and HPGRs is presented. The research notion is that in order to further develop compressive breakage machines and operations, fundamental understanding is needed with regard to specific details inside the machines.

The rock particles are modelled using the Bonded Particle Model (BPM) and particle shapes are based on 3D scanned rocks. The machine geometry is based on CAD modelling and 3D scanning. The interactions between rock particles and between rock particles and the machine boundaries are modelled using contact models that determine the reaction forces. A novel DEM calibration and validation framework based on design of experiments, surrogate modelling and multi-objective optimization has been developed for calibrating and validating particle flow and breakage.

Seven different simulation case studies are included in the thesis work. The simulation and modelling capabilities have successively progressed for each study conducted. The result and findings are both attributed to machine specific insights and generic modelling findings.

The work shows that the most vital machine and process performance responses, such as product size distribution, pressure distributions and power draw can be predicted using the DEM simulation platform.
Sammanfattning

Krossning och sönderdelning av berg- och malmmaterial förbrukar ~ 1.5-1.8\% av den totala energiproduktionen i gruvintensiva länder (Tromans, 2008). Ett flertal studier har visat att man genom att utnyttja kompressiv krossning kan uppnå mer energieffektiva lösningar jämfört med konventionella malprocesser baserade på stora tumlande kvarnar. Processer baserade på användning av konkrossar och HPGR maskiner (High Pressure Grinding Rolls) har visat sig vara mer energieffektiva. Dessa maskiner är därför lämpliga kandidater för ytterligare prestandaförbättringar och optimering.

I denna avhandling presenteras en simuleringsplattform, baserad på diskret elementmetod (DEM), för simulering av kompressiv krossmaskiner. Ett grundläggande antagande för forskningen är att ytterligare förståelse och kunskap angående maskinernas inre mekanismer är nödvändig för att möjliggöra fortsatt vidareutveckling samt kostnadseffektiv utveckling av nya maskiner.


Totalt sju olika simuleringsfallstudier är beskrivna i avhandlingen, där simulerings- och modelleringsfunktioner successivt har utvecklats för varje studie. De resultat och insikter som gjorts i dessa studier är både relaterade till maskinspecifika lärdomar samt ny förståelse gällande DEM simulerings och modellering.

Arbetet visar att de mest vitala processtekniska maskinparametrarna, såsom produktstorleksfördelning, tryckrespons och effektuttag kan förutsägas med hjälp av simuleringsplattformen.
Publications

The thesis contains the following papers:


Paper F: Quist, J., Evertsson, C. M., Calibration of DEM Contact Models, Submitted to Granular Matters, Feb 2017


Paper H: Johansson, M., Quist, J., Evertsson, C. M., Hulthén, E., Cone crusher performance evaluation using DEM simulations and laboratory experiments for model validation, Minerals Engineering, 2016

Paper I: Quist, J., Johansson, M., Evertsson, C. M., Calibration of DEM Bonded Particle Model Using Surrogate Based Optimization, Under review for Minerals Engineering

Paper J: Quist, J., Evertsson, C. M, Investigation of Roller Pressure and Shear Stress in the HPGR using DEM, in XXVIII International Mineral Processing Congress, 2016: Quebec, Canada

In papers A-G and J, Quist and Evertsson initiated the idea. The implementation was performed by Quist. Quist wrote the paper with Evertsson as a reviewer.

In paper H, Quist, Johansson and Evertsson initiated the idea. M. Johansson and Quist performed the DEM modelling and co-wrote the paper with Evertsson and Hulthén as reviewers.

In paper I, Quist initiated the idea. Quist and Johansson performed the DEM modelling and experiments and co-wrote the paper with Evertsson as reviewer.
Other publications


Evertsson, C.M., Quist, J., Bengtsson, M., Hulthén, Erik (2016), Monitoring and validation of life time prediction of cone crusher with respect to loading and feeding conditions. Comminution 16. 904 (1 Vol) ISBN 9781510826670
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Appended papers

Paper A:  Simulating Capacity and Breakage in Cone Crushers Using DEM
Paper B:  Application of discrete element method for simulating feeding conditions and size reduction in cone crushers
Paper C:  The effect of liner wear on gyratory crushing – a DEM case study
Paper D:  Simulating Pressure Distribution in HPGR using the Discrete Element Method
Paper E  Simulating breakage and hydro-mechanical dynamics in high pressure grinding rolls using the discrete element method
Paper F:  Calibration of DEM Contact Models
Paper G:  Cone Crusher modelling and simulation using DEM
Paper H:  Cone crusher performance evaluation using DEM simulations and laboratory experiments for model validation
Paper J:  Calibration of DEM Bonded Particle Model Using Surrogate Based Optimization
Paper I:  Investigation of Roller Pressure and Shear Stress in the HPGR using DEM
Preface

The research presented in this thesis was carried out at the research group Chalmers Rock Processing Systems at the department of Product and Production Development, Chalmers University of Technology. The research project has received support from the Sustainable Production Initiative - Production Area of Advance and from Ellen, Walter and Lennart Hesselmans foundation for scientific research. Without this support the project would not have been possible and it is gratefully acknowledged.

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Additionally, I would like to take this opportunity to express my gratitude to all past and present members of the CRPS research group and department faculty members for sharing expertise, guidance and all the valuable and senseless conversations.

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Finally I would like to thank my dad Gunno, mom Anna-Lena, my sisters Helena, Malin and Hanna for all the love, care and great support and especially my girlfriend Cecilia for always being there with all your motivation, inspiration and love.

Johannes Quist

Göteborg, April 2017
# Notation

<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>AG</td>
<td>Autogenous Grinding</td>
</tr>
<tr>
<td>ANOVA</td>
<td>Analysis of Variance</td>
</tr>
<tr>
<td>BEM</td>
<td>Boundary Element Model</td>
</tr>
<tr>
<td>BPM</td>
<td>Bonded Particle Model</td>
</tr>
<tr>
<td>CAE</td>
<td>Computer Aided Engineering</td>
</tr>
<tr>
<td>CSS</td>
<td>Close Side Setting</td>
</tr>
<tr>
<td>CCD</td>
<td>Central Composite Design</td>
</tr>
<tr>
<td>DEM</td>
<td>Discrete Element Method</td>
</tr>
<tr>
<td>DOE</td>
<td>Design of Experiments</td>
</tr>
<tr>
<td>FEM</td>
<td>Finite Element Method</td>
</tr>
<tr>
<td>HMNS</td>
<td>Hertz-Mindlin No Slip Model</td>
</tr>
<tr>
<td>HPGR</td>
<td>High Pressure Grinding Roll</td>
</tr>
<tr>
<td>IPB</td>
<td>Inter-Particle Breakage</td>
</tr>
<tr>
<td>MOO</td>
<td>Multi-Objective Optimisation</td>
</tr>
<tr>
<td>PBRM</td>
<td>Population Balance Replacement Model</td>
</tr>
<tr>
<td>SAG</td>
<td>Semi-Autogenous Grinding</td>
</tr>
<tr>
<td>SPB</td>
<td>Single Particle Breakage</td>
</tr>
<tr>
<td>PEM</td>
<td>Polyhedral Element Model</td>
</tr>
</tbody>
</table>

- $m_i$: Mass of particle $i$
- $\mathbf{f}_i$: Interaction force
- $T_i$: Interaction torque
- $\mathbf{x}_i$: Position vector, particle $i$
- $\mathbf{v}_i$: Velocity vector, particle $i$
- $\mathbf{\omega}_i$: Angular velocity vector, particle $i$
- $I_i$: Moment of inertia, particle $i$
- $g$: Gravitational field constant
- $h$: Euler time-step
- $\hat{n}$: Normal unit vector
- $\hat{t}$: Tangential unit vector
- $\mathbf{\delta}_n$: Normal interaction overlap
- $\mathbf{\delta}_t$: Tangential interaction overlap
- $R_i$: Radius of particle $i$
- $\mathbf{r}_{\text{rel}}$: Local contact position vector
- $\mathbf{\bar{v}}_{\text{rel}}$: Relative contact point velocity vector
- $\mathbf{\bar{F}}_n$: Normal contact force vector
\( \vec{F}_{d} \) Normal contact damping force vector
\( E^{*} \) Equivalent Young's modulus
\( v_{n}^{r} \) Normal relative contact velocity
\( v_{p}^{\tau} \) Tangential relative contact velocity
\( \beta \) Damping constant
\( k_{n} \) Normal contact stiffness
\( n_{i} \) Poisson's ratio
\( e \) Coefficient of restitution
\( \mu_{s} \) Coefficient of static friction
\( \mu_{r} \) Coefficient of rolling friction
\( \vec{F}_{b} \) Bond resultant force
\( \vec{F}_{n,b} \) Bond normal force
\( \vec{F}_{\tau,b} \) Bond tangential force
\( M_{n,b} \) Bond normal moment
\( M_{\tau,b} \) Bond shear moment
\( k_{n}^{\tau} \) Bond normal stiffness
\( k_{\tau}^{\tau} \) Bond shear stiffness
\( \Theta_{n} \) Normal angular overlap
\( \Theta_{\tau} \) Tangential angular overlap
\( \sigma_{c} \) Critical normal stress constraint
\( \tau_{c} \) Critical shear stress constraint
\( \hat{y} \) Prediction function
\( \Phi \) Vandermonde matrix
\( \Delta_{j} \) Error deviation constraint
\( \varepsilon \) Error estimate
\( \mathbf{x} \) Optimization design variable vector
\( \mathbf{g} \) Inequality constraints
\( \mathbf{h} \) Equality constraints
\( \mathbf{p} \) Parameters
\( \lambda^{j} \) Variable set constraint
\( w_{j} \) Objective function weight constant
\( \varphi_{j} \) Roller angular position
\( \dot{m} \) Mass flow rate
\( R \) Roller radius
\( b \) Roller width
\( v_{0} \) Particle bed initial velocity
\( T_{1,2} \) Left and right roller torque
1 Introduction

Computer simulations have been used since the late WWII years to understand complex systems that are not possible to solve using only analytical closed form expressions. The two mathematicians Jon von Neumann and Stanislaw Ulam realised during the years 1946-47 that computers could be used for implementing a statistical approach to problems related to neutron diffusion and other questions of mathematical physics (Eckhardt, 1987). This idea spawned into becoming the Monte Carlo method and the era of modelling and simulation using computers was born.

Since then, simulation methods have become increasingly important for scientists to better understand the world and for engineers in their efforts of designing machines and processes. Today many industries heavily depend on simulations as virtual environments for testing concepts and principles before creating any physical objects or prototypes. This reduces the need for physical prototypes and experiments which leads to decreased development costs and lower risks. However, the risk and cost reduction is dependent on that the simulation results conform to reality within the domain under study. Calibration and validation of models are therefore of utmost importance if decisions are to be made based on simulation outcomes.

This thesis work targets discrete element modelling and simulation (DEM) of the comminution of rock and ore materials in two different machines used in the minerals processing industry. The research project is named Optimal Crushing and Grinding and is conducted within the research group Chalmers Rock Processing Systems (CRPS) at Chalmers University of Technology, Sweden.

Comminution is the process in which ore particles are broken and reduced in size to a level where the minerals of interest can be separated, or in other words liberated from the waste gangue rock. The size reduction is done in sequential steps through a series of comminution, classification and separation stages. The choice of circuit design is governed by several factors including ore type and grade, ore deposit variability, scale of operation, company culture and available capital investment.

Schönert (1972; 1979) stated that the compression of single particles between two plates is a suitable benchmark experiment in order to evaluate the breakage efficiency in comminution. As reported in the study by Fuerstenau (2002) there are controlled laboratory particle cleavage tests that are capable of achieving breakage efficiencies above 1.0 m²/J. In comparison the Schönert compression tests showed a breakage efficiency of 0.02 m²/J and the Hukki (1944) experiments using a pendulum crusher resulted in 0.01 m²/J. However, when comparing these values with ball milling experiments reported in the Fuerstenau study, with an efficiency of ~0.0028 m²/J, it is clear that the compression breakage mode requires less energy to produce the same amount of
new surface area. The efficiency and breakage probability can be directly linked to the number of contact points per particle and hence the stress concentration level for each contact point given a certain force level applied. An illustration of different particle loading conditions can be seen in Figure 1.

Figure 1. Schematic illustration of particle breakage crack patterns for four different loading conditions a) Impact particle breakage b) Point contact loading on a particle supported by a soft compression plate c) Two-point contact loading d) Multiple-point contact loading. (modified from Schönert, 1979).

The analytical and experimental findings made by Schönert resulted in the invention and design of the high pressure grinding roll machine (HPGR) (Schönert, 1988). Compared to conventional tumbling mill based circuit designs, see Figure 2, the HPGR based circuits have the potential to operate at significantly higher energy efficiency levels (Kellerwessel, 1993; Norgate, 1994; Schönert, 1988). This has also been confirmed in work done by Morrell (2009). Two different types of HPGR circuits are shown in Figure 3.

Based on the work on cone crushers and interparticle breakage by Evertsson (2000), Lee (2012) proposed a methodology for optimization of the interparticle breakage crushing sequence. The research mainly targeted the compressive breakage sequence in a cone crusher but could be expanded to a general formulation of any compressive breakage sequence. In principle, the results indicate that each rock type has an optimal compression sequence in terms of number of compressions and compression ratios depending on the objective function.
The HPGR machine commonly receives its feed material from an upstream crushing circuit with cone crushers. This way of utilizing compressive breakage sequentially in two different machines leads to a new optimization problem of finding the best design and settings for both machines so that the total system performs at its optimum. The performance objective should be evaluated in terms of energy efficiency, throughput capacity, operational costs, liberation and operational robustness.

Figure 2. A conventional SAG mill - BALL mill - Crushing (SABC -type) comminution circuit. (modified from Wang (2015))

Figure 3. Two different HPGR comminution circuits. a) Cone Crusher – HPGR – Ball Mill circuit b) Cone Crusher – Double HPGR – IsaMill (Modified from Wang (2013))
The simulation platform is based on the discrete element method (DEM) originally proposed by Cundall (1971; 1979). Since proposed, the method has been developed into the most powerful methodology for simulating particle flow and breakage. A thorough portrayal of how DEM has been utilized within the research field of comminution has been presented by Weerasekara et al. (2013).

As a part of the development of the simulation platform the cone crusher and the HPGR have been modelled and investigated. The presented results are hence both related to new knowledge in the research field of DEM as well as in the research field of crushing and comminution.

1.1 Industrial Challenges

The mining and minerals industries are facing a series of challenges related to fluctuating market demands, business competition and environmental constraints with the continuous need for new technology and innovation as a consequence. What is regarded as a technology challenge will differ from country to country and between minerals processing operations. However, four major trends that have been identified and could be regarded as general are listed below:

- Market demand for commodities and metals increases globally
- The ore grade is generally decreasing since most easily accessible ore deposits have already been exploited
- The ore material competence is generally higher as material is mined at greater depth where weathering effects are less likely to have affected the ore over time
- Due to reduced grade, larger masses of fresh ore need to be processed in order to produce the same amount of finished product

Summarizing the four listed challenges leads to the conclusion that, if the technology is not changed, energy consumption will have to be increased in order to keep operations at the same levels. The energy consumption used by comminution processes in USA, Canada, Australia and South Africa have been investigated by Tromans (2008). The estimated energy consumption related to comminution was between ~1.5-1.8 % of the total energy usage in the mentioned countries (based on governmental statistical data from 2004-2006). This is a significant portion of the total energy usage and the logical consequence of the three above mentioned challenges is that the energy demand will increase without further development and implementation of new technology.

Minimizing the energy demand in comminution operations could be achieved by:

1. Making existing machines more effective within the processes where they currently operate.
2. Optimizing existing machine concepts with geometrical alternations, sensor technology and control.
3. Developing novel machine concepts or hybrid solutions.
4. Total transformation of comminution circuit designs i.e. technology quantum leap
All four of these activities require engineering design and innovation in order to be realised in action. It is unlikely that a major new breakthrough will be possible without new engineering methodologies, providing the capacity of achieving more accurate and less expensive predictions of not yet materialised concepts of machines and processes. Therefore, the notion of this thesis is that a virtual simulation environment is needed so that existing and novel comminution equipment, such as rock crushers, can be tested and understood on a fundamentally new level.
2 Research Approach

The objective of this work is to develop a virtual platform for evaluating the performance of compressive breakage machines. The platform can be used for gaining fundamental understanding and optimization as well as evaluation of novel comminution concepts.

In order for the simulation platform to be of any value in the research and development decision making process, the results need to be trustworthy. Hence the development of validation and calibration methods is an integral part of the work.

2.1 Research Questions

A set of research questions have been formulated with the purpose of guiding the research process. The focus of the research questions varies from modelling specific to more holistic development perspectives.

RQ.1 How should DEM be used to understand and diagnose operating equipment in comminution processes?
RQ.2 How should DEM be used as a design evaluation tool in a product development context?
RQ.3 How should rock breakage in compressive breakage machines be modelled in order to balance validity and computational economy?
RQ.4 What is a suitable method for calculating the particle size distribution of a bonded particle model?
RQ.5 How should DEM material models be calibrated with regards to particle flow?
RQ.6 How should DEM material models be calibrated with regards to particle breakage?
RQ.7 How should machine geometry be modelled in DEM?
RQ.8 How should machine dynamics be integrated with DEM?
RQ.9 How should the estimation of forces, pressure and power draw be conducted?
RQ.10 How should DEM simulations be validated?

2.2 Delimitations

To limit the scope of the research a set of delimitations have been formulated as listed below:

- Commercial and/or open source software will be used. No software will be developed, because of time constraints.
- Classification and separation machines or operations will not be simulated or studied in detail.
2.3 Scientific Approach

This work has been carried out at the Chalmers Rock Processing Systems (CRPS) research group at the department of Product and Production Development. The background of the group is within the field of machine elements and machine design. The philosophical foundation of the research is that optimization and improvement of machine design can be achieved by fundamental understanding of the components and machines studied. The research process can further on be described as problem oriented. The problem oriented research approach has been previously described by Hultén (1998) and Evertsson (2000). A schematic diagram of the research approach can be seen in Figure 4.

The structure can be applied on different activity levels in the research process and is applied in iterations. In order to match this particular research project the structure has been slightly modified to include the validation and optimization phases.

Knowledge gaps are defined based on observations either in the beginning of the project or arise as a consequence of the research activity itself. Model calibration and simulation validation methods have been developed in the project and are based on the comparison between simulations and physical experiments. Once a simulation model is calibrated and validated it can be used for optimization of specific machines, components or processes. This results in an understanding regarding performance and characteristics of the machine or operation and can be formulated as design considerations and new knowledge.
3 Background

3.1 Minerals Processing

In minerals processing the overall goal is to extract metal and mineral material from mined ore. The conventional process is to extract ore material from the ground via drilling, blasting and excavation (mining), reduce the size of particles (concentration) via crushing and grinding (comminution) until classification and separation of gangue and valuable materials can be achieved. A simplified representation of the process, excluding the excavation process, is illustrated in Figure 5. The separation technique chosen is dependent on the properties of the specific material. Release of valuable minerals from their waste gangue minerals is called liberation and the particle size at which this can be achieved is defined as the liberation size. For a more thorough description of minerals processing the reader is referred to e.g. Wills (2006).

![Figure 5. A simplified representation of a comminution (B), classification (C) and separation (S) circuit](image)

3.2 Cone Crushers

Cone crushers are the most commonly used crusher type for secondary and tertiary crushing stages in both the aggregate and the mining industry for hard competent rock types and ores. Due to the vast number of active operating crushers in the world, a very strong global common incentive should be to maximize performance and minimize operational costs related to e.g. energy and wear.

The functional principle of a cone crusher is to compress particles between two metal surfaces. Particles can be subjected to two different modes of compressive breakage; *single particle breakage* (SPB) and *inter-particle breakage* (IPB). The particle bed in the inter particle breakage mode can either be confined or unconfined (Fuerstenau, Gutsche, & Kapur, 1996; Schönert, 1996).

The compressive action is realised by inflicting a nutational motion on an inner cone (mantle) while an outer cone (concave) remains fixed. The nutational motion is actualised by an eccentric bush attached to the main shaft. There are two main types of cone crusher designs as can be seen in Figure 6; the *Hydrocone* (a) and *Symons* (b) concepts.
The concept to the left has a main shaft that is horizontally supported by a plain bearing at the top and by the radial plain bearing bushing at the bottom. The vertical force component is handled by an axial plain bearing at the end of the shaft connected to a hydraulic piston that enables vertical positioning of the shaft and mantle assembly. The vertical position of the main shaft and the mantle is used to control the close side setting (CSS), which is defined as the narrowest distance between the mantle and the concave. The eccentric throw of the nutation is determined by the eccentricity deviation of the bushing and is hence not easily adjustable during operation.

The Symons type of crusher design can be seen to the right in Figure 6. In this concept the horizontal force component is solely handled by the eccentric radial plain bearing bushing. The vertical force component is handled by a plain bearing with a spherical sector shape placed on top of the main shaft. The radius of the spherical sector defines the pivot point of the nutation. The eccentric throw is controlled by the eccentricity of the bushing. As the mantle is fixed in its vertical position the CSS has to be adjusted by changing the vertical position of the concave instead of the mantle. This limits the possibility of adjusting the CSS during operation, because of the risk for scoring in the threads.

Figure 6. Schematic illustration of the cross-section of two types of cone crushers. 
   a) Hydrocone type b) Symons type

The size and shape of the mantle and concave define the crushing chamber i.e. the space available between the crushing surfaces. The choice of crusher size and chamber design is evaluated in
regard to the requirements on throughput capacity, ore/rock strength and the feed particle size distribution. A comprehensive study of the performance of cone crushers has been presented by Evertsson (2000). Hulthén developed, based on the models developed by Evertsson, a methodology for real-time performance optimization of crushing circuits (Hulthén, 2010).

### 3.3 High Pressure Grinding Rolls

The high pressure grinding roll machine, often called HPGR, was originally proposed by Schönert (Kellerwessel, 1990; Schönert, 1988) and is an adaptation of the conventional roller mill. Many studies have been performed targeting modelling as well as circuit implementation studies, for instance (Austin, Van Orden, & Pérez, 1980; Benzer, Aydogan, & Dündar, 2011; Daniel & Morrell, 2004; Lim, Campbell, & Tondo, 1997; Schneider, Alves, & Austin, 2009; Schönert, 1988; Schönert & Sander, 2002; Torres & Casali, 2009; van der Meer & Gruendken, 2010).

In order to achieve high compressive loads in interparticle breakage mode one of the rollers is allowed to float freely and a force is applied by a hydraulic system. The hydraulic system is built up by 2-4 pistons equipped with nitrogen gas loaded accumulators. The accumulator volume and pressure dictates the force-displacement behaviour of the floating roller. Since a defined force-damper action is applied the resulting working gap between the rollers will depend on the properties of the feed material. Hence if the feed material properties change in e.g. size or strength, a new equilibrium position will be found.

The HPGR is a complex machine system due to that the operational performance and machine dynamics are dependent on the characteristics and compressed bed response of the feed material. Hence it is also problematic to achieve statistical process control since obtaining low variation in feed size distribution and ore competency is difficult in practice.

A schematic illustration of the internal mechanical layout of an HPGR can be seen in Figure 7. The nitrogen pressure in the accumulators gives a start pressure in the hydraulic cylinders. Before beginning to feed material, the floating roll is pressed against the stop gap bars. When proceeding to feed material the floating roll is pressed back until force equilibrium is reached at a specific working gap, also increasing the pressure in the system. If the characteristics of the feed material change during operation, a new force equilibrium state will be reached. Both the working gap and the operating pressure are second order variables depending on partly the same parameters. This means only one of them at a time can be kept as a control variable. The operating pressure and working gap at which force equilibrium is achieved are e.g. dependent of the roller speed i.e. at what rate uncrushed material is introduced to the compressive zone. Some HPGRs have variable speed control using frequency drives. This enables better control of the operation in order to balance throughput and breakage ratio.

Apart from the pressure build-up functionality, the gas accumulators also act as gas springs controlling the stiffness of the system and acting as shock absorbers. The nitrogen pressure level and the volume of the accumulator determines the spring properties. The system damping properties are determined by the configuration of the hydraulic system as well as friction forces.
3.4 Compressive Breakage

Consider a single particle captured between two plates (see Figure 1c) that are either parallel or at an angle limited by a slip condition. A contact positioning system can be defined as the number of contact points and their positions on either side of the particle. If a particle is resting on a surface it will commonly be in contact at three points given that it is not either balancing on two points, or has a surface with a flatness tolerance in parity with the plate surface (Söderberg, 1995). Both alternative conditions can be considered unlikely for an irregular rock particle resting on a surface. Line contact conditions are possible for cylindrical test specimens used for diametral compression tests such as the Brazilian tensile strength test (ASTM, 2008). The six contact positioning systems can be seen in Figure 8.

Figure 8. Illustration of six contact positioning systems 1:1, 1:2, 1:3, 2:2, 2:3 and 3:3.
The circles in the illustration represent arbitrarily positioned contact points. Lines are drawn between each contact point in order to visualise the relationship between the number of contact points and the resulting stress field in the particle body.

For a particle being compressed between two surfaces at an off horizontal angle such as in a cone crusher it is more likely that the particle may get nipped in a 2:1 or even a 1:1 contact position system. Other positioning systems such as 2:2, 2:3 or even 3:3 are possible given certain geometrical conditions. However for irregular particles such as rock particles those positioning systems are less probable. The special case seen in Figure 1b with a distributed load can be achieved by using a soft compression plate (Schönert, 1979). The position 2:1 system has been utilized specifically in the cone crusher liner design developed by Kawasaki. The liners have a ribbed structure on both mantle and concave in order to achieve a three point bending loading condition.

For interparticle breakage, see Figure 9, the number of contact points for each particle will vary depending on the size distribution and shape of the particle population. Large particles will be surrounded by smaller particles hence experiencing a high number of contact points. Small particles on the other hand may only have 2-5 particle contacts. This means that large particles will have a stress state closer to a hydrostatic pressure if surrounded by many particles. This effect leads to the perhaps unintuitive result that large particles usually have lower breakage probability than small particles if looking at sequence of breakage i.e. the small particles break first since they experience higher stress concentrations due to a lower number of contact points (Schönert, 1979).

![Figure 9. Illustration on interparticle bed breakage. To the left the compression stroke (s) and the bed height (b) are visualised. In the right image the blue colour illustrates the void volume in the packing matrix.](image-url)
When studying single particle breakage under slow compression, there is a strong size-energy relationship where more energy is needed to fracture small particles (Yashima, 1987). For impact breakage the same relationship holds as presented by for instance, Tavares (2007). The increase in strength with decrease in size is due to that flaws, pores and grain boundaries are distributed in the rock body. These inhomogeneities cause stress concentrations leading to inelastic deformations and cracks. Since the probability of finding a flaw is reduced when the particle size decrease, the stress has to increase in order to break the particle.

Further on this leads to the conclusion that for interparticle breakage the outcome is highly dependent on the feed size distribution and the resulting packing behaviour. For a mono sized distribution the coordination number distribution variance will be low compared to e.g. a bi-modal distribution where the coordination number will have a large spread.

The number of contact points in a particle packing will increase with a power law if increasing the proportion of fine material. At each contact point there will be a stick-slip friction process triggered by the compression. If the contact point density is high the congregated potential for particle rearrangement also increases. Such particle rearrangement leads to heat losses due to friction and possibly also particle surface breakage.
4 Discrete Element Method

The discrete element method (DEM) is a numerical method for simulating discrete matter in a series of events called time-steps. By generating particles and controlling the interaction between them using contact models, the forces acting on all particles in the system can be calculated. Newton’s second law of motion is then applied and the new positions of all particles are calculated and updated for following time-step. As this process is repeated, it gives the capability of simulating how particles move and interact in particle-boundary systems.

The method was first proposed by Cundall (1971) for the purpose of modelling blocks of rock and later on generalized for granular materials by Cundall and Strack (1979). About 10 years later Hart and Cundall extended the method to three dimensions (P. A. Cundall, 1988; Hart, Cundall, & Lemos, 1988). DEM has evolved into a conventionally used 3D simulation computer aided engineering (CAE) tool and is used by engineers and scientists in a wide range of fields. In the pharmaceutical industry the method is e.g. used for simulating coating processes and the mixing of powders. In the agricultural industry, DEM is used for predicting the performance of complex sowing mechanisms. Most relevant to this work, DEM has become one of the most important tools for simulating machines and processes within the field of minerals processing and comminution (Weerasekara, 2013).

The calculation cycle of a DEM iteration can be seen in Figure 10. After initiating the simulation, or after a previous time increment, the choice is made whether or not to generate new particles. If particles are to be added, the type, size, orientation and position will be defined by a particle generator function. When new particles have been created, a contact detection algorithm finds all interaction contacts between objects in the simulation domain. The contacts may be categorised as contacts between particles or contacts between particles and boundary objects. For all contacts identified, the corresponding contact forces are calculated. There are several different contact force models available, suitable for various kinds of applications, materials and conditions. For dry granular flow the linear spring or the hertz mindlin (no slip) (HMNS) models are commonly used. If the material has other characteristics such as moisture leading to a sticky behaviour, a cohesion model can be used in combination with for instance HMNS. Once the forces given by contact interactions have been calculated, any other external body forces may be added. External forces are commonly added due to the existence of one or multiple physical fields such as gravitational, electromagnetic or hydrodynamic fields.

In the following sections a more detailed overview of the calculation steps is presented. However, it should be noted that the commercial software EDEM® has been used for all simulations. Hence the descriptions on for instance equations of motion and numerical integration given below are meant to provide coherence rather than as vital aspects of the research outcome.
4.1 Equations of Motion

If the forces $f_{i,n}$ acting on a particle due to interaction with other bodies or external fields are identified, the movement of the particle is may be defined by integration of Newton’s equations of motion for the rotational and translational degrees of freedom:

$$m_i \frac{d^2 \vec{x}_i}{dt^2} = \sum_{i=1}^{\infty} f_i + m_i g$$

and

$$I_i \frac{d \omega_i}{dt} = \sum_{i=1}^{\infty} T_i$$

with the mass $m_i$ of particle $i$, the position $\vec{x}_i$, the total sum of forces $f_i$ due to interactions with other particles or bounding geometries, the gravitational acceleration $g$, the particles moment of inertia $I_i$, the angular velocity $\omega_i$ and the total sum of torques $T_i$ from tangential contact forces. These equations of motion for a system of particles are a system of coupled ordinary differential equations (ODE) to be solved in $D = 2,3$ dimensions.

4.2 Numerical Integration

The topic of numerical integration and different ODE solvers is a wide area of research in computational science. Since the use and implementation of different solvers have not been part of this thesis, only the basics of numerical integration will be mentioned here. Fundamentally, numerical integration of ODEs is performed using finite difference methods in order to provide approximations of the derivative of a function so that a new point further ahead can be estimated. The Euler method, discussed by Leonhard Euler in his book *Institutionum Calculi Integralis* (Euler, 1768), is the most basic and simple method for solving initial value problems (IVP). An IVP is stated on the form,
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\[ y'(t) = \frac{dy}{dt} = f(y, t), \quad t \in [a, b], \quad y(a) = \alpha \]  

(2)

Using the notation by Mattutis (2014), replacing the differential increments \( dy \) and \( dx \) by \( \Delta y = y(t_{n+1}) - y(t_n) \) and \( \Delta t = t_{n+1} - t_n = h \) over the time-steps \( t_n \) and \( t_{n+1} \), Equation (2) can be rewritten to obtain the Euler method,

\[ \frac{\Delta y}{\Delta t} = \frac{y(t_{n+1}) - y(t_n)}{h} = f(y(t_n), t_n) \]  

(3)

This can be rearranged to obtain the new point of interest a step forward in time \( y(t_{n+1}) \),

\[ y(t_{n+1}) = y(t_n) + hf(y(t_n), t_n) \]  

(4)

The expression in Equation (4) is termed the explicit Euler step. The slope or derivative could also be evaluated at the next time-step \( t_{n+1} = t_n + h \) which leads to the implicit Euler method,

\[ y(t_{n+1}) = y(t_n) + hf(y(t_{n+1}), t_{n+1}) \]  

(5)

In essence the Euler time-stepping method is hence based on the notion that \( f(t, y) \) will provide a slope of the line tangent to \( y \) at the point \( (t, y) \). If moving along the tangent line a small step \( h = \Delta t \) (in time), a new point \( y(t_{n+1}) \) will be reached. This new point will then hopefully be close to the actual point of interest. When this is expressed using the Taylor series theorem the local error truncation error \( O(h^2) \) is added and it will accumulate as \( t \) goes from \( a \) to \( b \). In addition to the truncation error there will also be a rounding error when implementing the algorithm using floating point numbers. The error will be reduced when smaller values for the step size \( h \) is applied. Based on the principle of finite differences, several different integration methods have been developed apart from the first order accurate explicit (forward) and implicit (backward) Euler methods, such as for instance the Runge-Kutta methods, \( n \)th order Taylor expansions and so called multi-step methods such as e.g. the Verlet scheme methods. The applicability of different numerical integration schemes for DEM simulations have been investigated by for instance Kruggel-Emden et al. (2008). Kruggel-Emden compared different integration schemes for a particle-wall contact case and concluded that significant differences in accuracy as well as computational efficiency could be found. In EDEM, the Symplectic Euler Method (Griffiths, 2010), also called the semi-implicit Euler method, is the implemented approach for numerical integration. It should however be noted that the option of choosing different integration methods have been included in the latest EDEM 2017 version.

In particle simulations the truncation and rounding errors related to the configuration and choice of integration method are not the only source of errors. A common occurrence in DEM is that when a too long time-step is chosen, objects with high velocities will travel a too long distance, so that when a contact interaction has been recognised the objects are intersecting in an unnatural way. When such high overlaps are recorded it leads to high contact forces, effectively leading to a chain reaction explosion behaviour. This is especially critical in simulation applications with fast moving geometry, large particle size ratios and for stiff particle systems.
4.3 Rotation of Particles

The dynamics of the angular degrees of freedom, i.e. the rotation of bodies, can be achieved by using the concept of Euler angles. However, the Euler angle formulation has some inherent disadvantages leading to stability concerns. Matuttis and Chen (2014) illustrate in their DEM textbook that for positions close to $\theta = \pi/2$ and $\theta = 0$ the equations of motion for the Euler angles approaches singularity conditions. In many modelling situations where Euler angles are used this problem is avoided by appropriate selection of variables. In DEM simulations, the number of objects are normally far higher in contrast to for instance rigid body simulations. A consequence is that the probability of divergence of the equation of motion for some particle increases when the number of particles increase in the simulation. A remedy to the singularity problem of Euler angles is to use the method called quaternions. The quaternion formulation is based on using complex number theory and provides a very stable representation of the equations of motion for the angular degrees of freedom (Matuttis, 2014). For the conventional full mathematical formulation of the Euler angle and quaternion concepts, see for instance Matuttis (2014) or Wittenburg (2008).

4.4 Contact Detection

Before being able to calculate the resulting force from a particle-particle or particle-wall interaction, the existence of all contacts needs to be identified. The algorithms for finding contacts are called broad phase collision detection. The most simple contact detection algorithm is sometimes called the brute force method. If assuming a system of $N$ particles, then the brute force method constitutes checking each particle against all other particles leading to order $O(N^2)$ operations (Weller, 2013). The method is computationally inefficient hence other more sophisticated approaches are normally applied in simulations where particles and other objects interact.

The neighbouring cell method, which is used in EDEM, divides the simulation domain into a grid of cells with a size usually specified by a ratio to the particle size. A list is kept of particles contained in each cell and for each particle, only the particles in the same and neighbouring cells needs to be checked for contact. A consequence of this strategy is the trade-off between having a large number of cells (small cell size) versus having too many particles to check in each operation (large cell size).

Other examples of contact detection algorithms used in DEM are the nearest neighbour method (Vu-Quoc, Zhang, & Walton, 2000; Zhao, Nezami, Hashash, & Ghaboussi, 2006) where a bounding volume is used to create a list of neighbouring particles, and the sweep and prune method (Perkins & Williams, 2001) where bounding volumes and edge projections to each axis direction are used to create a list that can be sorted in order to identify which bounding volumes that are candidates for overlap. These different methods have advantages and disadvantages depending on the simulation application. The nearest neighbour method is for instance suitable for quasi-static simulations where particles do not re-arrange or move large distances. In cases
with large particle size differences, sweep and prune as well as other hierarchy methods are suitable.

4.5 Contact Kinematics

When a contact interaction has been identified by the contact detection algorithm the characteristics of the contact may be used to calculate the corresponding reaction force using a contact force model of some type. In Figure 11 two particles $i$ and $j$ are represented schematically including notations for the vectors and scalars defining the particle properties, positions, velocities and overlaps.

![Diagram showing two overlapping particles](image)

Figure 11. Illustration of two particles overlapping during contact interaction. (modified from (Kruggel-Emden, Simsek, Rickelt, Wirtz, & Scherer, 2007)).

If particle $i$’s position is described by the position vector $\bar{x}_i$ and vector $\bar{v}_i$, respectively, then the position of $j$ relative to $i$ can be defined as $\bar{x}_j = \bar{x}_i - \bar{x}_j$. The unit vector aligned with the direction of the line through particle $i$ and $j$, pointing towards $j$, can be defined if dividing by the norm magnitude of the same vector:

$$\bar{n} = \frac{\bar{x}_j - \bar{x}_i}{\|\bar{x}_j - \bar{x}_i\|}$$  \hspace{1cm} (6)

The normal overlap distance between two spherical particles is given by subtracting the distance between the centre of mass positions with the sum of the radii of particle $i$ and $j$:

$$\delta_n = (R_i + R_j) - \|\bar{x}_j - \bar{x}_i\| > 0$$  \hspace{1cm} (7)

The location of the interaction point can now be calculated assuming that the interaction position takes place at half the displacement overlap by $\bar{x}_i = \bar{x}_i + (R_i - \frac{1}{2}\delta_n)\bar{n}$. The position vector from the centre of mass to the contact point for particle $i$ and $j$ respectively is given by,
\[
\begin{align*}
\tilde{r}_{i,e} &= (R_e - \frac{1}{2} \tilde{\alpha}_e) \tilde{n} \\
\bar{r}_{j,e} &= -(R_e - \frac{1}{2} \tilde{\alpha}_n) \tilde{n}
\end{align*}
\]

(8)

The velocity vector at the contact point can now be written as \(\bar{v}_{i,e} = \tilde{v}_i + \tilde{\alpha}_i \times \tilde{\bar{r}}_{i,e}\) for particle \(i\) and as \(\bar{v}_{j,e} = \tilde{v}_j + \tilde{\alpha}_j \times \tilde{\bar{r}}_{j,e}\) for particle \(j\). The relative velocity vector of the interaction contact point is given by \(\bar{v}_{rel,e} = \tilde{v}_{j,e} - \tilde{v}_{i,e}\). The normal and tangential unit vectors can further on be used to achieve a separation of the normal and tangential component so that \(\bar{v}_{rel,e} = (\bar{v}_{rel,e} \cdot \tilde{n}) \tilde{n} + (\bar{v}_{rel,e} \cdot \tilde{t}) \tilde{t}\). We can now write the tangential unit vector as,

\[
\tilde{t} = \frac{\bar{v}_{rel,e} - (\bar{v}_{rel,e} \cdot \tilde{n}) \tilde{n}}{\sqrt{\bar{v}_{rel,e} \cdot (\bar{v}_{rel,e} \cdot n) n}}
\]

(9)

The tangential overlap distance is then given by,

\[
\delta_t = \bar{v}_{rel,e} dt
\]

(10)

where \(dt\) is the time-step.

4.6 Hertz-Mindlin Contact Model

The particle to particle interaction can be described using different types of force contact models. The most commonly used models for modelling dry rock material are linear spring models and the Hertz-Mindlin No Slip model (HMNS). A schematic illustration of the Hertz-Mindlin model can be seen in Figure 12.

The HMNS model as implemented by DEM Solutions in EDEM™ is based on spring-dashpot configurations in the normal and tangential directions with a frictional element in between in the normal direction. The normal force component is based on Hertzian contact theory (Hertz, 1882). The tangential force component is based on work by Mindlin-Deresiewitz (R. D. Mindlin, 1949; R. D. Mindlin, Deresiewicz, H., 1953). The damping components in both normal and tangential direction include the coefficient of restitution as described by Tsuji (1992). The rolling friction is a contact independent directional constant torque model (Sakaguchi, 1993).
Damping components are added to normal and tangential force components where damping coefficients are linked to the coefficient of restitution. The normal force is given by considering the normal overlap $\delta_n$ according to,

$$F_n = \frac{4}{3}E^*\sqrt{R^*}\delta_n^{3/2} \quad (11)$$

The damping force is given by relative normal velocity and the stiffness characteristics according to,

$$F_n^d = -2\dot{\delta}_n\sqrt{5/6}\beta\sqrt{k^*_n}m^* \quad (12)$$

Where the equivalent Young’s modulus $E^*$, equivalent radius $R^*$, equivalent mass $m^*$, damping coefficient $\beta$ and stiffness $k_n$ are given by

$$\frac{1}{E} = \frac{1}{E_i} + \frac{1}{E_j} \quad (13)$$

$$\frac{1}{R^*} = \frac{1}{R_i} + \frac{1}{R_j} \quad (14)$$

$$m^* = \left(\frac{1}{m_i} + \frac{1}{m_j}\right)^{-1} \quad (15)$$

$$\beta = \frac{\ln e}{\sqrt{\ln^2 e + \pi^2}} \quad (16)$$

$$k_n = 2E^*\sqrt{R^*}\delta_n \quad (17)$$

Where $E_i$, $E_j$ is the Young’s modulus, $v_i$, $v_j$ is the Poisson ratio, $R_i$, $R_j$ the radius for spheres in contact and $e$ the coefficient of restitution.

The tangential force component is defined as the tangential stiffness times the tangential overlap. In addition the tangential damping force and tangential stiffness is given by

$$F_t = -k_t\delta_t \quad (18)$$

$$F_t^d = -2\dot{\delta}_t\sqrt{5/6}\beta\sqrt{k^*_t}m^* \quad (19)$$

$$k_t = 8G^*\sqrt{R^*}\delta_n \quad (20)$$

The tangential force component is limited by the Coulomb friction $\mu_s\dot{\delta}_t$ where $\mu_s$ is the coefficient of static friction. A rolling friction can be applied by assigning a torque to the contacting surfaces.

$$\tau = -\mu_s\dot{\delta}_tR_i\omega_i \quad (21)$$
With $\omega_i$ being the unit angular velocity vector of the object at the contact point, $\mu$, the coefficient of rolling friction and $R$, the distance between the centre of mass and the contact point.

4.7 Shape Representation

A particle can be mathematically described in many different ways. The most common approach in DEM codes is to describe particles as spheres. This is reasonable for some applications where e.g. the particle shape is close to spherical and if the particles are small compared to the surrounding environment. However for most applications the particle shape and coupled associated behaviour (i.e. spin) will have a very significant effect on the simulation accurateness. Therefore the capability of modelling irregularly shaped particles is important. In this work the particles are built up by a set of spheres, called a Multisphere, which mimics the shape of 3D scanned rock particles, see Figure 13. An optimization routine for the problem of best representing a 3D model of a rock with a specific number of spheres has been developed by Price et al. (2007).

With this method the shape representation depends on the number of spheres that are included in each particle. More spheres will give increased shape quality and correspondence to the real irregular shape of the 3D scanned rock. It is important to note that the relation between the number of spheres used and the flow behaviour quality is not linear. The use of two spheres limits two rotational degrees of freedom (DoF) and a third sphere limits the third rotational DoF, preventing spinning. The consequence is that the flow behaviour when simulating with a single sphere particle and a three-sphere particle is very different. This finding is also supported by Price et al. (2007).

Particles in DEM models may be defined using different kinds of shape representation. The first DEM codes used circular discs in 2D and for a long time single-sphere representations were the most commonly used for 3D simulations. However, early on in the DEM development, polygons and tetrahedral representations were used as they provided irregular and angular shaped particle shapes better resembling rocks. While the shape representation is superior, the polyhedral approach requires more complex contact detection and contact overlap algorithms. The spherical representation has the advantage that if the radiuses and positions of a spherical particle pair are known, it is easy to identify if they are in contact, independent of the orientation.

Figure 13. Left image shows a Multisphere representation of a rock particle using seven spheres. The middle image displays the 3D shape as an overlay on the spheres. The right image shows the stl-mesh generated by 3D scanning a rock particle.
4.8 Breakage Models

In systems where the loading condition does not significantly influence the particle integrity, no specific breakage model is needed. In these cases Hertzian or linear spring contact models are commonly used. However, when using DEM to understand the mechanisms in comminution devices, where the destruction of solids is the objective, a breakage modelling approach is needed. There are three main approaches for discontinuous modelling of rock breakage:

- Bonded Particle Model (BPM) - Spheres are arranged in a cluster and bonded together in each contact point using bonding beams (Potyondy & Cundall, 2004; Thornton, Yin, & Adams, 1996).
- Polyhedral Element Model (PEM) - Particles are modelled using a tessellated mesh structure using voronoi grains, trigons or tetrahedral elements (P. A. Cundall, 1988; Potapov & Campbell, 1996).
- Population Balance Replacement Model (PBRM) - Particles are replaced by a set of progeny fragments at the breakage event (P. W. Cleary, 2001).

All three of these approaches, illustrated in Figure 14, have been used for modelling rock breakage in comminution machines. The BPM has been applied to model breakage of rock, granules and concrete materials utilizing impact breakage modes (Antonyuk, 2006; Djordjevic, Shi, & Morrison, 2003; M. Khanal, Raghuramakrishnan, R., Thomas, J., 2008; H. Lee, Kwon, Kim, & Cho, 2008; Schubert & Jeschke, 2005) and compressive breakage modes (Brown, Chen, & Ooi, 2014; Cho, Martin, & Sego, 2007; C. Ergenzinger, Seifried, & Eberhard, 2012; C. Ergenzinger, Seifried, R., Eberhad P., 2011; Herbst & Potapov, 2004; M. Khanal, Schubert, & Tomas, 2007; Johannes Quist & Evertsson, 2016).

Several authors have also modelled comminution machines such as for instance the cone crusher using the PBRM approach (Paul W. Cleary & Sinnott, 2015; P. W. Cleary, Sinnott, Morrison, Cummins, & Delaney, 2017; G.W. Delaney, Cleary, Sinnott, & Morrison, 2010; G. W. Delaney, Morrison, Sinnott, Cummins, & Cleary, 2015; Li, McDowell, & Lowndes, 2014). An alternative approach to the spherical or polyhedral shape representation is used in the mentioned papers. A method is applied where a set of smooth corner cuboids with a pre-determined size distribution are packed inside a mother cuboid particle.
At the breakage event the daughter particles are released. This method does not demand an overlap packing approach as in the case for the spherical PRBM. Due to the cuboid shape the daughter particles can be packed to a high bulk density level. This also mitigates the issue of mass conservation. The spherical PBRM has also been used to model the HPG (G. K. P. Barrios & Tavares, 2016). The PEM approach has successfully been applied by Lichter (2009) for modelling cone crushers and by Refahi (2010) for modelling a jaw crusher.

In the previously mentioned cases the comminution machines have been the main objective of the study. In the rock mechanics and geomechanics research community the interest is usually focused directly on a rock block model or a rock specimen subjected to loads in a specific breakage experiment. This allows for 2D simulations to be of value in contrast to the machine simulation case where 2D simulations rarely provide any useful information regarding the internal particle flow or breakage phenomenon. A major difference is also that one single specimen is modelled in contrast to the machine modelling case where hundreds or preferably thousands of breakable particles need to be introduced. This is illustrated in Figure 42 where the rock breakage is modelled in an industrial scale Sandvik H6000 cone crusher. The rock particles, denoted as meta-particles are built up by a cluster of fraction particles bonded to all neighbours in contact with bonding beams according to the BPM approach. In this work the major difficulty was to facilitate enough fraction particles in each meta-particle in order to achieve reasonable breakage characteristics for single particles, while simultaneously maximizing the number of meta-particles in the population in order to capture the process behaviour.

The majority of studies found in literature do not explicitly handle this trade-off problem but focus on the rock breakage characteristics only. A useful review on the topic of discrete modelling methods for fracturing processes have been presented by Lisjak (2014). A number of researchers have applied the BPM approach for modelling the uniaxial compressive strength test and Brazilian tensile strength test, see for instance (C. Ergenzinger, Seifried, R., Eberhard P., 2011; M. Khanal, Schubert, & Tomas, 2005; Potyondy & Cundall, 2004; Yoon, 2007). Mahabadi et al (2010)
modelled the diametrical compression of rock discs in the split Hopkinson bar by applying a hybrid finite-discrete element method in 2D formulated by Munjiza (2004). In the FEM/DEM method each discrete element is discretised into a mesh of finite elements. This means that the discontinuum behaviour is modelled by discrete elements and the continuum behaviour through finite elements. The transition between the continuum and discontinuum is handled through defining a fragmentation process, leading to the ability of modelling breakage.

Kazerani (2013) used a discontinuum based model implemented in the UDEC (Itasca Consulting Group, 2008) software package for modelling both compressive and tensile failure of sedimentary rock. The 2D model is based on a finite difference discrete element approach where the microstructure is modelled as an assembly of distinct mesh elements whose boundary behaviour is controlled by contact models. The authors applied a central composite design experimental plan to calibrate and analyse the statistical influence of a set of micro parameters controlling the breakage behaviour. The resulting fracture patterns resemble the experimental results in both the uniaxial strength test as well as the Brazilian tensile strength test. Similar 2D studies have also been conducted by Tatone et al (2015) and Ghazvinian et al (2014) using a Delaunay algorithm to generate a Voronoi mesh which incorporates the possibility of modelling complex crystalline grain structures.

Recently, Behraftar et al (2017) presented a methodology for the evaluation of micro parameters of a breakage model based on a method named discrete spherico-polyhedral element method (DSEM). In three dimensions each element is modelled as a polyhedral where the corners have been rounded. Due to the smoothness of all geometrical features the handling of contacts between elements becomes simple and efficient. The simulation results showed a good agreement to the Cracked Chevron Notched Brazilian Discs (CCNBD) breakage experiment.

The smooth particle hydrodynamics (SPH) method has also been applied to fracture modelling by for instance Das and Cleary (2010) and Ma (2011). Since SPH is a mesh-free method it has an advantage over mesh-based models when it comes to for instance, high deformations and self-collision. The results presented by Das (2010) using the 2D SPH fracture model for different geometrical shapes showed a good qualitative behaviour when comparing fracture patterns. Nishioka (1997) states in his review article that for continuum modelling approaches for fracture mechanics, there are three primary methods; the finite difference method (FDM), the finite-element method (FEM), and the boundary-element method (BEM). These continuum based models will not be further elaborated on or discussed in the paper specifically. However, the general calibration approach presented below would principally be applicable also in these cases.

### 4.9 Bonded Particle Model

The bonded particle model (BPM) was formulated by Potyondy and Cundall (2004) for the purpose of simulating rock breakage. The approach has been applied and further developed by Cho (2007). The concept is based on bonding or gluing a packed distribution of spheres together forming a breakable body. The particles bonded together will hereafter be called fraction particles and the cluster created is defined as a meta-particle. The fraction particles can either be of mono
size or have a size distribution. By using a relatively wide size-distribution and preferentially a bi-modal distribution the packing density within the meta-particle increases. It is important to achieve as high packing density as possible due to the potentially problematic issue with mass conservation as the clustered rock body will not be able to achieve full solid density.

The forces and torques acting on the theoretical beam can be seen in Figure 15. The schematic graph in Figure 16 illustrates the relationship between different loading modes (tension, shear, and compression), bond stiffness and strength criteria.

Figure 15. Schematic representation of two particles bonded together with a virtual beam with radius $R_b$ and length $L_b$ subjected to a resultant force $F^e_b$ as well as a normal $M^s_b$ and shear $M^c_b$ torque (modified from (P. A. Cundall, Strack, O.D., 1979; Potyondy & Cundall, 2004)).

Figure 16. Schematic force-displacement plot of the different modes of loading on a bond beam. (Modified from (Cho et al., 2007))

Before bond-formation and after bond breakage the particles interact according to the HMNS contact model. The bonds are formed between particles in contact at a pre-defined time-step. When the particles are bonded the forces and torques are adjusted incrementally according to Equation (22) (Potyondy & Cundall, 2004).
\[
\begin{align*}
    dF_n^b &= k_n^b A_\delta^b_n \\
    dF_t^b &= -k_t^b A_\delta^t_i \\
    dM_n^b &= -k_n^b J_\Theta_n \\
    dM_t^b &= -k_t^b J_\Theta_i
\end{align*}
\]

(22)

Where the normal and tangential displacements are given by \( \delta^b_n = \nu_n \Delta t \) and \( \delta^t_i = \nu_t \Delta t \) respectively. The angular displacements are similarly expressed as \( \Theta_n = \omega_n \Delta t \) and \( \Theta_i = \omega_i \Delta t \) where \( \omega_n, \omega_t \) are the normal and tangential angular velocities. The beam dimensions are given by the bonding beam radius \( R_b \) defining the cross-sectional area \( A \) and the moment of inertia \( I = \frac{1}{4} \pi R_b^4 \) and polar moment of inertia \( J = \frac{1}{2} \pi R_b^4 \). The normal and tangential stiffness parameters \( k_n^b \) and \( k_t^b \) control the stiffness characteristics of the beam.

The normal and shear stresses are computed and evaluated if exceeding the critical stress constraint values according to Eq. (23).

\[
\begin{align*}
    \overline{\sigma}_{\text{max}} &= \frac{\tilde{F}_{n,\text{total}}}{A} + \frac{2M_n^b}{J} R_b < \sigma_c \\
    \overline{\tau}_{\text{max}} &= \frac{\tilde{F}_{t,\text{total}}}{A} + \frac{2M_t^b}{J} R_b < \tau_c
\end{align*}
\]

(23)

The critical strength levels \([\sigma_c, \tau_c]\) are set to scalar values defining the rock strength. While it has not been done in this work, it is possible in principle to define the critical bonding strength using a probability density function.

An improved bonded particle model called the Timoshenko beam model has been developed by Nicholas Brown (Brown et al., 2014). This model has the inherent ability to resist tensile force, bending and twisting moment in all directions. The increased functionality in constraining the degrees of freedom for each beam improves the capability of reproducing brittle fracture. In order to increase the rock fracture accuracy this model would be of interest to implement using the meta-particle method, even though it is more computationally expensive due to the increase in numerical operations.

### 4.10 Rock-Shaped Meta-Particles

The size distribution of the fraction particles building up the meta-particles needs to be chosen so that a sufficient quality of the breakage is achieved while keeping the computational load within the constraints. A schematic illustration of three different packing structures can be seen in Figure 17. When experimenting with different packing structures for the meta-particles, the findings suggested than a bimodal distribution resulted in a good balance between achieving high packing density and acceptable breakage characteristics, while using the least number of spheres. This finding is also supported in terms of optimal packing of particles (Brouwers, 2006; Groot, 2011).
When a satisfactory packing structure has been created it can be used to form meta-particles. A 3D model of a rock is used as a selection space as seen in Figure 18. The position and radius of each fraction particle within the selection space is exported to form a meta-particle cluster model that can be used in the crusher simulation. Several meta-particles can be configured to build up a sophisticated feed material population in terms of both size and shape distribution, see Figure 19.

In simulations where crushing machines are studied, the meta-particles needs to be introduced to the system in a practical manner. In this work the introduction of the clusters are realised by first generating a set of host-particles. The host-particles are practical since they may be allowed to settle and form a bed of rocks in the feeding hopper. At a specified time the host-particles are removed and replaced by the meta-particle cluster.
Figure 19. The bonded particle model implemented for simulation of a cone crusher using a population of bonded meta-particle in EDEM 2.5. Particles represented as the bonding beams.
5 Simulation Based Design

A methodology and work flow process for using DEM in problem solving and concept development has been developed in Paper B. A process diagram of the different steps can be seen in Figure 20. The process consists of three main steps each including a number of sub-processes. The purpose of the first step is to gain fundamental understanding of the problem and system that is the target for the analysis. It is important to have a good understanding of the particle flow behaviour and, before trying to solve it, identify what is causing the problem. Therefore it is also suggested that simulation and modelling efforts are done at the highest possible fidelity level in terms of geometry and particle population model. This naturally includes calibration of the particle material model.

Based on the outcome of the first step, conceptual solutions should be developed using conventional engineering design methodologies. It is recommended to generate multiple concepts to cover a larger solution space. In order to quickly evaluate the basic performance and behaviour of each concept, simulations may be carried out with a lower simulation fidelity level. At this point the approximate functionality of a concept is of interest, not the exact behaviour or performance. The simulation and evaluation activities may be performed iteratively at successively increased fidelity level. When a sufficient information quality is achieved the concept or concepts with the highest performance are selected. The selection should be made with the DEM simulation results evaluated together with other requirements such as e.g. cost, complexity, serviceability, safety, manufacturability. The decision may be aided by selection and scoring matrices as described by e.g. Ulrich (2008).

In the third and final step the chosen concept or concepts may be optimized in regards to performance objectives. It is common in particle flow applications that small differences in geometrical features have a significant influence on the material flow behaviour. A sensitivity analysis may also be performed in order to gain understanding, not necessarily of the best exact position of a feature, but more importantly within what interval it needs to be adjustable during implementation.

Depending on the development situation, all steps in the process may not be performed. A problem solving situation may require only step i and an optimization exercise in step iii, skipping the conceptual development step.
Figure 20. Process flow diagram for the use of DEM in design and engineering
5.1 Example: Cone Crusher Feeding Problem

In Paper B the work flow process was tested and evaluated. The case chosen for the study was a feeding operation of a cone crusher. The plant observed significant issues related to miss-aligned feed and segregation effects. After completing the first process step the phenomenon of the feeding problem was mapped and understood in principle. A set of concepts was developed and evaluated by means of low fidelity simulations. It was discovered that the best performing concept for minimizing the effects of miss-aligned feed and segregation was a static cylindrical smaller hopper above the crusher hopper, see Figure 21.

![Figure 21. Left: Low fidelity simulation result when feeding directly. Right: Simulation result when evaluating an early conceptual solution situation, in this case a static hopper.](image)

5.2 Example: Development of Cone Crusher Liner Design

In paper H the process in Figure 20 was used to solve a flow restriction problem in a laboratory cone crusher. The crusher was refurbished and rebuilt to incorporate frequency drives allowing for operation at significantly higher levels than normal. During testing it was found that a mass flow restriction and wear problem occurred when operating at high eccentric speeds. The machine was analysed according to step i in order to provide deeper understanding of the problem. Through DEM analysis it was found that the original liner design had a geometric feature in the end-zone that resulted in a pocket region where it was a low probability for the particles to escape from, see Figure 22.

A set of new liners were designed and DEM was used to evaluate the performance iteratively according to step ii, see Figure 23. The new liner geometry provided the ability to successfully run the crusher at high eccentric speeds without any flow restriction issues. The decision to go forward with manufacturing of the liners was based on the knowledge gathered through the DEM analysis. The DEM analysis also provided vital information regarding the breakage behaviour when operating at high eccentric speeds up to 20 Hz, which is about four times the normal operating speed and previously unknown operating range.
Figure 22. DEM simulation of flow restriction problem in the laboratory cone crusher

Figure 23. Snapshot from DEM simulation of the new liner design
6 Verification and Validation

Currently, the assessment of whether comminution and crushing equipment under development satisfies stated functional requirements is commonly expensive due to the need for numerous resource intensive physical experiments. Computer modelling and simulation-based methods for virtual comminution analysis and design evaluation currently have an increasing impact on the time and cost effectiveness of the development process (Forrester, Keane, & Sóbester, 2008). To accomplish the benefits of virtual testing, simulation-based predictions require verification and validation. Evidently, the virtual testing must also provide the sought insight and knowledge at a lower cost and time compared to physical tests.

The ASME (American Society of Mechanical Engineers) guide for Verification and Validation in Computational Solid Mechanics (2006), provides the following definitions,

- Verification: The process of determining that a computational model accurately represents the underlying mathematical model and its solution.
- Validation: The process of determining the degree to which a model is an accurate representation of the real world from the perspective of the intended uses of the model.

The desired outcome of the verification and validation (V&V) process is to validate the model for its intended use (ASME, 2006; Sargent, 2013). This is an important distinction to make in the context of design engineering. The model builder needs to define a set of predetermined requirements setting the constraints for the accuracy and predictive capability. The accuracy constraint may in some cases be balanced against computational load in a trade-off exercise. Meaning that some modelling accuracy is intentionally sacrificed in order to run fast simulations. This model fidelity reduction has to be accounted for in the decision making process. It is hence also advisable to exercise some caution regarding which stakeholders are to be included, and how they are informed, before entering the decision making process.

An example of this strategy is demonstrated in Paper B, previously described in §5 where a lower fidelity DEM particle material model is used for preliminary virtual experiments of several different conceptual design solutions for a cone crusher feeding problem. In many cases when dealing with particle flow problem solving it is difficult to, based on engineering heuristics and experience, make a decent prediction regarding the particle trajectories and segregation behaviour. This is especially the case when a stream of particles interacts with equipment put in place with the intended function to guide or mix the material. For initial conceptual evaluations it is advantageous to perform quick simulations with computational times below 2-15 minutes. This allows for iterative adjustment of geometrical features (shape and position) until the conceptual solution reaches a sufficient performance to be further analysed.
As a part of the previously mentioned study in Paper B, the design stage process is proposed for particle flow problem solving using DEM, see Figure 20. In this process the first step is to perform a high fidelity simulation with the purpose of understanding the current problem. For this step it is not advisable to reduce the model fidelity in order to save computational time. This analysis stage is important in order to reproduce the issues reported and measured in operation. If the operational problems (e.g. misaligned feeding, segregation effects or blockages) cannot be reproduced, the modelling needs to be revisited, improved and iterated. This kind of direct visual semi-validation leads to an interesting point regarding alternative validation methods for cases where it is expensive or not possible to perform direct measurements on complex and disruptive particle flow conditions.

Other methods for validation than to mathematically compare datasets are mentioned in the model validation research field. A number of different validation techniques have been listed by Sargent (2013). One of the approaches mentioned is defined as ‘face validity’ and is explained as validation based on the engineering individuals knowledge and experience about the system. In the case of DEM modelling and simulation, experienced engineers and researchers may comment that a simulation appears to be realistic. Such a comment obviously does not automatically validate a simulation model. However, in cases where direct measurement is difficult and expensive, it might still be of value to the model developer.

According to the terminology and framework proposed by Schlesinger et al (1979) validation may also be done in regards to the conceptual model of a system. There are several examples in this work where the simulation results have been evaluated in regards to an expected behaviour based on an already established conceptual model:

- The observed edge effect of the pressure distribution from the HPGR simulations in Paper D. Described by e.g. (Daniel & Morrell, 2004; Schönert & Sander, 2002; Torres & Casali, 2009) and measured by (Lubjuhn, 1994).
- The observed choke level in the cone crusher and its dependence on the position of the minimum cross sectional area in the crusher chamber (Paper A, C, F).
- The observed phenomenon of inter particle breakage above the choke level and single particle breakage happening below the choke level in (Paper A, F). As proposed by Evertsson (C.M. Evertsson, 2000).
- The relationship between the eccentric speed and the number of breakage zones (Paper A). As proposed by Evertsson (C.M. Evertsson, 2000).

Due to the difficulty of measurement directly in an industrial operation, the process validation needs to be decomposed into manageable sub-models. This decomposition into a hierarchical approach to validation is discussed in the ASME standard (ASME, 2006). A bottom-up approach, exemplified by a vehicle model application, is suggested where the lowest level of models, called component models, are validated using experiments. The component level is followed by subassembly models, assembly models, system models, finally achieving a validated model applicable to the top level reality of interest.
A similar bottom-up approach has been presented in paper D for DEM applications, see Figure 24. The model is based on the V-model normally found in product development methodologies for breaking down requirements, relating verification and validation test on different levels back to the corresponding defined requirement. In this case the subject of decomposition is changed from requirements to different levels of test environments. The top level is defined as full scale, normally referring to industrial operation measurements.

The lowest level refers to experiments on single properties that could be for instance coefficient of restitution, static friction, or particle strength tests. The mid-layer is emphasised in this work as an important validation step to perform. There is no guarantee that single property experiments and calibration, result in a valid particle bulk behaviour. When all model components come together including geometrical features, friction, rolling friction, cohesiveness, particle shape and size distribution, the resulting combined behaviour needs to be analysed.

6.1 Surrogate Modelling

DEM simulations are most often long running expensive computations. This leads to a problem in research or design analysis processes; there are never enough resources evaluable to investigate all possible variable combinations of interest. This problem becomes increasingly apparent when attempting to apply optimization schemes. As a remedy to this problem the design engineering community has for a long time suggested the use of surrogate models. The models are commonly also referred to as meta-models or response surface models.

A comprehensive overview of different surrogate modelling approaches has been presented by Forrester and Keane (2009). Surrogate modelling and optimization have developed as a consequence of situations where several variables are included in the study and where each simulation is expansive. It is simply not possible to either explore all possible variable combinations, or to apply direct search methods on the complete simulation model. In these cases an efficient strategy is needed where as much information as possible is retrieved from a limited set of simulations. Based on these carefully chosen variable combinations a surrogate model may be constructed, functioning as an interpolation between the data points gathered.
The first step in the response modelling process is to decide what parameters to explore and vary - in other words the variables. In some cases this choice may be obvious, however in many situations the first question is more targeted towards what parameters influence what responses. These relationships may be analysed and determined by performing sensitivity analysis (in some cases called statistical screening analysis) using design of experiments. Experimental plans such as for instance fractional factorial designs or Placket Burman designs are efficient while they have the inherent drawback of sacrificing information regarding higher order variable interactions (Box, 2005). These designs are useful in order to bring clarity regarding:

1. What variables have a statistically provable effect on what response variables?
2. From the identified significant variables, which ones are the most influential and which variable interactions are active?

The modeller needs to understand these relationships before proceeding with response modelling. In some cases these relationships may be already known a priori and the modeller may directly focus on experimental designs more suitable for response modelling. The response surface modelling experimental design may be developed as an extension of the screening design, or, a new separate experimental design dedicated for response modelling is created. The extension of an existing experimental plan can be realised through the concept of design augmentation (SAS Institute, 2015) (called sequential assembly in Box (2005)). The augmentation enables a controlled way of expanding the design plan while maintaining each variable column mutually orthogonal. Variable column vectors are mutually orthogonal if the sum of the products of their corresponding elements is 0. Orthogonality is important in design of experiments since it allows for analysis of main effects and interactions independently.

The four most common experimental plans for response surface modelling are shown in Figure 25. The full factorial design (here presented as a $2^5$ design) is perhaps the easiest to apply and understand. A more efficient space filling design is the Latin Hypercube design, in this example displayed using 20 points. In contrast to the full factorial design the Latin Hypercube is based on defining how many experiments that should be performed. The algorithm then positions the points in order to achieve an optimal spacing and filling. The central composite design (CCD) expands the full factorial design with a centre and star points (Box, 2005). The Box-Behnken design is an incomplete three-level factorial design where the points are specifically chosen to allow efficient estimation of coefficients for second order polynomial models. Points are not positioned in the corners however, instead they are inserted on the midpoints of each edge of the experimental space.
6.2 Polynomial Regression

Even though modelling methods such as artificial neural network, Kriging and radial basis functions (RBF) are becoming more and more popular in the engineering design field, polynomials remains one of the most used surrogate modelling techniques (Forrester & Keane, 2009). In this work neural network models were tested in both Paper F and Paper I. However, in both cases it was found that while the models displayed good correlation, the models tended to over fit the data when the surface profiler tool was used to inspect the functions in SAS JMP® 12. Hence, polynomials have been the basis for the surrogate modelling for the calibration. An advantage with polynomials is that the fitted coefficients have a direct relationship to the factor contrasts provided from the screening analysis. This enables verification of the model in terms of the relationship between a variable and an expected response. As an example, a positive coefficient should be expected for the normal stress limit variable, so that the predicted tensile strength of the rock increases when increasing the stress limit. Similarly, if reducing the particle size variable, it should result in increased number of particles (given an equal volume creation) which should be seen as an increase predicted simulation time.

A polynomial approximation of a function $f$ of dimension $n = 1$ and order $m$, can be written as,

$$\hat{y}(x, \alpha, m) = a_0 + a_1 x + a_2 x^2 + \ldots + a_m x^m = \sum_{i=0}^{m} a_i x^{i}$$

(24)
The coefficients \( \alpha = \{ \alpha_0, \alpha_1, ..., \alpha_m \}^T \) are estimated through the least square solution of \( \Phi \alpha = y \), where \( y \) is the vector of observed responses and \( \Phi \) represents the Vandermonde matrix,

\[
\Phi = \begin{pmatrix}
1 & x_1 & x_1^2 & \cdots & x_1^m \\
1 & x_2 & x_2^2 & \cdots & x_2^m \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
1 & x_n & x_n^2 & \cdots & x_n^m 
\end{pmatrix}
\]

(25)

The coefficients \( \alpha \) is calculated by the maximum likelihood estimation method according to,

\[
\alpha = (\Phi^T \Phi)^{-1} \Phi^T y
\]

(26)

The order \( m \) should be chosen carefully since including higher order terms usually improves the correlation. However, the increased flexibility of more terms may lead to over fitting noise present in the response. Such a model may also result in a poor generalization with many local minima that complicate the optimization. In this work the choice of \( m \) has been guided by the variable sensitivity/screening analysis.

6.3 General Model Calibration

The backbone of the validation process is to calibrate the model components. The concept of calibration of simulation models dates back to the early days of computer simulation, see for instance Barker and Watson (1969) and Wigan (1972). Calibration can be defined as the task of adjusting an already existing model to a reference system (or in some cases a trusted reference model). The reference system commonly refers to experimental measurement data. This is done by adjusting the internal model parameters until a satisfying congruence between the model and reference system is accomplished.

The adjustment may in very trivial cases be done by trial and error however in most situations a systematic approach is favourable. The most commonly applied strategy for calibration of DEM models is to apply a design of experiments approach.

When working with calibration in practice, most researchers and engineers will realise that there is a hidden complexity in defining under what circumstances it is possible to claim that a simulation model actually matches the reference system. These questions have been extensively explored in a paper by Hofmann (2005) where the author proposes a set of definitions that encapsulates a comprehensive calibration framework. In brief, Hofmann describes two theoretical states of congruence:

- **Strong congruence** - Denotes that there exists at least one model parameter configuration, so that for all possible system inputs the outcome of the model application equals the output of the system.
Weak congruence - Denotes that for every possible system input there exists at least one parameter configuration of the model, so that the outcome of the model application equals the output of the system.

The weak congruence definition implies that every time something in the reference system changes, the model needs to be recalibrated. In contrast, the strong congruence says that the model is perfectly valid for all possible changes in the system. None of these states are likely to be useful in a design context. The strong state is highly unlikely if not impossible to achieve for complex systems, and the weak state eliminates any predictive capability.

These two states clearly represent two extremities and some kind of constraint relaxation is needed in order to arrive at a useful congruence definition. Hofmann hence proposes to allow a problem specific error or tolerated deviation $\Delta_j$ and call this state pragmatic congruence. The model is then pragmatically congruent if the error, or distance, between the reference system (R) output and simulation (S) prediction output is less than the allowed error according to Eq. (27) (Hofmann, 2005).

$$|y_{i,j}^R - \hat{y}_{i,j}^S| \leq \Delta_j \quad \text{with} \quad \forall j \in \{1, ..., k\} \quad \forall i \in \{1, ..., I\}$$

(27)

Where $j$ represents the index for the response variable and $i$ represents the index for the samples. The samples may either for instance refer to data-points in a times-series, or discrete outputs from different operating conditions. In paper D and E discrete responses have been used such as for instance rock strength and stiffness in the case of breakage calibration and average mass flow and response angle in the flow calibration case.

### 6.4 Error Measures

There are several error measures and metrics defined for quantifying how well two systems, models or data-sets correspond to each other. In the work by Sarin et al (2010) a wide range of different error measures are investigated for the purpose of validation of simulation models based on time histories. In the case where two time discretised history vectors $A$ and $B$ of equal size $N$, a popular approach is the $L_p$ norm which measures the difference between the two vectors according to,

$$\|A - B\|_p = \left( \sum_{i=1}^{N} |a_i - b_i|^p \right)^{\frac{1}{p}}$$

(28)

The three most commonly used norms are the $L_1$, $L_2$ (Euclidean norm) and $L_\infty$. When applying the different norms on the same time historical data, Sarin et al (2010) showed that the use of a different norm may lead to different conclusions since the resulting error estimation differs. As pointed out by the authors, the major limitation for using norms for time histories is that they do not distinguish errors due to phase shift from error due to magnitude.
In this work the calibration and validation operations are not performed on time-series but rather on a vector of aggregated response variables extracted from the bulk behaviour. As further described below the A vector from the example above here corresponds to the experimental result dataset and the B vector the corresponding simulation-based surrogate model response. The Euclidian $L_2$ norm has been chosen as the error measure hence Eq. (28) becomes:

$$\| A - B \|_2 = \sqrt{\sum_{i=1}^{N} |a_i - b_i|^2}$$  \hspace{1cm} (29)

This general error measure is applied to the case of model calibration by defining the error function according to Eq. (30).

$$\|E(x)\| = \sqrt{\sum_{j=1}^{N} \varepsilon_j(x)^2}$$  \hspace{1cm} (30)

Where,

$$\varepsilon_j(x) = |\bar{y}_j - \hat{y}_j(x)|$$  \hspace{1cm} (31)

The index $j$ refers to the $j^{th}$ response variable. The notation $\bar{y}$ refers to the scalar mean experimental response and the notation $\hat{y}$ refers to the prediction response from the simulation based surrogate model. As an example, in the bonded particle model calibration, two response variables are included in the calibration. The two are the tensile strength ($y_1 = \sigma_t [Pa]$) and the stiffness ($y_2 = k [N/m]$). When working with the optimization formulation for the BPM calibration it was found that the difference in magnitude of the different response variables generated an unwanted bias. The formulation in Eq. (31) is hence adjusted in order to normalize the expression:

$$\varepsilon_j(x) = \frac{|\bar{y}_j - \hat{y}_j(x)|}{\bar{y}_j} = 1 - \frac{\hat{y}_j(x)}{\bar{y}_j}$$  \hspace{1cm} (32)

Equation (32) constitutes the error measure found to be most useful for calibration with respect to response variables with no trade-off relationships.

### 6.5 Calibration Optimization Formulation

Model calibration is essentially about finding the values for each variable in the vector $x$ that gives the smallest error possible in Eq. (32). This is inherently an optimization problem no matter how sophisticated a method the modeller uses to find the optimizer $x^*$. Since there are optionally multiple response variables included in the expression the optimization problem will be defined as a multi-objective function combining Eq. (30) and (32). The general calibration optimization formulation on negative null form (Papalambros, 2000) becomes:
\[
\text{minimize } f(x,p) = \sqrt{\sum_{j=1}^{N} \left( 1 - \frac{\tilde{y}_j(x,p)}{\bar{y}_j} \right)^2}
\]

subject to \( g(x,p) \leq 0 \)
\( h(x,p) = 0 \)
and \( x \in X \subseteq \mathcal{R}^n \) \( j = 1, 2, \ldots, N \) \hfill (33)

On this form, \( g \) and \( h \) are the functional constraints, inequality and equality constraints respectively, and \( X \) is the set constraint which is a subset in \( \mathcal{R}^n \). The vector \( p \) in the formulation refers to the modelling parameters. It should be noted that depending on the optimization strategy used these parameters could potentially mean different things. In the case of surrogate-based optimization where a response model is used, the parameters would refer to for instance the fitted least square polynomial coefficients or the weight and bias coefficients for the sigmoid functions in a neural network. However, if a direct search optimization algorithm would be linked to a DEM solver, the parameters could refer to DEM model parameters such as contact model and geometry parameters.

In cases where errors for different response variables have a conflicting trade-off relationship, the formulation in Eq. (33) may not be suitable. An objective function based on the weighted function approach (Papalambros, 2000) can be written as

\[
f(x,p) = \sum_{j=1}^{N} w_j \epsilon_j(x,p)
\]

Where,

\[
\sum_{j=1}^{N} w_j = 1
\]

(35)

The expression is similar to the \( L_1 \) norm, however in addition a set of weights \( w_j \) are added in order to explore trade-offs. The corresponding optimization formulation on negative null form is then:

\[
\text{minimize } f(x,p,w) = \sum_{j=1}^{N} w_j \epsilon_j(x,p)
\]

subject to \( g(x,p) \leq 0 \)
\( h(x,p) = 0 \)
and \( x \in X \subseteq \mathcal{R}^n \) \( j = 1, 2, \ldots, N \)

(36)

When using this form the weights \( w_j \) may be subject to a parameter sweep in order to generate the attainable set \( \mathcal{A} \). The set of solutions may be reduced to a subset of the attainable set, termed the Pareto set, consisting of Pareto optimal points. A solution point \( f_x \) in the attainable set \( \mathcal{A} \) is Pareto optimal if and only if there is no another \( f \in \mathcal{A} \) such that \( f_i \leq f_x \) for all \( i \) and \( f_i < f_x \) for at least one \( i \) (Papalambros, 2000; Pareto, 1971). The approach of using weighted sum of squares
has previously been applied for calibration of contact model parameters by Kruggel-Emden (Kruggel-Emden et al., 2007).

6.6 Bonded Particle Model Calibration

In order for a breakage model to be useful in a machine simulation setting it needs to be calibrated against experimental data; regardless of the specific choice of modelling approach. Several studies have contributed to the topic of calibration of DEM models. However they may be divided into two main areas where the first concerns applications related to granular flow. In this area several researchers have made important contributions, see for instance (G. K. P. Barrios, de Carvalho, Kwade, & Tavares, 2013; C. J. Coetzee, 2017; C. J. Coetzee, Els, D.N.J., 2009; C. J. Coetzee & Nel, 2014; Combarros, Feise, Zetzer, & Kwade, 2014; Favier, 2010; Frankowski et al., 2013; González-Montellano, Ramírez, Gallego, & Ayuga, 2011; Grima, 2009, 2011; Gröger, 2006; J. Quist, 2015).

In the area of calibrating bonded particle models, Hanley (2011), Yoon (2007), Wang and Tonon (2009) as well as Jiang et al (2015) have made important advancements by introducing statistical design of experiment methods and calibration optimization. In the calibration procedure demonstrated by Wang and Tonon (2009), sensitivity analysis is performed by varying one factor while fixing the others. This approach is simple and requires few simulations, however the method does not capture the second and third order interaction relationships between micro parameters. Even though the researchers succeeded in modelling the breakage behaviour well for both unconfined uniaxial strength and direct tensile strength tests they concluded that only two parameters are not sufficient for achieving a good calibration outcome. In the work presented by Hanley (2011), ANOVA analysis was used to analyse the responses from 3D simulations of particle agglomerate compression with micro parameters configured by a Taguchi L27 array (27 runs with 13 columns). The results from the study displayed reasonable prediction capabilities for some of the response variables. It was however concluded that more simulations, hence a larger experimental design plan, was needed in order to resolve significant factor aliases.

In the work by Yoon (2007) the design of experiments approach is both well described and applied by using both Placket-Burman (PB) designs for factor screening and Central-Composite Designs (CCD) for response surface modelling. The proposed approach is based on first applying a PB screening design in order to evaluate which micro parameters are statistically significant for each response variable. When this has been done a CCD is created in order to fit response models for each output response variable. The results show good predictability and correspondence to experimental data. However, the simulations are performed using 2D DEM code hence it is not possible to utilize the outcome directly for 3D DEM simulations.
Figure 26. The developed general calibration framework as implemented for calibration of the bonded particle model (BPM)
In Paper I the general statistical and optimization framework developed by Yoon (2007) is expanded by introducing 3D DEM and a multi-objective optimization approach for calibration. Two main objective functions are included in the multi-objective optimization formulation. The first objective function is formulated according to Eq. (33) including the response variables tensile strength $\hat{y}_1$ and bulk stiffness $\hat{y}_2$. The second objective function models the computational time and a fracture quality property. Since DEM simulations in three dimensions are computationally demanding, it is of interest to explore the trade-off between the calibration error, the computational time and the breakage quality.

The calibration procedure follows the diagram in Figure 26. The calibration follows two tracks; experiments and simulations and is concluded by the optimization stage. The fracture test used in this case is the Brazilian indirect tensile strength test (ASTM, 2008). The experimental setup can be seen in Figure 27 showing a granite rock core specimen. Two flat surfaces were used for the loading plates. It should be noted that the contact boundary condition influences the estimation of the tensile stress, were a small angular strip results in an underestimation of the tensile stress (Fairhurst, 1964; Perras & Diederichs, 2014). In order to prepare the rock core specimens a diamond core drilling machine was used for extracting specimens out of ~200 mm rock specimens sampled from the same blast batch as for the material used in Paper H. The result from the Brazilian breakage experiments can be seen in Figure 28.

The concept of sequential assembly of the design of experiment test plan was used iteratively until the response model converged, resulting in a total of 102 simulations including nine factors. The stress-strain curves for 16 of the performed simulation runs can be seen in Figure 29. As can be seen the breakage process does not match the smooth monotonically increasing (to failure point) characteristics of the stress-strain curves from the experiments. An example of a breakage sequence for one of the simulations can be seen in Table 1.
Figure 28. Experimental result of the breakage of 10 core specimens.

Figure 29. Simulation results for 16 simulation runs with different BPM settings.
Table 1. Breakage sequence for one of the performed simulations in the DOE simulation plan.

<table>
<thead>
<tr>
<th>$t_{\text{break}} - 3t_s$</th>
<th>$t_{\text{break}} - 2t_s$</th>
<th>$t_{\text{break}} - t_s$</th>
<th>$t_{\text{break}}$</th>
</tr>
</thead>
</table>

The multi-objective optimization algorithm was used to generate pareto-fronts displaying the relationship between first objective function $f_1$ (experimental congruence calibration error) and second objective function $f_2$ (simulation time and fraction of broken bonds). The weight factor $w_2$ controls the priority between simulation time and breakage or fracture pattern quality. The fracture quality is modelled using a proxy parameter based on the ratio of broken to non-broken bonds. A distinct fracture crack is desirable and the assumption is that this ratio parameter can capture and respond to if there is a narrow or wide crack pattern. The pareto-sets from the optimization, presented in Figure 30, show that an increase in simulation time produces a lower congruence error. This indicate that in a design engineering context, the DEM model developer can make an active decision regarding computational cost and the corresponding accuracy. If there is less time available, the decision making process needs to consider a lower predictive accuracy capability.

Figure 30. Set of feasible non-dominated points forming a Pareto front for three levels of $w_2$. 

![Figure 30](image-url)
6.7 Particle Flow Calibration

In Figure 31 the process for calibration of DEM models that have been developed in Paper F is presented. In principle it has the same layout as the diagram in Figure 31, however presented with less complexity. In the first stage the material characteristics are measured. The second stage is described as a parallel process of performing experiments and simulations. The actual calibration is realized by an optimization routine where the error between the experimental response and simulation response is minimized as previously presented.

![Figure 31. Schematic flow diagram of the proposed calibration process](image)

6.8 Calibration Device

A new device has been designed and built for the purpose of contact model calibration of granular material. An illustration of the device can be seen in Figure 32. In order to be able to measure several flow properties in the same experiment the device is built up by different features. Each feature can be individually adjusted in order to create different flow regimes. The concept of performing experiments at different flow regimes is based on the idea that the calibrated material model should be unbiased by one fixed flow condition.
Figure 32. Illustration of the calibration device configuration including possible adjustments that allows for different flow regimes.

The experiment is conducted in the following sequence:

i. Hopper angle, sliding plane angle and top section height is adjusted
ii. The material sample is placed in the hopper
iii. High speed camera is triggered ON
iv. Trap door mechanism released
v. Material flows through hopper
vi. Material slides on sliding plane
vii. Material bounces on left wall
viii. Material settles forming a sloped bed
ix. High speed camera triggered OFF
x. Repose angle of sloped bed is manually measured
xi. Material is discharged

The flow passing through the device can be monitored directly due to the transparent front glass sheet. In Figure 33 an example of the flow in the experiment and the corresponding flow in the simulation are shown. As the figure indicates it is possible to subjectively make a judgement if the simulated material flows and behaves in the same way as the experiment. However, a subjective judgement is not good enough if the objective is to calibrate the model in a more strict sense. The response variables measured from the experiment are the mass flow through the hopper and the repose angle of the bed formation. An example of a bed formation is shown in Figure 34.

A high speed camera is also used to capture the flow for further analysis using motion tracking. This gives the possibility to measure the actual flow pattern of particles, not only indirect bulk flow properties. Guiding experiments shows that colour-marked particles can be tracked as part
DEM MODELLING AND SIMULATION OF CONE CRUSHERS AND HIGH PRESSURE GRINDING ROLLS

of the bulk material. Six particles have been marked and positioned in the same pattern for each flow experiment. In Figure 35 an example of particle flow profiles in the x- and y-direction is presented. The red and blue colour represents two different flow regime settings and the particles tracked are placed in the same position at test initiation.

The particles tracked in the plot are placed cantered on top of the trap door latch. The particle route through the device can be described as a sequence as seen in Table 2. The experiment represented by a blue line (R4) in the plot has a flow regime setting with a steeper sliding plane angle than the red line experiment (R1).

Figure 33. The left hand side image shows a frame from a recording of a real experiment after the first stream of particles has travelled through the device. The right hand side image shows the corresponding flow in the simulation environment.

Table 2. Flow sequence of marker particle in Figure 35 through the device.

<table>
<thead>
<tr>
<th>Seq.</th>
<th>Time [s]</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0 – 0.16</td>
<td>Free fall flow out of the hopper</td>
</tr>
<tr>
<td>2</td>
<td>0.16 – 0.4</td>
<td>Particle slides on the sliding plane</td>
</tr>
<tr>
<td>3</td>
<td>0.4 – 0.8</td>
<td>Particle bounces on the left wall and settles in the pile formation</td>
</tr>
<tr>
<td>4</td>
<td>0.8 – 1.6</td>
<td>Particle is settled</td>
</tr>
</tbody>
</table>
Figure 34. Image of the repose angle formation at the end of a flow experiment.

Figure 35. Example of motion tracking data showing x- and y-coordinate positions during the flow for a marker particle with the same starting position.
The developed method for direct measurement of the particle flow is promising and gives a good subjective understanding. However, so far the method has not been developed far enough to enable a statistical comparison to simulation data. One identified issue is that the random variance in the trajectories was high due to the particle shape, initial starting positions and orientation of particles in both the simulation and experimental case. In order to utilize the high-fidelity trajectory data, a higher level of automated data analysis for both motion tracking trajectory data and simulation trajectory data is needed. Due to this reason the bulk flow measures in terms of mass flow and repose angle was used as response variables for the calibration.

The calibration outcome using the device and process have been successful in terms of calibrating size, shape, static friction and rolling friction. However, it was found that the coefficient of restitution did not have a statistically significant effect on the mass flow and repose angle response variables. When analysing the simulation results it was found that the coefficient of restitution clearly affected the flow behaviour, however mostly in the interaction with the sliding plane and most importantly with the left wall. Hence, in order to improve the calibration method an idea would be to utilize the trajectory data. If an algorithm could identify a set of particle-particle and particle-wall collisions, the velocities before and after the collisions could be calculated, providing a data-set where the restitution behaviour is captured with a good statistical basis. Using this data, an estimate of the average coefficient of restitution could possibly be calculated.
7 Cone Crusher Modelling

7.1 Cone Crusher Eccentric Speed

The research objective for the study in Paper A was to investigate the influence of eccentric speed on the crusher capacity. The BPM was used to bond together fraction particles to meta-particles with the shape of spheres. The spherical shape of the meta-particles is of course not representative of the irregularly shaped rock particles normally feeding a crusher. However, the simplistic shape results in distinct and distilled flow behaviour that is of interest when looking at the principle dynamics of the particles in the crushing chamber.

Evertsson (1999) established an analytical framework for modelling flow of particles in a cone crusher. From this work it can be derived that if the eccentric speed is increased the number of breakage zones and events will increase. The obvious reason is that the mantle will interact with the particles more frequently. Further on, particles will have less time to accelerate and move down the chamber between each compression event.

Five different eccentric speed settings were evaluated by simulating a 40 degree section fed by a batch of BPM meta-particles. The crusher type was a Sandvik CH430 equipped with a fine chamber, 12 mm close side setting and 36 mm eccentric throw. The eccentric speed was varied from 269 to 579 rpm. In Figure 36 an image from the 269 rpm simulation is shown.

In Figure 37 two detail images of the crushing chamber are presented. The left image shows the particle trajectory of a randomly chosen fraction particle. The trajectory path suggests that the particle has been subjected to at least 4-5 compressions before leaving the chamber. An important feature of the trajectory is that the direction of the particle is sometimes turning back upwards. The effect of a particle moving upwards during compression instead of getting broken is referred to as squeezing by Evertsson (2000).

In the image to the right the particles in the chamber are represented by bonding beams. From this view as well from the view in Figure 36, it can be seen that below the choke level, only single particle breakage occurs. An important qualitative simulation result is that meta-particles break apart sequentially. In the early phases in this work it proved to be difficult to find a simulation setup where particles didn’t break apart totally during the first compression. The successful modelling of sequential breakage is one of the key findings in Paper A.
Figure 36. Overview of CH430 Cone Crusher simulation at 259 rpm

Figure 37. Crushing chamber simulation details. The image to the left shows a single fraction particle trajectory when moving through the crushing chamber. The right image shows the meta-particles represented as bonds.

In Figure 38 the throughput capacity is plotted for the five eccentric speed settings simulated. As can be seen there is a clear decreasing trend with increasing speed. This result corresponds well to the findings by Evertsson (1999).
The main finding in Paper A is that the conceptual idea, previously based on analytical closed form expressions, that the flow through the crusher is limited at the position of the minimum cross-sectional area, can be proved using numerical models. In the same way it has been demonstrated that the eccentric speed controls the number of compressions events. The integrated dynamic behaviour from all compression events further on has an effect on the mass flow superimposed on the effect due to the choke level minimum cross-sectional area.

### 7.2 Primary Gyratory Crusher - Influence of Liner Wear

In Paper C a primary gyratory crusher is investigated in terms of the influence of liner wear on crushing performance. The performance is evaluated with respect to the throughput capacity and the bond breakage rate. A primary gyratory crusher was 3D scanned on two different occasions, when liners were new and when they were about to be replaced. Three different cases are simulated:

- New liners
- Worn liners (without CSS compensation for wear)
- Worn liners (with CSS compensation for wear)

In Figure 39 the simulation of the worn case is shown. As can be seen the BPM model is used with meta-particle shapes based on 3D scanned rocks. The feed size distribution was chosen based on input data from the site operation.

Figure 40 shows the total mass within the simulated crusher cavity over time. In essence, the derivative of the linear regression line plot illustrates throughput capacity. As can be seen the capacity is higher when the concave is worn as compared to when the concave is new. As the gap is compensated in the worn case to meet the target CSS the capacity decreases. However, due to the change in chamber profile and cross-sectional area the capacity is still higher in the worn
compensated case compared to the new case. This implies that the gap is not the only limiting parameter, but the chamber geometry and choke level area also have an effect.

![Image from simulation of the worn compensated case showing the bonds forming particles and colored by the level of normal force. Beams with black color are representing recently broken bonds.](image)

**Figure 39.** Image from simulation of the worn compensated case showing the bonds forming particles and colored by the level of normal force. Beams with black color are representing recently broken bonds.

![Total mass within system domain over time. The trend line slope indicates the mass flow rate for the three simulated cases.](image)

**Figure 40.** Total mass within system domain over time. The trend line slope indicates the mass flow rate for the three simulated cases.

In Figure 41 the ratios of the number of intact to broken bonds over time are plotted. The derivative of the regression line represents the bond breakage rate. The shape of the plotted curves has a characteristic staircase pattern resulting from the sequential compression. This pattern is a consequence of that a section of the crusher has been simulated. The breakage rate is highest for the worn uncompensated case. The reason for this can be found when observing the simulations more closely. The simulated particle feed size distribution fits very well in the gap
between the liners when liners are worn. Particles hence fall further down in the cavity and are broken. In the other two cases the gap is slightly less open and the particles have a higher tendency to be squeezed upwards instead of being crushed.

![Graph](image.png)

Figure 41. Ratio of intact bonds over time. The trend line slope indicates the bond breakage rate for the three simulated cases.

The main finding of this work is that 3D scanned geometry was successfully implemented in a gyratory crusher DEM simulation and for the first time reported. It is also the first time meta-particles based on real rock shapes have been used. The simulation results show that liner wear has a dramatic influence on crusher performance in terms of throughput capacity and breakage rate.

### 7.3 Cone Crusher Close Side Setting

In Paper G all the levels in the V-model, see section 6, are tackled and explored for a cone crusher application, see Figure 42. The operation studied was a Svedala H6000 secondary crusher at an aggregate plant. The material model was calibrated both in regards to flow properties using the calibration device and in regards to breakage using single particle breakage experiments.

The crusher mantle and concave geometry was modelled based on 3D scanned point clouds of the real physical geometry. The experiments performed on site included high frequency sampling of the power draw and hydraulic pressure signals which revealed an important insight on how the feeding influences the crushing. In the signals a distinct cyclic pattern at 5 Hz can be observed. This frequency is the same as the eccentric speed of the crusher indicating a strong causality between the misaligned feed visually observed and the pattern seen in the data.
The simulation was configured so that a section of the crusher was modelled. The main reason is to keep the total number of particles below a certain limit in order to keep the simulation time within possible ranges. Simulating a section of the crusher can be regarded as forcing a very distinct miss-aligned feed into the crusher, only feeding on one side. If the crusher operates in nominal conditions, rotational symmetry is normally expected.

As seen in Figure 43 the cyclic behaviour of the pressure signal is very distinct in both the simulated case and the experimental data. The same cyclic behaviour can be seen for the power draw but the correspondence in terms of amplitude does not match equally well as the pressure, see Figure 44. Due to the modelling of a section of the crusher the comparison should be made with regards to the envelope of the peak of each cycle in the simulated pressure and power draw. The peak envelopes are in the same region of operation as the measured data.
Figure 43. Measured and simulated hydrostatic pressure for 1.5 seconds of operation.

Figure 44. Measured and simulated power draw for 1.5 seconds of operation.
In order to be able to calculate the product size distribution of the broken meta-particles a new method was developed based on image processing. Images of the discharge region are exported and the size and shape of the clusters are evaluated using the image processing tool ImageJ (Rasband, 2015). The calculated size distribution, see Figure 45 shows that the 34 mm CSS case results in a finer distribution then the 50 mm CSS case. Qualitatively there is a similarity in the trend, however the quantitative results could be improved.

The main finding in paper G is that it is the first work published that compares a simulated cone crusher with experimental data on industrial scale level. It is also the first published, at least known to the author, method to evaluate the cluster size distribution of broken BPM meta-particles in a crusher application.

Figure 45. Comparison between measured and simulated product size distribution.
8 HPGR Modelling

The High Pressure Grinding Roll (HGPR) machine utilizes the inter-particle compressive breakage mechanism at high pressure levels and has been a popular subject for modelling efforts the last couple of years. The mechanical layout of the machine is similar to metal rolling machines. As a consequence there are also similarities in terms of defining the governing equations for the relationship between the mechanics and the material. However, due to the different nature of granular brittle solids compared to deformable hot metal, several boundary conditions, constitutive relationships and continuity assumptions are not the same. When modelling metal rolling it is possible to utilize continuum based FEM models, due to the discrete nature of granular materials these modelling techniques are not applicable. Particle based simulation methods such as the discrete element method are the only realistic options for modelling and understanding what happens between the rollers.

The framework needed in order to describe the components of the HPGR system can be decomposed in two main categories; modelling the machine, and modelling of the material. When it comes to the machine model there are no major difficulties with either implementation of advanced CAD models or applying kinematics, other than the expertise and skill needed in order to prepare such datasets. The challenge lies in the modelling of the rock material as well as modelling the dynamics of the floating roller.

In this work the HPGR has been modelled in terms of particle flow and roller pressure distribution (Paper D), analytically modelled in terms of dynamics (Paper E) and with dynamics implemented (Paper J). The overall modelling objective is to be able to simulate particle flow and breakage including the dynamics of the floating roller. All DEM model components have been developed consecutively in order to achieve this.

8.1 Roller Pressure Distribution

The roller pressure distribution has been the subject of scientific discussion within the research community since the HPGR crushers were implemented in the mid-1980s. Due to the operational complexity and high pressure levels the pressure distribution is difficult both to determine analytically and measure experimentally. In Paper D the pressure distribution on the roller surface is investigated using DEM simulations. The effects of feed size distribution, roller speed and the use of cheek-plates are studied specifically.

In an HPGR machine the feed material is presented from above, guided by feeding plates to be compressed between the rotating rollers. The compression process in an HPGR resembles the continuous rolling of metal sheets or bars but instead a granular rock material is continually
subjected to pressure. Johanson (1965) has made important contributions to the theory of rolling compaction of granular materials. The theory was originally related to the compaction of powders between fixed rolls. The insights from these theories are a good starting point for analysing the HPGR. However, in an HPGR the material is coarser and, as previously mentioned, instead of both rolls having fixed positions one of them is pre-loaded with a hydraulic piston-accumulator spring system and allowed to move. The possibility of applying a force on the roller enables a higher force and pressure on the material in the compressive zone compared to a fixed roller press.

In Figure 46 the different zones of the compression process can be seen. The first zone, where the material first comes into contact with the rollers, is called the entry zone. There is none or a very low amount of breakage in this zone. The material in the entry zone experiences a pressure from the material resting above. It is important to run the machine at choke fed conditions in order to keep the feed pressure at a sufficient level. In the lower region of the entry zone the particles rearrange and a force network is established. During this process the material is subjected to shear stresses and slip behaviour between the roller and the particles. The pressure is successively built up in the compaction zone until particles begin to break and the packing density increases. The air filled void spaces between particles are reduced as the volume available is limited further down. The maximum compaction and pressure level is reached just before the neutral line between the roller centres. In the exit zone below the neutral line, the material is relaxed and the packing density is slightly reduced.

The shear stresses and material slip have been investigated and measured by Schönert and Sander (2002). In the paper the authors claim that while the normal pressure is possible to calculate based on piston and die experimental data, the tangential shear component is difficult to establish. Based on the discrepancy found between their measurements and calculations, Schönert and Sanders proposes that a reasonable explanation is that there exists one region with slip and another without slip. This hypothesis is strengthened by the performed DEM simulations in this work. In Figure 47 the velocity of each particle is displayed as they are being compressed between
the rollers. It can be seen that there is a uniform velocity at the $\alpha = 0$ position. The velocity is uniform in the region noted as the compression zone. Above this zone, in the entry zone, the particles are accelerated hence a velocity gradient can be observed. The gradient begins from the moment particles begin interacting with the roller surfaces. An interesting phenomenon regarding the velocity gradient is that the increase in velocity is first observed on the edges of the rollers. In the mid-section the particles are still at a lower velocity (blue colours). It is only when the particle contact interaction force network becomes rigid enough to lock the particles in position that the particles move together at uniform velocity. This means that, according to the simulation results there exists an internal slip behaviour in the material. We can hence observe two slip mechanisms; the material slip against the rollers, and the internal slip in the particle bed. This indicates that there are internal non-uniform shear stresses in the region above the compaction zone.

![Figure 47. Image from DEM simulation of HPGR displaying the velocity of particles during compression](image)

The pressure distribution in the compaction zone is a vital parameter when modelling HPGR machines analytically. Several researchers propose a parabolic shape of the distribution along the direction of the roller axis (Hasanzadeh & Farzanegan, 2011; Nadolski, 2012; Torres & Casali, 2009). The reason for the non-uniform shape is that the confinement of the particle bed is affected at the edges of the rollers and due to symmetry has its maximum in the middle. Most HPGR machines are equipped with so called cheek plates. These plates restrict the material from falling out at the end of the rollers. If there are no cheek-plates installed the pressure approaches zero at the edge. If cheek plates are installed the pressure is still reduced at the edges, but to a lower extent. The reduced pressure at the edges and hence possibility to achieve breakage is commonly known as the edge effect. Lubjuhn provided measurements of the roller pressure at three different positions, from the centre to close to the edge, on a laboratory roller using custom manufactured load cells. The pressure levels was investigated without edge confinement, when using cheek plates and when using so called dam rings. The dam ring solution is based on attaching an
additional metal ring on the outside of one of the rollers. The ring rotates together with the roller and covers the edge region. In this solution the confinement moves in the same direction as the material flow. Lubjuhn showed in his study that the use of dam rings greatly improved the pressure levels close to the edges, limiting the edge effect, suggesting that dam rings are superior to static cheek plates.

In Figure 48 the compressive force on the particles can be seen when operating with cheek plates (left) and without cheek plates (right). It is very clear that the use of these plates has a significant effect on the pressure distribution. Apart from reducing the amount of breakage in the edge region, a substantial amount of material flows over the edge unbroken. Therefore, if measuring the throughput capacity of an HPGR operating without cheek-plates the capacity will be deceivingly higher than operating with cheek-plates. However, as reported in Paper D, if only considering the material actually passing between the rollers the throughput capacity is higher when using cheek-plates than without. Some HPGRs are equipped with edge splitters that separate the material passing through the edge region from the rest of the flow.

Figure 48. The images show the compressive force on particles under compressive load between the two rollers. The left image shows the compression with cheek plates on the side. The right image shows the compression without cheek plates on the side.

The simulated pressure distributions on the roller surface for two different cases are presented in Figure 49. The upper plot shows the pressure distribution when the cheek plates are in place, restricting material from flowing at the edge of the rollers. As can be seen in plot b) the pressure is significantly lower as material is allowed to flow across the edge.

It can be seen that the edge effect appears both when operating with cheek-plates and without however the effect is more dramatic in the latter case. A wider size distribution generally increases the pressure distribution. The reason can be attributed to a higher initial packing density. The performance effect of the feed size distribution was recently investigated by Ozcan, Aydogan and Benzer (2015) by adjusting the recirculating load in order to control the amount of fines in the feed. A general conclusion is that the feed size distribution is a vital operational parameter and can be controlled by the recycle ratio in order to achieve optimum breakage efficiency. The DEM simulation results support these findings in terms of displaying a significant relationship between the feed size distribution and the pressure distribution.
Figure 49. Bar plots of the pressure distribution on the roller surface for the six simulated cases in Paper D. a) with cheek plates, narrow feed PSD, 120 deg/s, b) without cheek-plates, wide feed PSD, 120 deg/s
8.2 Modelling Dynamics

In order to include the dynamics of the floating roller a force-feedback model between the roller and the material bed is needed. In paper E the theoretical framework for this was developed. The conceptual model of the system can be seen in Figure 50. The left part of the schematic illustration shows the geometrical details and the forces due to the interaction with the material. In the DEM simulation all normal and tangential forces acting on the rollers are integrated for each time-step as input to a force balance equation. This integration is performed as an internal function in the EDEM multi-body dynamics coupling interface. The hydraulic piston and accumulator system is modelled using a spring-dashpot model as seen on the right hand side of the illustration. For each time-step the interaction between the floating roller and the compressed material bed will be balanced and the horizontal position of the floating roller updated accordingly.


debug

![Figure 50. Forces and dynamics of a high pressure grinding roll](image)

The throughput can be described as the slab of material with a specific packing density passing through the rollers per unit time according to Austin (1980),

$$m = (1 - \varepsilon) \rho s g b R \omega $$  \hfill (37)

Where $\rho$ is the solid density, $(1 - \varepsilon)$ is the volume fraction, $s_g$ the working gap and $b$ the roller width. The throughput equation works relatively well for estimating capacity in real operations when data is available for the flake density and working gap. However, as the working gap and flake density are output parameters it is not a good predictive model. A mechanistic model based on input parameters such as material characteristics, start pressure and machine design would be preferable.

The shear stress on the roller, see Figure 51, is not straight forward to define since the slip condition changes along the arc of the roller. However, an approximation can be made by assuming non-slip in the compression zone, enabling the definition of shear as the roller pressure multiplied by the coefficient of friction:

$$\tau(\varphi) = \mu p(\varphi)$$ \hfill \(38\)
It should be noted that it is possible to extract both the normal pressure as well as the shear stresses on each arbitrarily sized roller segment from the DEM simulation. This enables the possibility to evaluate the approximation used in Eq. (38).

In Figure 51 the force balance on the rollers and a material element is presented. The torque can be calculated as the roller radius multiplied by the integral of tangential forces acting on the roller surface due to shear stresses:

$$ T = R \int_{\alpha_{\omega}}^{\alpha} \tau(\varphi) b R d\varphi = R \int_{\alpha_{\omega}}^{\alpha} \mu p(\varphi) b R d\varphi = \mu b R^2 \int_{\alpha_{\omega}}^{\alpha} p(\varphi) d\varphi $$

(39)

Figure 51. Force balance between a feed material segment and the roller surface.

The pressure response profile on the roller can be described in cylindrical coordinates according to Figure 52. The shape of the profile is dependent on several parameters and hence complex. However some measurements have been made (Lubjuhn, 1994; Schönert & Sander, 2002) that indicate that the maximum pressure is located slightly above the neutral line. The results from paper D also support the principle shape of the distribution as presented in Figure 52.

Figure 52. Schematic illustration of the pressure response cross-section on the roller surface

The total grinding force is derived as the projected pressure multiplied with the area defined as the projected arc of contact multiplied by the roller width,

$$ f(t) = b R \sin(\alpha) \int_{\alpha_{\omega}}^{\alpha} p(\varphi,t) \cos(\varphi) d\varphi $$

(40)

The force balance for the spring-dashpot system seen in Figure 50 is given by Newton’s second law according to Eq. (41).

$$ m \ddot{x} = -c \dot{x} - k x + f(t) $$

(41)
Dividing by the roller mass $m$,

$$\ddot{x} + \frac{c}{m} \dot{x} + \frac{k}{m} x = f(t) \quad (42)$$

Let $\omega_0 = \sqrt{k/m}$, be the natural frequency of the system and $\zeta = c/2\sqrt{km}$, the damping ratio. We get the following second order differential equation,

$$\ddot{x} + 2\zeta \dot{x} + \omega_0^2 x = f(t) \quad (43)$$

In the Laplace domain, we have

$$X(s) \left[ s^2 + 2\zeta s + \omega_0^2 \right] = F(s) \quad (44)$$

Which gives the transfer function,

$$G(s) = \frac{X(s)}{F(s)} = \left( \frac{1}{s^2 + 2\zeta s + \omega_0^2} \right) \quad (45)$$

The differential equation in Eq. (41) can also be modelled using block-diagrams in e.g. MATLAB Simulink as seen in Figure 53. For each time-step the accumulated forces on the floating roller are calculated according to Eq. (40) and communicated via the EDEM dynamics coupling to either MATLAB Simulink or MSC Adams where the updated roller position is calculated and sent back to EDEM.

![Figure 53. Block diagram representation of the differential Equation (41).](image)

In this work no external software has been used in the coupling. The EDEM development team supplied a valuable tool in the API where external .dll code could be implemented using the dynamics coupling server. This template based solution for a linear spring dashpot is easily adaptable using the EDEM API C++ solution. The template dynamics coupling for the spring-dashpot used here was developed in a collaboration between EDEM engineers and G. Barrios at UFRJ, Brazil (2014; 2016). The solution has also been implemented by Kumar (2014) for modelling the HPGR.
8.3 Investigation of Dynamics, Shear and Pressure Profile

Even the most advanced mechanistic models will at some modelling level include assumptions related to physical effects and boundary conditions in order to solve the differential equations. One of these boundary conditions concerns the interaction between the particles and roller surface where it is commonly assumed that there is a slip behaviour in the entry zone and non-slip in the compaction zone. The pressure profile on the roller surface has been successfully measured by implementation of a strain gauge load cell on the roller (Lubjuhn, 1994). A new load cell design was later developed by Schönert and Sander with the ability to, apart from the pressure, also measure the shear stresses on the rollers (Schönert & Sander, 2002), see Figure 54. In the work by Schönert and Sanders the normal and tangential forces were measured and compared with calculations based on first principle equations and a set of assumptions needed in order to solve the differential equations. The results displayed a good congruence between the measured and calculated surface normal pressure. However, the surface tangential shear component showed a significant discrepancy between measured and calculated values. Some elaboration on the potential reasons behind the differences were provided in the paper and any significant measurement errors have been ruled out by the authors. This leaves the modelling assumptions regarding friction and shear slip behavioural conditions as the most probable source of inconsistency.

![Figure 54. Re-digitalization measurement plot of Fig 3 in Schönert and Sanders work. Positive shear direction is defined as the mass flow direction (downwards).](image-url)

In the work by Sander (1998) the dynamic behaviour of the moving roller was investigated. The laboratory mill setup used was equipped with a set of three different disc springs. The results showed dramatic fluctuations for certain operating conditions and speeds. Since only springs were used in order to apply the force a limited damping was introduced to the system. In the hydraulic system, in contrast to a mechanical spring system, there is natural damping due to the...
fluid flow conditions through pipes, valves and gas accumulators. Hydraulic flow restrictors may also be included to control the system damping properties more carefully.

In Paper J the force feedback dynamic model for the floating roller is implemented using a multi-body dynamics coupling to EDEM mentioned in the previous section. The scope of the paper is to evaluate the pressure and shear distribution and qualitatively compare to the measurements done by Schönert and Sander.

In the initial phase a range of different stiffness and damping coefficients was tested in order to evaluate the response and verify a realistic movement for the roller. The roller dimensions were 300 mm in diameter and 50 mm in length. The edge confinement was modelled in order to treat the modelled section as a sub-section hence the boundary wall friction was set to 0.01. Three multisphere template shapes were used each with 10 sub-spheres. The HMNS contact model was used for the contact force calculation.

In Figure 55 the roller position relative to the start gap can be seen for five different stiffness settings. A higher stiffness leads to a smaller gap. However, as can also be seen the simulation has not been run long enough in order to reach a steady state. The dotted line representing the highest stiffness also displays a more erratic curvature. This can be seen even more clearly when investigating the roller velocity for the corresponding time-series, see Figure 56. The dotted line representing the highest stiffness shows higher velocity amplitudes. The relationship between the stiffness and the gap can also be seen by studying the mass flow. In Figure 57 the average mass flow is presented for each stiffness level simulated. As anticipated the mass flow decreases when the stiffness is increased.

Figure 55. Floating roller position during the duration of simulation
In Figure 56 the velocity pattern for one of the simulations can be seen. The colour gradients indicate the entry, compression and relaxation zones. The particle velocities at the roller surface were investigated and compared with the circumferential speed of the roller. This allows identification of where the slip occurs.

In Figure 57 a new simulation not included in Paper J is presented. In this simulation a batch of bonded meta-particles has been included as an embedded batch in-between the normal multispher particles. The colour on the geometry represents the vertical force component where the red colour represents the positive z-direction (upwards). As can be seen the colour turns from red to blue just above the mid-plane meaning the tangential force changes sign. When comparing
this result with the Schönert and Sander measurements presented as a re-digitalization in Figure 54, it can be seen that the shear should change sign about 1 degree before the 0 degree mid-line. This position is close to the peak pressure angular position. These results indicate that the DEM model currently provides a qualitative correspondence.

Unfortunately the stiffness coefficient values have been found to be estimated incorrectly leading to too low pressure levels. This can be seen both when calculating the roller pressure distribution as well as due to that the meta-particles in Figure 59 are not fracturing due to too low forces.

![Figure 58](image_url)

Figure 58. Snapshot image from EDEM displaying the flow pattern with colour representing the velocity magnitude.

![Figure 59](image_url)

Figure 59. Snapshot image showing the bonded meta-particles embedded among the multisphere particles that have been hidden in order to provide visibility of breakage.
9 Discussion and Conclusions

The research results presented in this thesis aim at facilitating a simulation platform for modelling the compressive breakage in cone crushers and HPGRs including a general methodology for model calibration and validation. In every research activity performed in the project a sub-set of the stated research questions is targeted at different levels of detail and completeness. It is natural that new problems and challenges arise in the process for each area targeted. Hence, the iterative nature of the problem oriented research approach, see Figure 4, has been important. This learning process leads to an increased insight, understanding and experience for the researcher, and hopefully at least a subset of this knowledge is also new to the scientific community.

In every DE modelling and simulation project there are seemingly small problems that need to be resolved in order to arrive at a satisfying simulation result. Usually 10-50 iterations of a simulation model are required before an acceptable outcome is achieved. This iterative process is in many cases more related to verification rather than validation activities. In other words making sure that the model is developed according to the intended theory and functionality. This notion is equally true for the modelling in the DEM environment as well as for the data post-processing.

This aspect of an iterative work process has not been reflected upon in the papers appended in this thesis, but it is vital for DEM model developers to understand that for complex DEM model design, numerous minor problems on the detail level need to be identified and solved.

The proposed methodology for validation and calibration outlined in §6 should in principle be applicable in a general DE modelling perspective independent of the application specific examples presented for DEM flow and breakage calibration. The modelling insights regarding the cone crusher and HPGR are more specific. However, if conceptualizing the functions of each model component as a modular system, this modular platform has the capability to model most types of compressive breakage machines. The major limiting aspect is the scale of the machine potentially modelled, which in turn is limited by the computational power available. Hence, the capacity of DEM simulations is dependent on the evolution of computers and computational power. Even though claims have been made that Moor’s law is now dead (Simonite, 2016a, 2016b) it is likely that computational power will continue to increase even though transistors are not made any smaller. During recent years GPU computing has become more frequently used for particle based simulation. However, most implementations have to date only handled particle flow problems without breakage models. When the hurdles of including breakage models have been overcome, GPU or CPU-GPU computing will most likely allow for a step change in machine DE modelling scale feasibility.
9.1 General Conclusions

From the studies presented on the cone crusher and HPGR machines, some general conclusions can be formulated:

- Compressive crushing of rock material has been successfully modelled using the BPM model
- The BPM modelling approach is more computationally extensive than the PBRM, however the BPM provides more information regarding the damage history and does not have a predefined progeny size distribution.
- A framework for calibration of contact models has been presented based on design of experiments, surrogate response modelling and error optimization.
- A framework for using DEM in a problem solving context has been proposed and evaluated on materials handling problems
- It has been demonstrated that DEM can be used to validate previously analytically derived principles of operation and hypotheses regarding internal interaction between rock material and the machine
- The cone crusher has been modelled and validated against industrial as well as laboratory experiments with acceptable qualitative congruence and a reasonably good quantitative conformation.
- A laboratory scale HPGR has been modelled and compared to laboratory experiments presented by Lubjuhn (1994) and Schönert and Sanders (2002).

9.2 Answering Research Questions

Each of the research questions formulated represents a research or knowledge gap that was identified early on in the project. The gaps can also be defined as the limiting factors for achieving the goals of the project. In the project, all research questions have been targeted and new insights have been elicited so that further, more detailed questions, may be stated for future work. The relationship and dependency between each research question and the appended papers can be seen in Table 3.
Table 3. The dependency matrix displays the relationship between research questions and papers, with the size of the circle (three levels) corresponding to a subjectively judged level of focus.

<table>
<thead>
<tr>
<th>Research Question</th>
<th>Papers</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>A</td>
</tr>
<tr>
<td>RQ.1 Operational understanding</td>
<td>●</td>
</tr>
<tr>
<td>RQ.2 Design process methodology</td>
<td>●</td>
</tr>
<tr>
<td>RQ.3 Rock breakage modelling</td>
<td>●</td>
</tr>
<tr>
<td>RQ.4 Product particle size</td>
<td>●</td>
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<tr>
<td>RQ.5 Flow calibration</td>
<td>●</td>
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<tr>
<td>RQ.6 Breakage calibration</td>
<td>●</td>
</tr>
<tr>
<td>RQ.7 Machine geometry modelling</td>
<td>●</td>
</tr>
<tr>
<td>RQ.8 Machine dynamics</td>
<td>●</td>
</tr>
<tr>
<td>RQ.9 Power draw and pressure</td>
<td>●</td>
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<tr>
<td>RQ.10 Simulation validation</td>
<td>●</td>
</tr>
</tbody>
</table>

9.3 Scientific Contribution

The results presented can either be attributed to strengthening previous findings by others, contradicting findings by others, or being novel findings and models. In the section below a list of detailed findings is presented for each research question respectively.

RQ.1 How should DEM be used to understand and diagnose operating equipment in comminution processes?

- A new framework for the work flow of problem solving and concept development using DEM has been proposed.
- The framework has successfully been demonstrated for a cone crusher feeding materials handling problem
- The framework has successfully been demonstrated for laboratory cone crusher liner design development

RQ.2 How should DEM be used as a design evaluation tool in a product development context?

- The use of DEM as a concept evaluation tool has been demonstrated using the proposed DEM design process.
- For the first simulation in step i, a high level of fidelity should be applied in terms of geometry and particle material model
- Lower fidelity simulation models, balanced against the predictive accuracy, may be used in order to quickly evaluate the principle of different conceptual solutions
RQ.3  *How should rock breakage in compressive breakage machines be modelled in order to balance validity and computational economy?*

- A new bonded particle model utilizing a bimodal size distribution for fraction particles has been developed. It was found that a bimodal distribution offers the best packing behaviour and minimum number of particles in each meta-particle while maintaining sufficient quality of the breakage behaviour. The bimodal distribution affects bond beam strength distribution enabling a heterogeneous strength distribution within the body of the meta-particle.
- A novel method for creating realistic irregular rock particle shape of meta-particles has been developed based on 3D-scanned rock particles.

RQ.4  *What is a suitable method for calculating the particle size distribution of bonded particle model?*

- A new method, based on image processing, for calculating the product size distribution of progeny meta-particle clusters has been developed and compared with industrial scale data.

RQ.5  *How should DEM material models be calibrated with regards to particle flow?*

- A new calibration and validation framework has been proposed and demonstrated. The framework is based on optimization of the pragmatic congruence tolerance between the experimental and the simulation based surrogate model.
- The concept of multi objective optimization has been applied to the calibration process of contact model parameters for multiple flow regimes.
- A new device for calibration of DEM contact model parameters has been designed and demonstrated.
- It has been demonstrated that contact model parameter calibration needs to be conducted using multiple flow regimes.
- A new method for capturing real single particle flow within a bulk flow using marker particles and motion capture technique has been developed.
- A new method for calculating the repose angle from simulated pile formation by fitting a sigmoid function to the free surface layer particle positions has been developed.
- It has been demonstrated that the dynamic repose angle is dependent on the flow velocity of the bulk material prior to bed formation.

RQ.6  *How should DEM material models be calibrated with regards to particle breakage?*

- A calibration procedure for BPM models using Brazilian tensile strength test has been developed.
- A framework for breakage model calibration has been proposed based on surrogate-based optimization.
- It has been demonstrated that the model accuracy has a trade-off pareto relationship to the computational load.
RQ.7 *How should machine geometry be modelled in DEM?*

- 3D-scanning point cloud data has been used for modelling primary gyratory crushers. This is the first time reported in literature that point clouds are used for DEM simulations of crushers.
- 3D scanning point cloud data has been used for modelling machine geometry of industrial scale cone crusher liners in order to match the actual worn state of the machine.

RQ.8 *How should machine dynamics be integrated with DEM?*

- The framework for the dynamic model of the HPGR has been developed and implemented.
- The dynamics of the HPGR rollers have been analysed for different stiffness and damping characteristics.

RQ.9 *How should the estimation of forces, pressure and power draw be conducted?*

- The power draw and pressure have been calculated and compared to an industrial scale cone crusher.
- A method for calculating the pressure distribution on the roller surface based on DEM simulation data has been developed.

RQ.10 *How should DEM simulations be validated?*

- The V-model for DEM calibration and validation has been proposed.
- So called *face validity* has been demonstrated when comparing DEM simulation results with findings from analytical modelling (C.M. Evertsson, 2000) (using first principle mechanics) regarding choke level positions, breakage mode and particle squeezing behaviour in the crushing chamber.
- A laboratory scale validation case of a cone crusher operated at high eccentric speed has been demonstrated.
- An industrial scale validation case of a secondary stage cone crusher has been demonstrated.

### 9.4 Vision for further development

The utopia of discrete element breakage modelling would be to simulate a machine in full operation with a material model in which the real rock grain size distribution and texture is represented. Let’s exemplify by considering a small machine where 1 litre of material is subjected to a mechanical breakage environment. If the mean grain size of a relatively coarse rock texture would be 1 mm and we approximate the shape of each grain by a cube. Then the corresponding number of particles or voxels needed for one litre of rock would be 1 million. The current state of the art DEM codes are capable of simulating up to 5-15 million particles. However, the benchmark simulations then usually referred to are flow simulations and do not consider the computational
load increase due to modelling of any bonding interactions. It is hence currently not a feasible approach forward to model 10-100 litres of material in full resolution at once.

What can be done however, is to model each meta-particle rock with a larger grain size distribution than the actual, which is what is being done in this work. Through the calibration process the bonds between the fraction particles may still be defined in order to achieve a useful breakage response. While useful for evaluating machine design, it may not be useful for rock mechanics research purposes.

A novel alternative solution (future work) to the PBRM and BPM implementation would be to combine them. Before proceeding to explain this solution, a brief repetition on how the BPM is currently created may be needed. In the current BPM implementation the fraction particles are included in the simulation at a pre-defined time. At this time step a set of dummy particles is replaced by the cluster of fraction particles forming the meta-particle. This replacement action is in fact very similar to the breakage replacement algorithm implemented for the PBRM model. The difference is that in the BPM case the replacement is triggered at a pre-defined time, and in the PBRM case it is triggered by a loading condition.

Instead of triggering the replacement action either by load or time, the replacement could take place based on position. Hence, instead of creating all BPM meta-particles in a batch, they would stay as inexpensive non-breakage particles, until they enter a ‘risk of damage domain’. The damage domain would correspond to the crushing chamber in a cone crusher and the compression zone in the HPGR.

This model has not yet been implemented in this work but would require the following changes:

- The replacement to be initiated based on a spatial domain constraint
- Improve the replacement function to allow for the use of multi-spheres as dummy particles
- Implement orientation compliance in the replacement function in order to align the multisphere domain and the meta-particle.
- The BPM model needs to be re-formulated so that the bonding action is activated at each continuous replacement event. However, only the most recently fraction cluster particles are allowed to bond. Otherwise there is a risk of bonding chunks of meta-particles together or re-bonding already broken clusters.

This approach would allow for a high resolution BPM or TEM modelling approach by taking advantage of the engineers/modellers knowledge regarding where the breakage will take place.

Finally, the results presented in this thesis demonstrate that fundamental novel understanding regarding comminution equipment can be achieved using DEM. For research and development departments operating in the mining and aggregate industry that seriously begin using particle based simulation methods, there will be no way back.
10 References


