

Simulating operational improvements on mine compressed air systems

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Abstract

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As operational costs of deep level mines increase, and gold ore grades decrease, profitability in the South African gold mining sector is becoming a challenge. Electricity tariff increases have contributed to this rise in the cost of operating mines.

Compressed air supply systems are the largest energy users in a mine, contributing to approximately 20% of their total power usage. It has been shown that mining compressed air networks are systemically inefficient. Improving the efficiency of these systems would result in a reduction of energy costs.

Previous studies have shown the usefulness of simulations to develop improvements for deep level mining systems. However, these studies have not followed a structured methodology for developing compressed air simulations. Previous studies also used simplified compressed air models, therefore, reducing the simulation precision and testable scenarios.

In this study, a simulation methodology was developed. Investigations into the compressed air systems are also performed. A compressed air system is then modelled in adequate software to recreate the system operation accurately. Finally, a proposed means of improvement is simulated, analysed and quantified regarding improvements in energy savings and service delivery.

Two case studies were evaluated. For each case study, a variety of scenarios were simulated. In Case study 1, two scenarios were simulated on a compressed air system. The results showed that energy cost savings of R0.91M could be achieved. The simulation results were very similar to tests later performed on the physical systems.

The results of case study 2 showed that by reducing air usage at refuge bays, an average power reduction of 1 MW could be achieved. The improvement in efficiency would potentially lead to R 5.2m in annual energy cost savings. Additionally, a significant improvement of 15 kPa to system pressure during the drilling period were identified. Other scenarios showed annual energy cost savings of up to R 2.5m.

An additional analysis was implemented to assess the use of periodically repeated simulations. The results demonstrated that through repeated simulations, operational changes in a system could continuously be identified. This information can then be used for further improvement and cost savings.

The study showed that a simulation is a valuable tool for the identification of improvements in compressed air systems. By utilising a structured methodology to develop detailed compressed air simulations, inefficiencies and operational improvements were successfully identified.

Keywords: Simulation, compressed air, energy, mining, operational improvements

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Acronyms

CALDS	Compressed Air Leakage Documentation System
COLS	Corrected Ordinary Least Square
DCS	Dynamic Compressor Selection
DSM	Demand Side Management
E.E.	Energy Efficiency
MAE	Mean Absolute Error
MRD	Mean Residual Difference
MSE	Mean Square Error
p.a	Per Annum
P.C.	Peak-Clip
PGM	Platinum Group Metal
PI	Proportional-Integral
PLC	Programmable Logic Controller
PTB	Process Toolbox
RMSE	Root Mean Square Error
SCADA	Supervisory Control and Data Acquisition
SI	International System of Units
SP	Set-Point
THS	Thermal-Hydraulic Simulation
TOU	Time Of Use
VFD	Variable Frequency Drive
VSD	Variable Speed Drive

Nomenclature

Celcius	The SI measure for tempature	C
Correlation coefficient	The statistical measure of linear relationship between two variables	r
kilopascals	The international measure of pressure	kPa
Mass flow	The measure of mass of a fluid moving per unit of time	kg/s
Polytropic coefficient	The thermodynamic coeifient used when describing the heat transfer due to compression or expansion	—
Rand	The South African monetary unit	R
Relative error	The absolute error divided by the magniture of the exact value.	$Err\%$
Stopes	Underground mining area	—
Tonne	The non-SI measure for 1000 kilograms	T
Volumetric flow	The measure of the volume of flow that moves per unit of time	m^3/s
Watt	The SI measure of power	W

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CHAPTER 1

Introduction and background

“Precisely one of the most gratifying results of intellectual evolution is the continuous opening up of new and greater prospects.” - Nikola Tesla

1.1 Preamble

This chapter firstly discusses background regarding deep level mining in South Africa. Next, the need to reduce costs of operation in the mining sector is examined. From this, a focus on reducing the energy consumption of compressed air systems is developed. Next, background on compressed air operation and energy interventions are discussed. Simulations and their value in the industry are discussed; leading to a problem statement and objectives of the study. Finally, an overview for the dissertation is provided.

1.2 Background on deep level mining

1.2.1 Mining profitability

Various technical, economic, social and operational challenges are posing a risk to the profitability of the South African mining sector. One of the challenges the sector faces is a rise in the cost of operation [1].

A considerable factor that is contributing to the increase in operational costs in South African gold mines has been the rise of electricity costs. As shown in Figure 1.1, the general cost of electricity has increased at a rate greater than inflation since 2008¹.

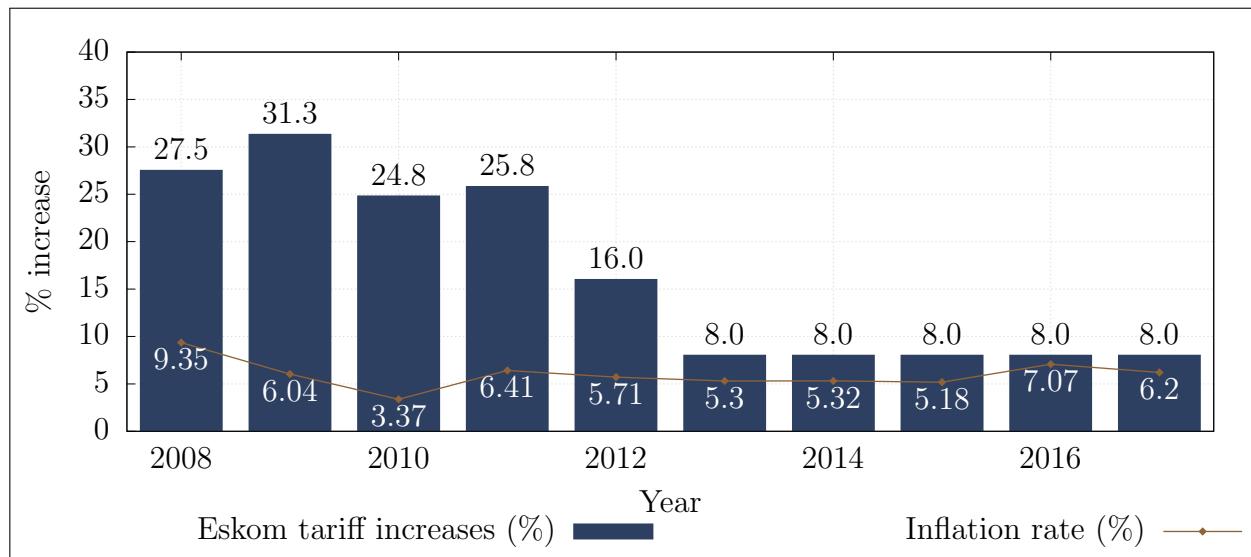


Figure 1.1: Electricity price increases between 2007 and 2017¹ compared to the inflation rate in South Africa²

¹ Eskom, "Revenue application - multi year price determination 2013/14 to 2017/18 (mypd3)," [Online] <http://www.eskom.co.za/CustomerCare/MYPD3/Documents/NersaReasonsforDecision.pdf>, 2013, [Accessed 22 March 2017].

² inflation.eu, "Historic inflation South Africa." [Online] <http://www.inflation.eu/inflation-rates/south-africa/historic-inflation/cpi-inflation-south-africa.aspx>, [Accessed 25 March 2017].

In addition to rising electricity costs, gold ore grades of South African mines have fallen substantially over the last few decades [2]. As ore grades decline, the energy utilised per unit of metal increases exponentially [3].

Additionally, mines are having to extend their operations deeper in order to reach the valuable ore reefs [CITATION NEEDED]. Therefore, mines require significantly more energy per unit of metal produced. This combination of tariff increases and increased energy usage per unit have led to significant rises in mining operation costs.

1.2.2 Process of a deep level mine

South Africa's mines are some of the deepest in the world. Some mine shafts are reaching depths greater than 4000m below the surface [4]. The process of extracting ore at this depth is dependent on the essential services, mainly cooling and ventilation, pumping, compressed air and hoisting, as shown in Figure 1.2.

Cooling and ventilation systems are required to maintain a safe working temperature underground. Pumping is critical to remove service and fissure water, preventing flooding. Compressed air is used to power underground drills and machines safely. A hoisting system is needed to bring the ore to the surface and to transport workers in the mine. The hoisted materials are transferred to a plant for processing.

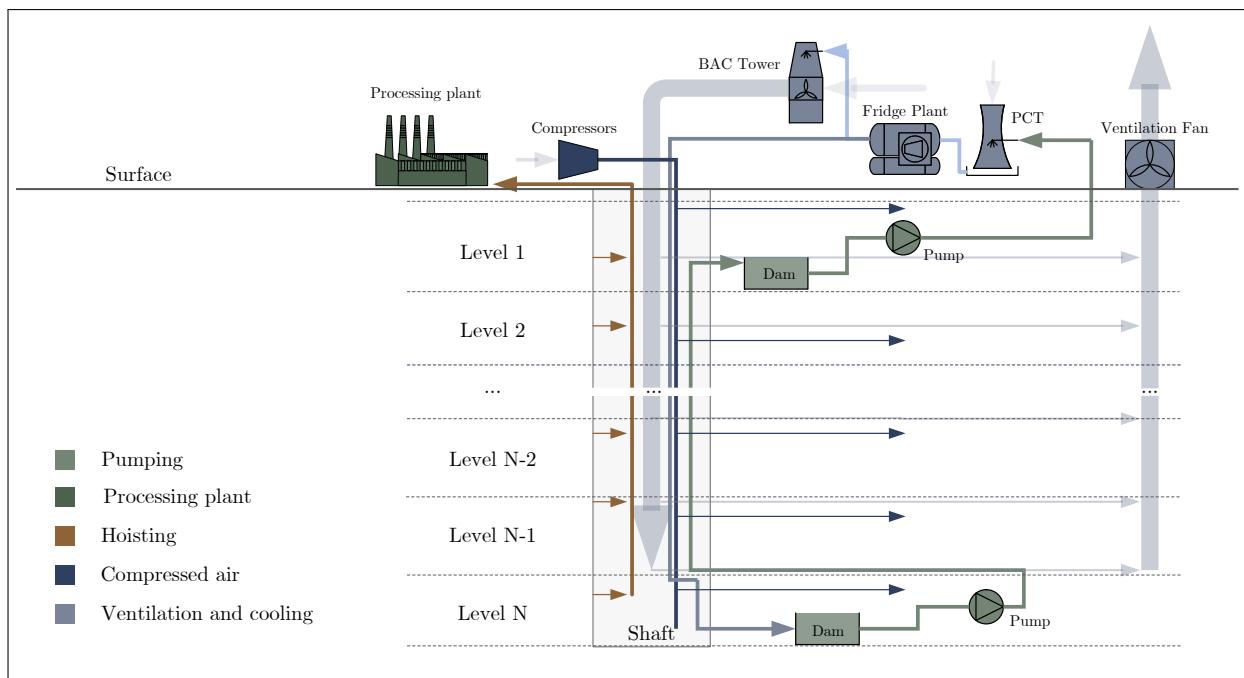


Figure 1.2: A schematic showing the mining processes.

1.2.3 Energy usage of mining services

The mining industry uses X MW of Energy [Citation needed]. In South Africa, the industry utilises approximately 15% of the national electricity supplier's annual output, of which, gold and platinum mines use 80%¹.

Figure 1.3 illustrates the division of energy use within the mining industry. The chart indicates that compressed air systems utilise the most energy within a mine. Therefore, energy can be most effectively reduced through the implementation of energy saving interventions on compressed air systems.

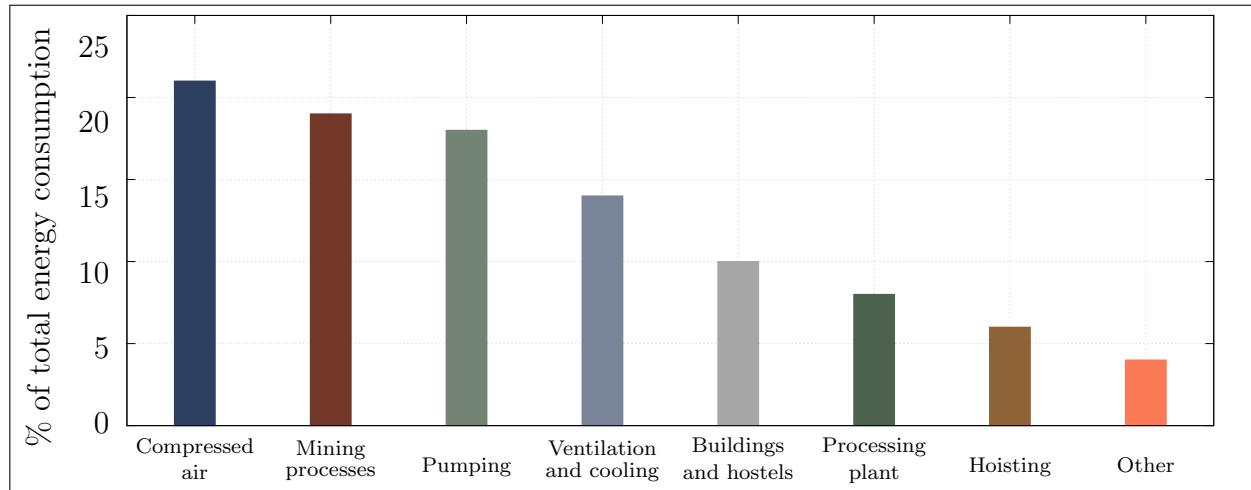


Figure 1.3: The percentage energy consumption for each mining system [5]

1.3 Compressed air systems in mining

1.3.1 Compressed air in operation

Largely due to their reliability, versatility and ease of use, the South African mining industry has installed extensive compressed air networks. These systems can have compressors with capacities of up to 15 Megawatt (MW)[6]. However, the supply of compressed air is a high energy demanding and costly process [7].

The energy used for compressed air production contributes to between 9% and 20% of the total mining energy consumption¹ [8]. Additionally, the compression process is highly inefficient. It has been estimated that the efficiency of the process of converting electrical energy to power pneumatic drills is as low as 4% [9].

¹ Eskom, "The energy efficiency series - towards an energy efficient mining sector," [Online] <http://www.eskom.co.za/sites/idm/Documents/121040ESKD>, February 2010, [Accessed 19 March 2017].

Large compressed air systems are likely inefficient. Internationally, the expected energy savings potential of a large compressed air network has been proven to be 15% [10]. Marais [11] showed that energy efficiency interventions could lead to energy and cost savings of between 30% and 40%.

1.3.2 Inefficiencies found in compressed air systems

Compressed air distribution networks in the mining industry consist of multiple compressors and working areas up to eight kilometres away from the source on the surface [6]. Due to their size and complexity, these systems are prone to significant energy losses [Citation Needed].

Compressed air leakage accounts for as much as 35% of the energy losses of a compressed air network [12]. Other systemic losses include, faulty valves, pipe diameter fluctuations, obstructed air compressor intake filters and inefficient compressors.

Leakage and inefficiency detection strategies are not often pursued in the South African mining industry [13]. Many mines do however perform leak inspections either internally or by an outside company. In these inspections, an ultrasonic detector is used to locate the leak.

Alternatively, some mines employ the “walk and listen” method to identify leaks from the audible sound that it produces [13]. Once the inspection is completed, the findings, including the locations and estimated costs of all identified leaks, are reported.

1.3.3 Compressed air savings interventions

From literature, it has been observed that energy saving interventions on compressed air systems have historically implemented one or a combination of following strategies [14]:

- Reduce leaks
- Reduce demand
- Reduce unauthorised air usage
- Increase supply efficiency
- Optimise supply

Often a combination of energy strategies will lead to the most savings [6]. In Chapter 2, successfully completed compressed air interventions on mining systems will be discussed from literature.

Once an energy saving measure has been identified, it is most often necessary to make

estimations to determine the potential costs and benefits of the intervention. The estimations have typically been performed using first principle calculations, simplified mathematical models and practical tests where possible.

However, new tools have enabled quick, accurate compressed air model development. Through simulations, accurate estimations can be obtained quickly, with no risk and at comparatively low resource requirement.

1.4 Use of simulation in industry

1.4.1 Background on industrial simulation

Continuous improvements in computing hardware have led to major advancement in software technology. Consequently, the use of computational simulation has become an increasingly valuable tool for many industries [15].

In *Handbook of Simulation: Principles, methodology, advances, applications, and practice*, the advantages of the use of simulation in the industry are discussed as follows [16]:

- The ability to test new policies, operating procedures and methods without disrupting the actual system.
- The means to identify problems in complex systems by gathering insight in the interactions within the system.
- The facility to compress or expand time to investigate phenomena thoroughly.
- The capability to determine the limits and constraints within a system.
- The potential to build consensus about proposed designs or modifications.

1.4.2 Simulation usage in compressed air optimisation

Simulation has been used to test and identify energy and operational improvement modifications in mining compressed air systems. However, in the past producing complex models for mining systems was not feasible as simulation software required difficult to obtain data inputs [11].

Simplified "vessel" model

Before new software tools allowed for the development of detailed mining compressed air simulation models, Marais [6], [11] created a simplified compressed air model to estimate and quantify the performance of potential energy interventions. Marais simplified the mining compressed air system, comparing the network to an air source and a vessel with many leaks.

The simplified model is illustrated in Figure 1.4.

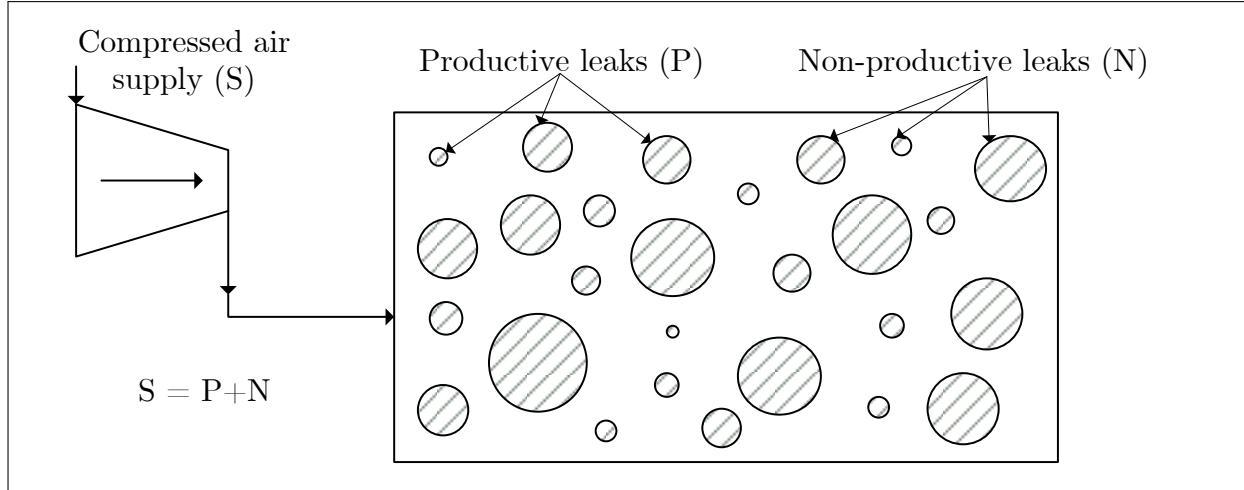


Figure 1.4: Simplified compressed air network model. Adapted from Marais [6].

A calculation methodology was developed to estimate the expected energy savings impact on the system quickly. From this, energy saving estimations and rules were designed as listed in Table 1.1.

Intervention	Estimation rule
Reducing compressor deliver pressure	$x\%$ pressure reduction $\propto (1.6 \text{ to } 1.8) \cdot x\%$ power reduction
Reduce control valve pressure	$x\%$ pressure reduction $\propto p \cdot x\%$ power reduction. Where p is the valves' relative flow contribution to the system
Reduction of flow	$x\%$ flow reduction $\propto x\%$ power reduction

Table 1.1: Summary of energy saving estimation rules [6].

There is not a high degree of precision in this approach as specific details regarding the air network are not taken into account. The simplified approach cannot be used to estimate more complex scenarios. The method also does not estimate other potential benefits of interventions such as pressure delivery improvements.

Simplified air network model

Kriel [17] used simulation to estimate the performance of energy projects on mine compressed air systems. The KYPipe GAS software tool was utilised to develop simulation models for

the systems. Kriel simplified the air networks for the simulations to a single compressor representing the supply processes and an outlet flow to each underground level in the network. The model is shown graphically in Figure 1.5

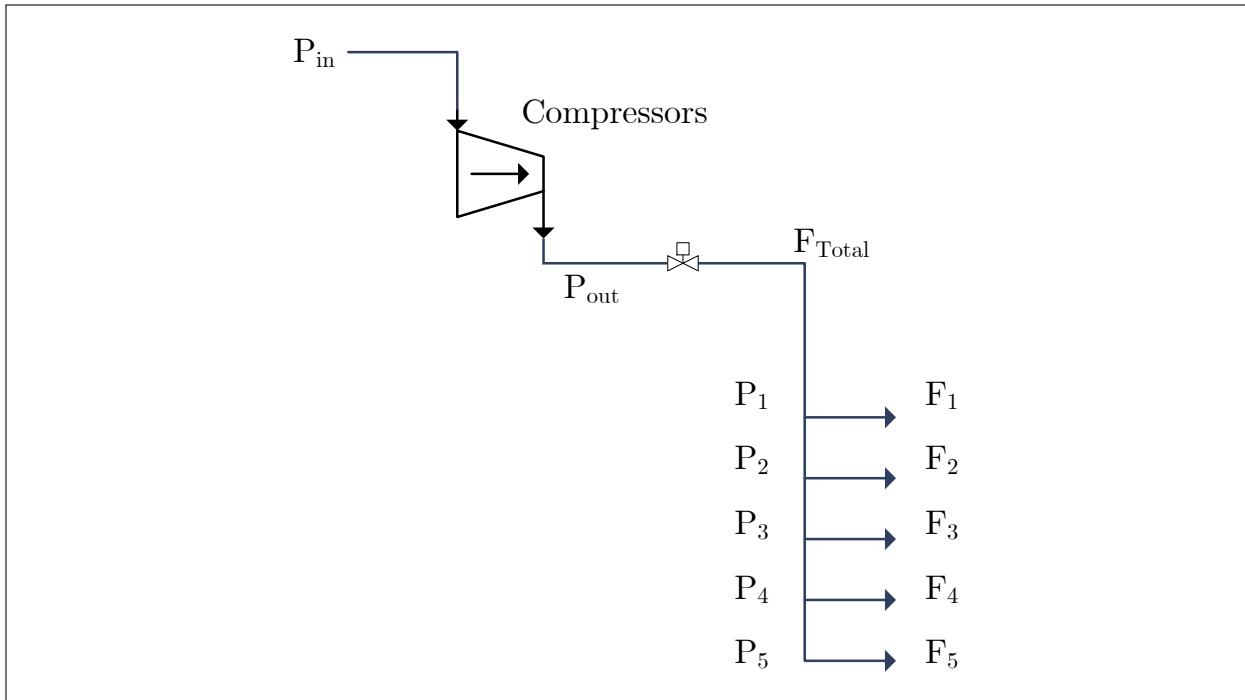


Figure 1.5: Simplified system model. Adapted from Kriel [17]

The simulation was performed to quantify the savings from underground network interventions. The interventions were designed to reduce flow to the network. The estimated savings from the simulations varied between 10 and 25% compared to the actual performance of the interventions. Like the vessel model discussed previously, the simplified air network model can not be used to estimate the energy savings potential of more complex scenarios.

The simulation procedure in this study could be improved by using a more detailed model and a more precise verification method. The method could lead to savings predictions with higher accuracy.

The use of planned manual measurements, estimations and new software technologies can be used to develop more detailed compressed air models [18], [19]. Using a structured procedure may allow for the development of more detailed and accurate mine compressed air simulations.

1.5 Problem statement and objectives

1.5.1 Problem statement

Rising costs and falling ore grades are driving the mining industry to reduce operational costs. The industry can save significant energy cost through interventions in compressed air systems. However, manual testing of these interventions is risky and cumbersome.

Computer modelling and simulation of compressed air systems can be used to quantify and prioritise operational interventions with minimal risk. However, integrated simulations have not been used to their full potential in compressed air system studies in the past. With new tools and more detailed simulation models, the energy and operational efficiency of mining compressed air systems can be improved, leading to significant energy and cost savings as well as other potential improvements for a mine's operation.

Therefore, a need exists for an integrated compressed air simulation approach to identify energy and operational improvements for mines.

1.5.2 Research objectives

The main aim of this dissertation is to identify energy savings through the identification of operational improvements in mining compressed air systems. A simulation process will be developed to achieve this goal. The other objectives for this study are:

- Develop an integrated approach to develop an implement compressed air simulations;
- Use simulations to apply and rank compressed air operation interventions;
- Model mining compressed air networks components accurately
- Develop an approach to verify compressed air simulation models

1.6 Dissertation overview

Chapter 1 - This chapter serves as an introduction to the dissertation. The chapter provides the relevant background in mining, compressed air and simulation to establish the problem statement. The objectives of the study are then outlined.

Chapter 2 - Chapter 2 provides a review of the literature and necessary background relevant to the study. The chapter firstly provides background on mining compressed air networks. A review of compressed air energy interventions is then performed. A review of simulation usage in the mining industry then follows. Finally, a review of simulation usage in mining compressed air systems.

Chapter 3 - Chapter 3 provides a simulation methodology. The method outlines the processes used to investigate a compressed air system, develop and calibrate simulation models and finally to implement simulations and obtain results.

Chapter 4 - Chapter 4 provides validation of the methodology through case study results. Results of three case studies are provided. Finally, the impact of large scale implementation of the simulation methodology is discussed.

Chapter 5 - Chapter 5 concludes the dissertation. The chapter also provides recommendations for further studies regarding compressed air simulation.

CHAPTER 2

Overview of simulation and compressed air applications

“If I have seen further, it is by standing on the shoulders of giants.” - Isaac Newton

2.1 Introduction

This chapter firstly provides relevant background on the operation and subcomponents of compressed air networks. A review of the literature regarding compressed air energy interventions was then be performed. This review summarised previous work to improve the supply and demand efficiency of mining compressed air networks.

The use of simulation in mining systems was then be reviewed. The section summarised the usage of mathematical estimation and simulations tools in the mining industry. From the literature, available simulation tools, as well as simulation and verification procedures, were discussed.

Finally, simulation usage in compressed air systems was discussed. The section summarised the successes and shortfalls in previous compressed air simulation studies.

2.2 Background on mining compressed air networks

2.2.1 Preamble

Compressed air is used extensively in a mine in surface and underground operations. This section provides background regarding mining compressed air networks. The section firstly discusses the components that make up a mining compressed air system. The typical functioning of the system is then examined. Finally, the instrumentation that is typically installed in compressed air networks is discussed.

2.2.2 Compressor air network components

Compressors

Compressed air in mining is most commonly supplied by a centrifugal-type dynamic compressor [20], [21]. These machines achieve compression as a result of the centrifugal force from the high-speed rotation of impellers in the air. An electric motor drives the rotating impeller.

Multi-stage impeller compressor designs, as shown in Figure 2.1, are used to obtain higher pressure ratios [20]. The compression process is inefficient. Only about 5% to 10% of the input energy of the process is converted into energy that is used [22].

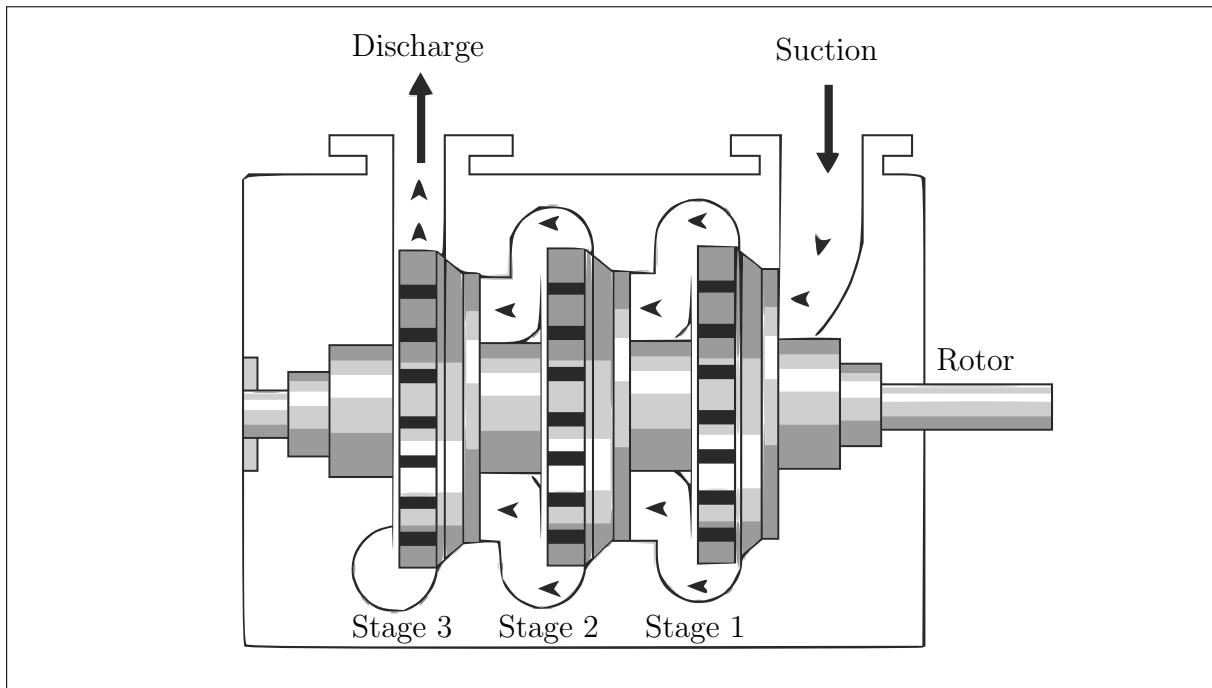


Figure 2.1: A schematic of a multi-stage centrifugal compressor¹

Pneumatic rock drills

Drilling is mainly performed in the production areas or stopes of a mine. Drill machines are used to drill holes into the rock face. Once the holes have been drilled, explosives are then installed to break up the rock [23].

Compressed air is used to power pneumatic rock drills within a mine. Pneumatic rock drills run at an efficiency of 2%. The efficiency is low when compared to alternative rock drills such as electric, oil electro-hydraulic and hydro-powered drills that convert energy at an efficiency of between 20-31% [9], [13].

Refuge bays

Refuge bays are installed underground in deep level mines to provide safety to miners in the event of an emergency. Due to safety regulations, most mines will utilise compressed air to deliver fresh air to the chamber [24]. Figure 2.2 shows an example of a compressed air inlet at an underground refuge bay. A muffler is installed to the end of the inlet air pipe to reduce noise.

The provision of 1.42 l/s of air per person at a pressure between 200 and 300 kilopascals is required to provide oxygen and prevent any poisonous gas entering the refuge [24].

¹ Abc oil refining, "How centrifugal compressors operate," [Online] <http://abcoilrefining.blogspot.co.za/2012/03/how-centrifugal-compressors-operate.html>, [Accessed 15 August 2017].



Figure 2.2: An example of compressed air inlet in an underground refuge bay chamber of a mine.

Airflow in the refuge bays is controlled by a manual valve within the room. The manual valves are often misused by mine workers to cool the bay through decompression of the air.

Processing plants

Processing plants are constructed near mines. They are used when extracting metal from the ore that is obtained from the mining operation. These plants use compressed air for various systems, processes and equipment.

Processing plants often share a compressed air network with mine to save costs [6]. The plants use relatively small amounts of air compared to mines. However, plant processes have pressure requirements that differ from the rest of the air network. If the plant air supply is not isolated from the shaft's air system, energy optimisation can be complicated [17].

Other compressed air users

Due to the availability underground, compressed air is utilised for several other applications. These usages include, pneumatic loaders or rock shovels, pneumatic cylinders, dam sediment agitation, cooling and ventilation and many other applications. This vast variety of applications also leads to misuse of compressed air. This leads to inefficient operation [6].

2.2.3 Operational schedule

On a typical mine, various operations are scheduled different times of the day. Depending on the activity taking place, many mines will control the pressure to meet the requirements of the tasks [6], [17]. Figure 2.3 shows the schedule and pressure requirement on a typical deep level mine.

As shown in Figure 2.3, the pressure requirement changes depending on the activity taking place. The drilling shift typically has the highest pressure requirement while blasting shift requires the lowest. This shifting pressure require shows scope for optimisation. However, it has been observed that many mines do not effectively optimise their compressed air supply [7]. Schedules and operation philosophies can differ between mines. Different operational schedules require alternative pressure requirement profiles.

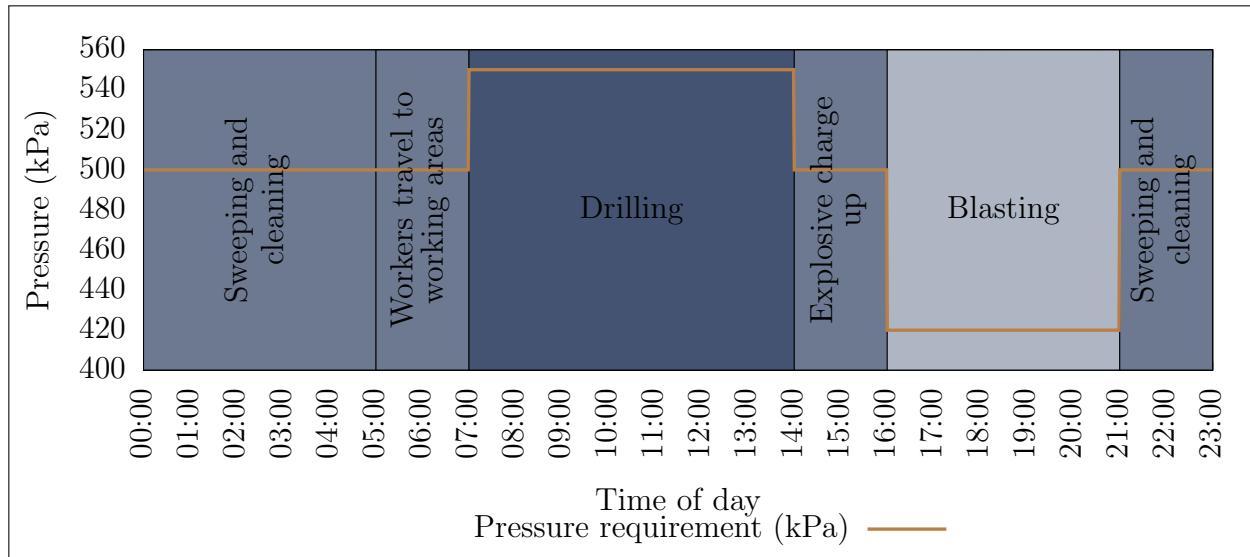


Figure 2.3: The typical operation schedule of a deep level mine [17]

2.2.4 Instrumentation

For large industrial systems, instrumentation is necessary to monitor performance and equipment conditions throughout the system. In a mining compressed air network, instrumentation is installed to monitor flows, pressures, temperatures and other process parameters. Electrical instrumentation is also installed for sensing currents, power factors, voltages and power. Parameters that are often measured include valves input/output pressures and flows, valve and guide vane positions and air condition metrics are usually measured with instrumentation.

Supervisory Control and Data Acquisitions (SCADAs) systems are used to monitor and control processes throughout the mine. The instrumentation and SCADA are connected on a communication network. This communication between the underground Programmable Logic Controllers (PLCs) and surface SCADA is achieved using a substantial fibre optic network[25]. The SCADA centralises instrumentation data from PLCs and instrumentation in the mine. The SCADA is also used to control machines and control instrumentation by transmitting control signals over the network.

2.2.5 Summary

In this section, various subcomponents of compressed air networks were discussed. These sub-components included compressors, processing plants as well as underground air users. The daily process scheduling and the effect on compressed air requirements were then reviewed. Finally, the typical instrumentation found in a mining compressed air system was summarised.

2.3 Review of compressed air energy interventions in industry

2.3.1 Preamble

Compressed air improvement can be obtained through intervention in either the supply or demand processes of a compressed air system [17]. Improvements in supply are achieved by increasing the efficiency of compressed air supply. Examples air supply optimisation measures include Dynamic Compressor Selection (DCS), compressor relocation, repairs and maintenance [14].

Due to the size of mining compressed air networks, there is often a larger scope for improvement in air demand. Improving the air usage is achieved by optimising air flow consumers, reducing leaks and through other interventions. Air demand optimisations include reducing unnecessary air use, optimising control and reducing system pressures usage [26].

This section will review compressed air supply and demand interventions that have improved energy or operational efficiency in the mining industry. From the literature, successes and shortcomings in studies will be discussed and analysed.

2.3.2 Strategies to improve compressed air supply

Optimising compressor control

The amount of available compressors as well as the compressor ratings can differ widely from mining compressed air systems. Compressor selection is crucial in these systems to match the correct compressors with the requirements of the system [27].

In a study, Booysen [21] found that many mines control compressors using fixed pressure points that are much higher than required. In one system, compressor controllers were set to target 650 kPa to ensure pressure underground did not fall below 500 kPa. Use of

high set-points can lead to excessive, wasteful blow-off air flow when the pressure exceeds maximum operating points.

Booysen [28], showed through dynamic pressure set-point control¹ and optimal compressor selection, energy savings can be achieved. In a case study, an average power demand reduction of 1.07 MW was achieved. The energy improvement lead to an estimated cost saving of R 3m.

Optimising control of compressors to match the demand of the system can be complicated. Variable Speed Drives (VSDs) and guide-vanes are used to control the capacity of the network. More effective power reductions can be achieved through the use of VSD control [29]. However, installing the controllers can be very costly.

Running compressors at part load will reduce the efficiency. From literature, it has been shown that electric motors will typical use 60-80% of their rated power when operating at less than 50% load [30].

Reconfiguring compressed air networks

Some old mining compressed air systems have not been adequately maintained and improved. Often they cannot sufficiently supply air to meet the demand or air is provided from non-optimal sources. In a study by Bredenkamp [29], reconfiguring of the air network was investigated to improve these systems.

In the study, Bredenkamp investigated interconnecting the compressed air systems of two mining shafts and relocating of a compressor. This strategy leads to an average power reduction of 1.7 MW and an estimated annual energy cost saving of R 8.9m at the time.

2.3.3 Strategies to reduce compressed air demand

Reducing the airflow demand is achieved by reducing unnecessary airflow such as leaks, optimising the operating pressure to match the demand² and by improving the efficiency of air usage or replacing equipment with non-pneumatic alternatives [14].

¹ Matching the supplied pressure with the specific required air demand

² Matching the supplied pressure with the required air demand. Optimising the supply is achieved by identifying the minimum required operating pressure for all times of the day as illustrated in Figure 2.3

Leakage detection

Air leaks are a major inefficiency in mining compressed air systems. Improving leaks is an effective measure to reduce air demand and improve the efficiency of the system [31]. However identifying leaks is difficult and time consuming.

Air leaks occur as a result of open pipes, fissures and breaks. Losses depend on the size of the leak and pressure in the network. Figure 2.4 shows the theoretical airflow through a pipe orifice as a function of leakage area and pressure¹. van Tonder [31] showed that the system power consumption linearly increases with the amount of air leakage. Therefore, energy savings can be achieved through either reducing pressure or detecting and fixing leaks.

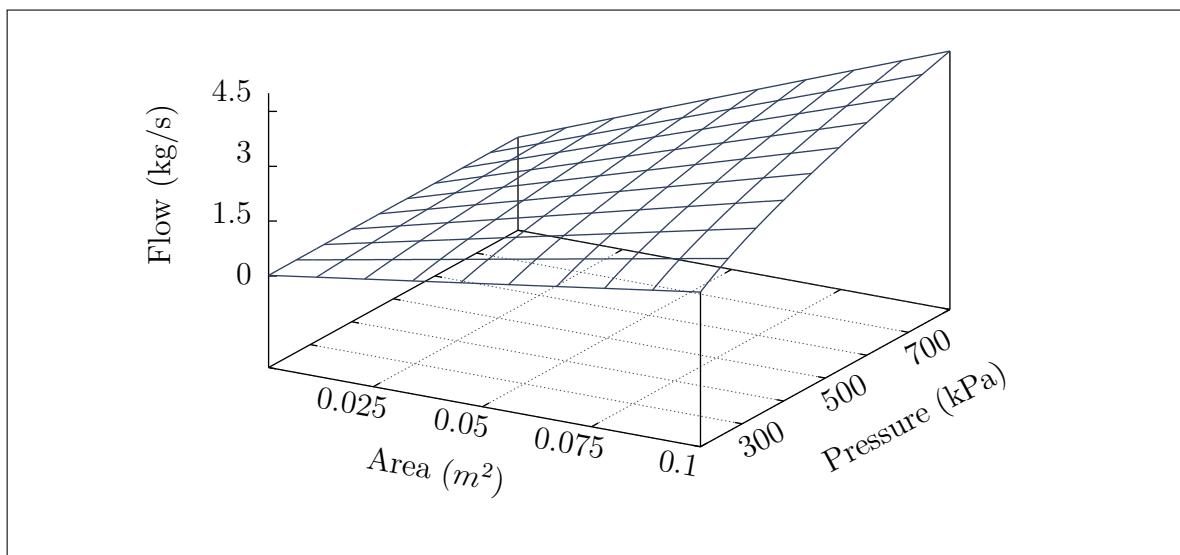


Figure 2.4: The flow due to leakage as a function of inlet pressure and leakage area¹

Leaks are often not easily detected through visual methods. In industry, many techniques can be employed to detect air leaks. Pascoe [32] and van Tonder [13] summarised these techniques as follows:

- Audible detection (Walk and report)
- Ultrasonic detection
- Intelligent systems²
- “Pigging”³
- Soap water/visible dyes

¹ efunda, “Orifice Flowmeter Calculator.” [Online] http://www.efunda.com/formulae/fluids/calc_orifice_flowmeter.cfm, [Accessed 18 October 2016]

² Leakage detection by using strategically placed measurement equipment and smart computer systems

³ The use of a device, “pig”, within the pipe to perform inspections

These methods are time and resource intensive, and many mines do not actively employ dedicated leakage detection and repair teams. Marais *et al.* [33] investigated streamlining the leakage detection and repair process to increase energy savings through the use of Compressed Air Leakage Documentation System (CALDS). The system was developed to allow centralised mobile leakage reporting. Usage of CALDS in mines resulted in an increased leak detection rate. One mine reported 24 leaks in a single month. It was noted in the study by Marais that there was difficulty quantifying the actual energy savings of the leakage repairs due to other interventions co-occurring.

Underground control valves optimisation

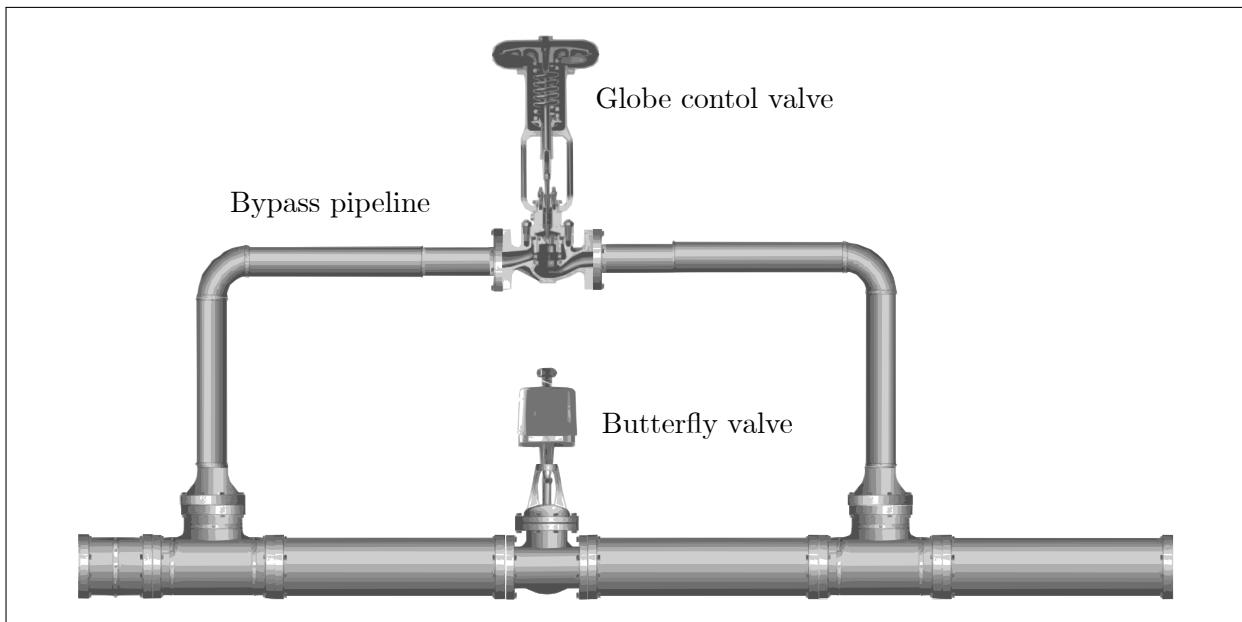


Figure 2.5: An illustration of a control valve [17]

Many mines utilise automated valves, as illustrated in Figure 2.5, at critical locations or on levels in the compressed air network. These valves control the pressure, restricting airflow from that point in the air network. Restricting airflow reduces losses resultant from network inefficiencies and leaks.

Kleingeld and Marais [26] found that optimising control valve control on mining levels can conservatively lead to 20% energy improvement on mines where no control valves are installed. For systems that already have some form of network control, between 10 and 15% savings can conservatively be achieved.

From literature, the advantage of control valve optimisation is a significant savings that can be achieved with relatively short set-up up time. Savings can be reached incrementally with each control valve installation. Previous studies have looked at simplified estimations of

energy improvement that may result from underground control valve optimisation. However, integrated simulation has not been used to accurately determine the energy savings and shaft pressure improvements that may occur [26], [33], [34], [25]

Improving pneumatic rock drill efficiency

Pneumatic rock drills are one of the largest air consumers in a mine. However, Pneumatic drilling systems convert energy very inefficiently. Replacing pneumatic drills with more efficient alternatives such as hydraulic or pneumatic drills would lead to large energy savings [32]. Alternatively improving the efficiency of pneumatic drilling can have a significant energy impact on the system, without the cost and safety concerns of alternative drilling technologies.

In a study by Bester *et al.* [35] looking at the effect of compressed air pressure on energy demand, Bester showed that between 2002 and 2013 compressed air and energy consumption per tonne of ore produced had steadily increased. The shift in energy and air volume per tonne of produced ore over time is illustrated in Figure 2.6.

The increase of air consumption per Tonne was a result of reduced air pressure at the mining areas. The pressure reduction caused a drop in the drilling rate, leading to higher air consumption. Pressure measurements as low as 300 kPa were recorded in these regions. In 2002 the drilling pressure at the mining section (stopes), was maintained above 500 kPa at most mines¹.

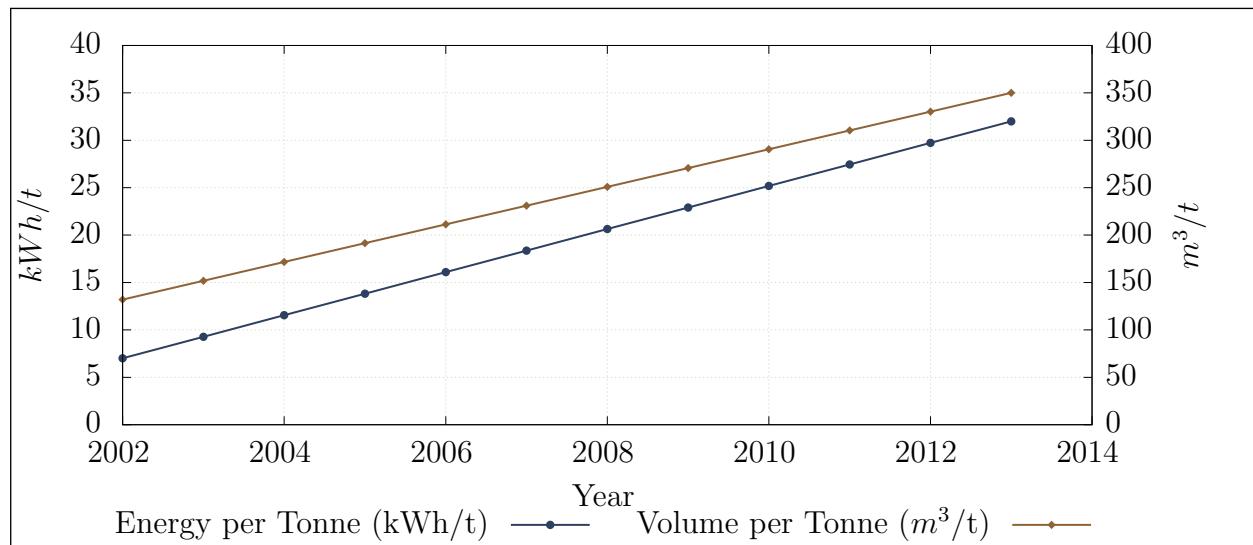


Figure 2.6: The Compressed air energy and flow consumed per Tonne of ore produced. Adopted from Bester *et al.*[35]

From literature, it is shown that lowering the pressure reduces the efficiency and drill rate

¹ H. Heller, ‘Compressor Performance Data,’ Impala Platinum Mine UTS , 2002 - 2013.

of rock drilling, leading to higher air consumption. Interventions that reduce systemic air losses or optimise supply can increase the operating pressure. Increased pressure, during the drilling shift, may add more value than the energy cost savings that can be achieved at a lower pressure [35].

2.3.4 Summary

This section reviewed previous studies that had achieved energy improvements in compressed air. The literature was divided into studies that focus on improving the supply of compressed air and those that optimised the compressed air demand.

Compressed air supply interventions included optimising compressor control and reconfiguring compressor networks. On the demand side, studies that investigated reducing leaks, optimising underground valve control and improving rock drill efficiency were discussed.

2.4 Use of simulations to identify improvements in mining systems

2.4.1 Preamble

The value of simulation in the mining industry has been widely shown through its usage in Demand Side Management (DSM) initiatives. Simulation and estimation procedures have been widely applied to identify, and pre-emptively quantify the effect of energy interventions strategies for water reticulation, cooling, compressed air and ventilation systems of mines.

The purpose of this section is to summarise and discuss the work that has been done concerning estimation and simulation of mining systems. This section will also discuss simulation development procedures and verification methodologies that have been applied in literature.

2.4.2 Estimating energy savings on mining systems

Estimation has been a vital tool to obtain the potential energy impact that can be achieved from interventions on a mining system. Before new tools allowed for the quick development of simplified simulation models, estimation techniques were frequently used to determine the feasibility of energy interventions.

The shortcomings of the estimation approaches are that they typically rely on simplified system models. The simplifications lead to high prediction error, limit the complexity of

scenarios and input/output variables.

Snyman [14] used mathematical estimation to determine the expected power savings from initiatives on mining compressed air systems. Due to uncertainty in the estimations, [14] predicted results were provided as a range between conservative and best-case estimates. The actual achieved energy impact would fall between these estimations. Snyman's model only estimated the average energy impact and could not provide a resultant power profile or other output variables such as pressure.

2.4.3 Benchmark modelling

Cilliers [36] developed “best practice models” using the Corrected Ordinary Least Square (COLS) benchmarking method. These models provide an energy benchmark that can be used to identify the scope for energy improvement on mining systems. An example of a benchmark model for a mining compressed air system is shown in Figure 2.7. The model can be used to estimate optimal energy required as a function of the quantity of ore mined (T) and the depth of the mining shaft (Z). Cilliers also developed benchmark models for mine cooling, water reticulation and ventilation systems.

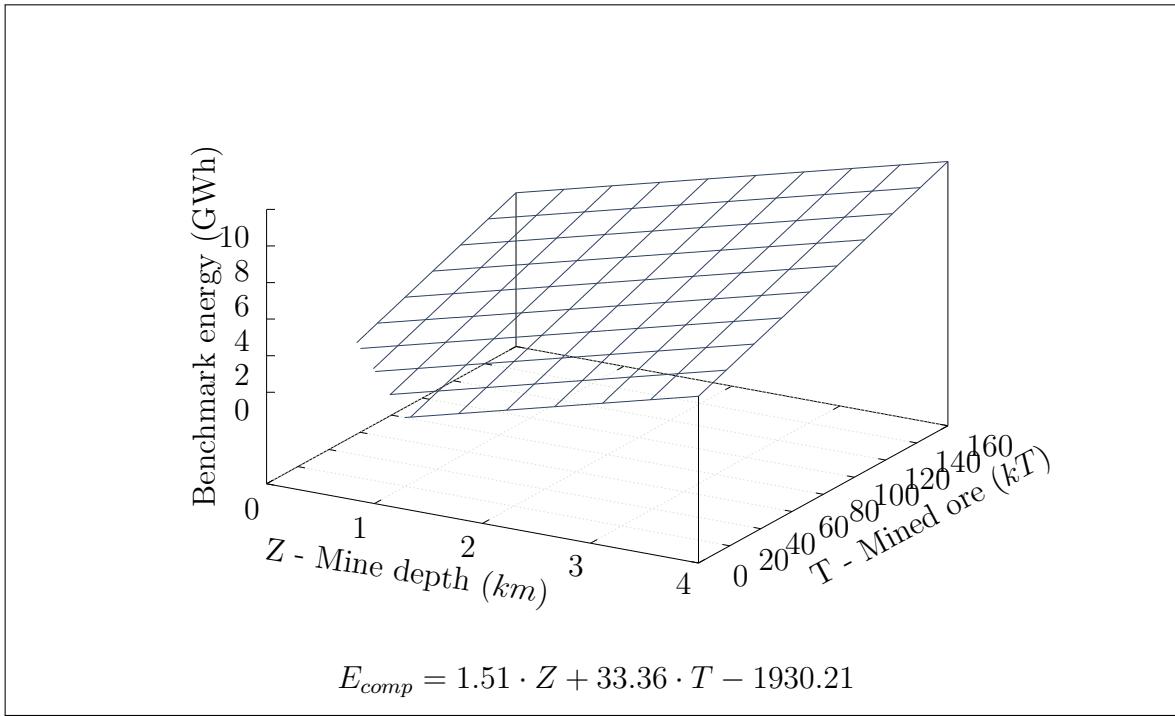


Figure 2.7: Energy benchmark model for a mining compressed air system, adapted from Cilliers [36]

2.4.4 Value of simulation in mining DSM

Van Niekerk [37], [38] investigated the value of simulation models in mine DSM projects. Van Niekerk developed simulation models for compressed air and water reticulation systems using KYPipes's simulation engine.

Simulation has been used in studies as a tool to improve mine cooling. Holman [39] investigated improvements to mine cooling systems that improve performance and efficiency. In the study, Process Toolbox (PTB) was used to develop and solve simplified simulation models to investigate cooling interventions.

The scenario Holman simulated showed potential average power reduction of 136 kW which would lead to an annual energy cost saving of R 0.55m. The study could be improved by increasing accuracy of the simulation. Power difference of as high as 31% between the simulation and actual was observed for some time periods.

Simulation has added great value for mining compressed air DSM. A detailed literature discussion of compressed air simulation applications is discussed in Section 2.6.

2.4.5 Simulation procedure

A structured simulation procedure is required to achieve the objectives of the study. Previous studies from literature have focussed on how simulation software tools work rather than the simulation procedure that was followed [40].

Bouwer [41] developed a software tool that he used to investigate extensive thermal and energy systems. Bouwer used the following procedure to implement simulations:

1. Create a detailed schematic of the system
2. Obtain data from installed instrumentation
3. Perform manual measurements where necessary
4. Gather data for a typical operational period
5. Convert data into useful formats
6. Setup the simulation model
7. Calibrate the model until simulated outputs match measured data

Bouwer's simulation procedure provides a good guideline for mining system simulation. The procedure could be simplified by merging steps. For example step 1 to 4 could be merged into 1 step "Gather system data and information".

2.4.6 Periodic/repeated simulation procedure

The concept of a periodic or repeated simulation is the execution a simulation multiple times or on a recurring schedule while altering input parameters. Hollander and Lui [42] used repeated simulations to estimate travel time distributions for traffic networks. Hollander and Lui developed the following methodology for the repeated simulations:

1. Gather input parameters
2. Run simulations
3. Read simulation outputs
4. Repeat for desired iterations

Within the mining context, Snyman [14] used repeated estimation to determine an upper and lower bound for the expected result. Another example from industry is real time simulation or estimation systems.

Van Heerden *et al.* [43] developed a dynamic control system for a mine compressed air network. The controller utilised repeated simulation to estimate unmeasured and future parameters of the network. The estimated results were used to optimise the compressor control.

2.5 Simulation software tools

Available tools

A software tool is required to create and execute/solve the compressed air simulations. In literature, a variety of software tools have been used to simulate industrial systems. A suitable software system for this study must be capable of producing accurate models of complex compressed air networks. The system should be able to solve for transient compressed air simulation scenarios with dynamic data and control inputs.

The development of system models should also not be excessively time-consuming. The software tools should also be able to handle missing data inputs and information that is common in mining systems. These were shortfalls of simulation tools noted by van Tonder [44] and Marais [40].

The following simulation tools that may meet the criteria were identified :

- Arrow ¹
- PTB
- Flownex ²
- KYPipe GAS ³
- AirSim
- REMS

PTB was used for this study. PTB was selected as it was developed specifically for simulation of mining systems. The design allows simpler and speedier development of system models as many components such as compressor mathematical models have already been pre-built [40].

PTB background

2.5.1 Simulation model verification strategies

Due to a lack of instrumentation, measurement inaccuracies and the sheer scale of mining systems, it is impossible for a simulation model to match the actual system's operation perfectly. From the literature, methods of verifying simulation precision were investigated. The verification techniques identified from literature were the Mean Residual Difference (MRD), Mean Absolute Error (MAE), the coefficient of determination or correlation and Mean Square Error (MSE).

Mean residual difference method

The average difference method looks at the mean of the actual and simulated time series. Relative error is then calculated with Equation (2.1). The simulation percentage error from the physical system is then calculated by dividing the error by the Actual data-points, Equation (2.2).

$$\bar{R} = \left| \frac{1}{N} \sum_{n=1}^N (A_n - S_n) \right| \quad (2.1)$$

The equation is rewritten to get a relative error percentage:

$$Err\% = \left| \frac{1}{N} \sum_{n=1}^N \left(\frac{A_n - S_n}{A_n} \right) \right| \times 100\% \quad (2.2)$$

¹ <http://www.aft.com/products/arrow>

² <https://www.flownex.com/>

³ <http://kypipe.com/gas/>

Where:

- A Actual system time series
- S Simulation time series
- n Data point
- N Number of data points in simulation period

A major disadvantage of this method is that for transient simulation, the positive and negative errors for individual points can cancel out. This leads to a smaller resultant error than would be expected. The resultant error value can therefore not direct to any conclusive statements regarding the accuracy of the model [45]. This strategy is not recommended if used alone to verify transient simulations.

Yu-jie Xu *et al.* [46] developed a steady state simulation of an absorption chiller. In the study, residual difference was utilised as a measure of the steady state error for an absorber model. The accepted margin of accuracy in the study was a relative steady state error of 5%.

Mean absolute error method

The MAE verification method follows a similar calculation as in the Average residual difference method. However, as shown in Equation (2.4), the error is calculated individually for each point in the series. The average of the individual errors is the resultant error, Equation (2.3). The relative error percentage is obtained by dividing each error is by the Actual value at that time step as in Equation (2.4).

$$MAE = \frac{1}{N} \sum_{n=1}^N |A_n - S_n| \quad (2.3)$$

The equation is rewritten to get a relative error percentage:

$$Err\% = \frac{1}{N} \sum_{n=1}^N \left| \frac{A_n - S_n}{A_n} \right| \times 100\% \quad (2.4)$$

Where:

Coefficient of determination

The Coefficient of correlation is the measure of how accurately a data series (x) can be represented in a linear relationship with Data series (y), i.e. $y = mx + c$. The value for the coefficient ranges between -1 and 1 where a value of 1 indicates a perfect linear relationship

- A Actual system time series
- S Simulation time series
- n Data point
- N Number of data points in simulation period

between the series and a value of -1 represents a perfect negative linear relationship. A value of 0 indicates that there is no connection between the data series. The correlation coefficient can be calculated using Equation (2.5) [45].

$$r = \frac{\sum_{n=1}^N (A_n - \bar{A})(S_n - \bar{S})}{\sqrt{\sum_{n=1}^N (A_n - \bar{A})^2 \cdot \sum_{n=1}^N (S_n - \bar{S})^2}} \quad (2.5)$$

Where

- A Actual system time series
- S Simulation time series
- n Data point
- N Number of Data point in simulation period

The coefficient of determination or R-Square value can be calculated by squaring the correlation coefficient (r).

Kurnia *et al.* [47], [48] developed a simulation for a novel underground mining ventilation system. Kurnia selected the mathematical model with the highest precision when compared with historical data points. The chosen model had an R-Square value of 0.96 and a relative error 30 %, using the mean absolute error method.

Mean squared error

In statistics, the MSE or Root Mean Square Error (RMSE) is the average of the square of the error between the actual and estimated value. The value is always positive. A smaller value relates to a more accurate model¹.

$$MSE = \frac{1}{N} \sum_{n=1}^N (A_n - S_n)^2 \quad (2.6)$$

¹ University of Kentucky Department of Mathematics, “Estimators, Mean Square Error, and Consistency,” [Online] <http://www.ms.uky.edu/~mai/sta321/mse.pdf>, [Accessed 3 March 2017].

Where:

- A Actual system time series
- S Simulation time series
- n Data point
- N Number of data points in simulation period

Other methods

Many alternative methods or variations of the methods discussed are available to verify transient simulations. As an example, Arndt [49] looked at the percentage of relative errors under certain limits as well as a maximum relative error. The method is an improvement compared to the residual difference in ensuring transient simulation accuracy however the results are difficult to interpret.

Sarin *et al.* [45] compared many methods. Some methods that have not been discussed in this study include:

- Vector Norms.
- Sprague and Geers Metric [50], [51]
- Russells error Measure [52], [53]
- Normalized Integral Square Error
- Dynamic Time Warping (DTW).

The calculations in the above methods are relatively complex. However, they provide additional measures such as phase, magnitude and slope errors. These metrics could add value in verifying simulations.

Comparing verification methods

The difference between the strategies is best shown using an example. Figure 2.8 shows the output and actual power of a simulation of a mining system for a 24-hour duration. In the study Maré [40] used MRD, Equation (2.2), to determine the accuracy of a simulation model.

With MRD, positive and negative differences between the simulated and actual profiles cancel out. The average of the two power profiles were very similar. The calculated residual difference relative error was therefore calculated as 1.17%. However, Using the relative MAE, Equation (2.4), applied to the same data series results in a relative error of 15.2%. There is a significant 14% difference between the two measures.

The results of other verification methods applied on the example are provided in Table 2.1.

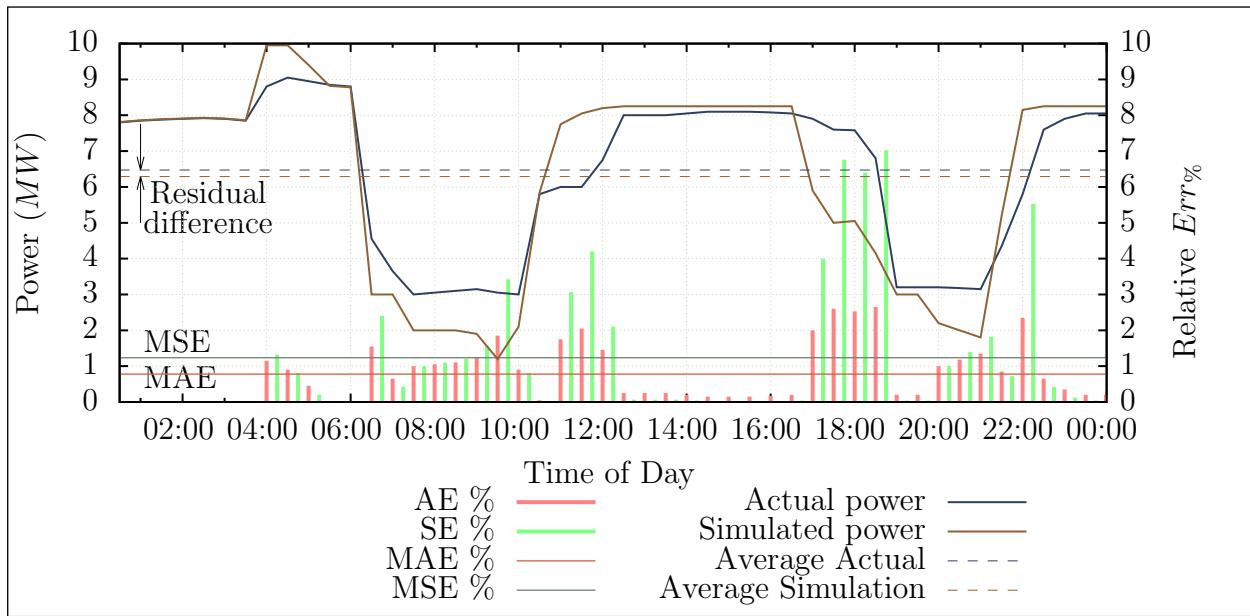


Figure 2.8: Example of simulation error calculations. Data adapted from Marè [40]

From the table it is clear that the accuracy of the simulation model is not as high as interpreted using the residual difference alone.

Verification method	Result	$Err\%$
Residual difference	0.06 MW difference	1.17%
MAE	0.778 MW error	15.2%
MSE	1.236	-
Coefficient of determination	$r^2 = 0.857$	-

Table 2.1: Results of the comparison of verification methods.

Willmott [54] studied the Advantages of the use of mean absolute error MAE over the RMSE method in assessing model accuracy. In the study Wilmott concluded that the MSE measure is a function of MAE and therefore does not describe average error alone. From the analysis, MAE was described as the most natural and unambiguous measure of average error magnitude.

Verification usage in previous simulation studies

Previous studies verified simulation accuracy through different methods and varying degrees of precision. Table 2.2 summarises these approaches. The majority of local literature utilised the mean residual difference to determine the estimation accuracy. However, from the literature it clear that the mean residual difference does not necessarily indicate model accuracy.

Table 2.2 shows the majority of local studies utilised residual difference to determine simulation

accuracy. However, literature has shown this is not valid method for transient simulation verification. International studies used correlation, MAE and MSE for verification. The accepted accuracy margin ranges from 3% relative error to up to 30%. The majority of studies used either 5% or 10% as the relative error cut-off.

Study	Year	Verification method	Accepted margin
Arndt [49]	2000	Mean and maximum absolute % error	% of time where $Err\% < 10\%$ and $Err\% < 5\%$
Bouwer [41]	2004	Mean residual % difference	$Err\% < 10\%$
Marais [6]	2012	Mean residual % difference	Not specified
Van Niekerk [38]	2012	Mean residual % difference	$Err\% < 10\%$
Bredenkamp [29]	2013	Mean residual % difference	$Err\% < 10\%$
Holman [39]	2014	Mean residual % difference	Not specified
Kriel [6]	2014	Mean residual % difference	$Err\% < 10\%$
VanTonder [44]	2014	Mean residual % difference	$Err\% < 3\%$
Kurnia <i>et al.</i> [47], [48]	2014	Coefficient of determination Mean absolute error	$r^2 > 0.95$ $Err\% < 30\%$
Dominic [55]	2014	Mean squared error	$< 1.7e^{-3}$
Du Plessis <i>et al.</i> [56]	2015	Mean residual % difference	$Err\% < 7\%$
Pascoe [32]	2016	Mean residual % difference	$Err\% < 5\%$
Peach [57]	2016	Mean residual % difference	Not specified
Maré [40]	2016	Mean residual % difference	$Err\% < 5\%$
Yu-jie-Xu <i>et al.</i> [46]	2016	Mean residual % difference	$Err\% < 5\%$ steady state

Table 2.2: Simulation verification methods that were implemented in previous studies.

2.5.2 Summary

In this section, the use of simulation and estimation techniques to identify improvements in the mining industry was reviewed. Estimation and benchmark procedures in literature were first discussed. A review of simulation usage in mining DSM for water, compressed air and cooling systems were then reviewed.

Simulation procedures in literature were examined. Various simulation software tools used in industry were then compared. Finally, an analysis and comparison of verification procedures were performed.

2.6 Review of the use of simulation in compressed air optimisation

2.6.1 Preamble

This section will review literature regarding simulation usage in optimising compressed air systems. Shortcomings identified from the literature that can be improved upon in this study will be addressed. The studies that were reviewed in this section are summarised in table Table 2.3.

Study	Year	Software	Application
Mousavi <i>et al.</i> [58]	2014	Airmaster AirSim	Compressor energy modelling
van Niekerk <i>et al.</i> [37]	2013	KYPipe	Compressed air DSM project design
Bredenkamp [29]	2013	KYPipe	Compressor relocation
Zahlan and Asfour [59]	2015	MATLAB	Determining the optimal compressor location
Pascoe [32]	2016	PTB	Optimised surface valve control Exchanging compressors
Maré <i>et al.</i> [19]	2017	PTB	Various compressor and air network optimisations

Table 2.3: Summary of reviewed compressed air simulation studies.

2.6.2 Compressor energy modelling

2.6.3 Compressed air DSM project design

2.6.4 Determining the optimal compressor location

Determining the optimal location for a compressor can improve the efficiency of an air network. This improvement can lead to significant energy cost savings. In a study, Zahlan and Asfour [59] investigated the use of simulation to solve this problem in the manufacturing industry.

To determine the optimal compressor location, Zahlan and Asfour divided the manufacturing facility into a matrix rectilinear zones. A mathematical model was developed in MATLAB to determine the energy consumption of the compressors when located at each zone.

User preference was taken into account by determining a parameter of the extent of the user

preference for each zone in the matrix. The user preference parameter and energy savings per zone were weighted and summed to arrive at the optimal zone. Figure 2.9 shows the simulated energy consumption for the compressors at each zone location. The energy consumption was thereafter combined with a user preference matrix to determine the optimum compressor location.

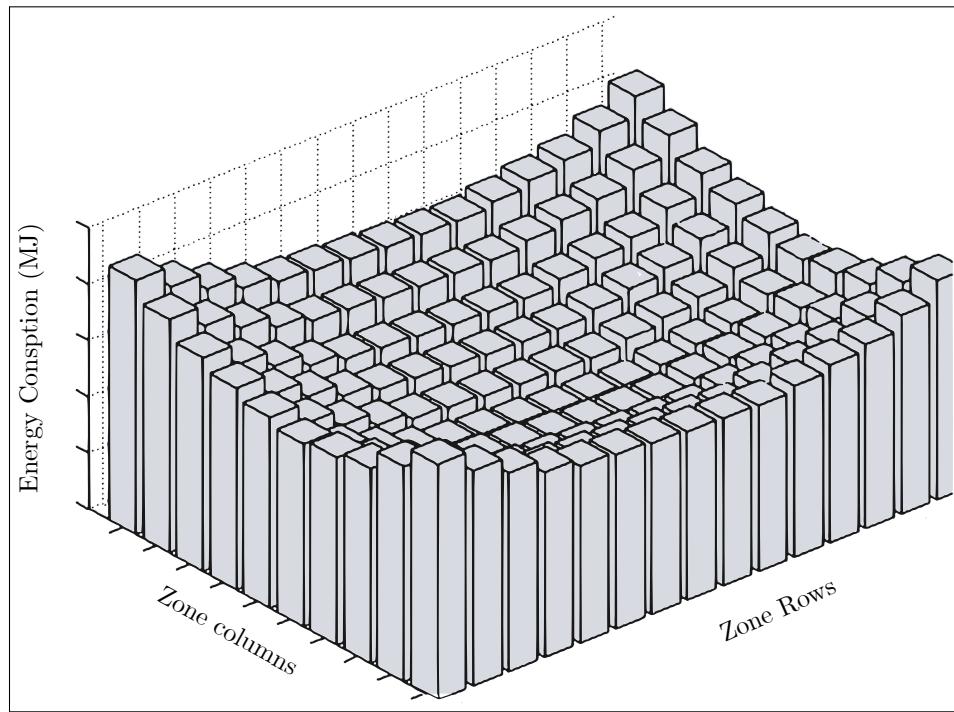


Figure 2.9: Resultant energy consumption matrix. Adapted from Zahlan and Asfour [59]

The approach to determining the optimum compressor location would not be suitable for mining compressed air system. Mining air networks extend for kilometres in the x, y and z planes. The study greatly simplified the compressed air network in order to easily formulate the mathematical model. This approach would not be suitable for larger and more complex mining air networks.

2.6.5 Compressor relocation in a mining system

Bredenkamp [29] utilised simulation models to test compressor relocation scenarios to improve the efficiency of a mining air network. The simulation model developed by Bredenkamp using KYpipe took into account the location, supply capacity of each compressor, as well as the surface pipe distances. The model, as visualised in Figure 2.10, simplified the air demand of the network to a single outlet flow per shaft.

The result of Bredenkamp's simulation showed a potential energy cost saving for the mine of R 170m over 13 years. The cost savings were proved it was feasible to purchase a new compressor at an estimated cost of R 15m. Additional surface pressure delivery improvements

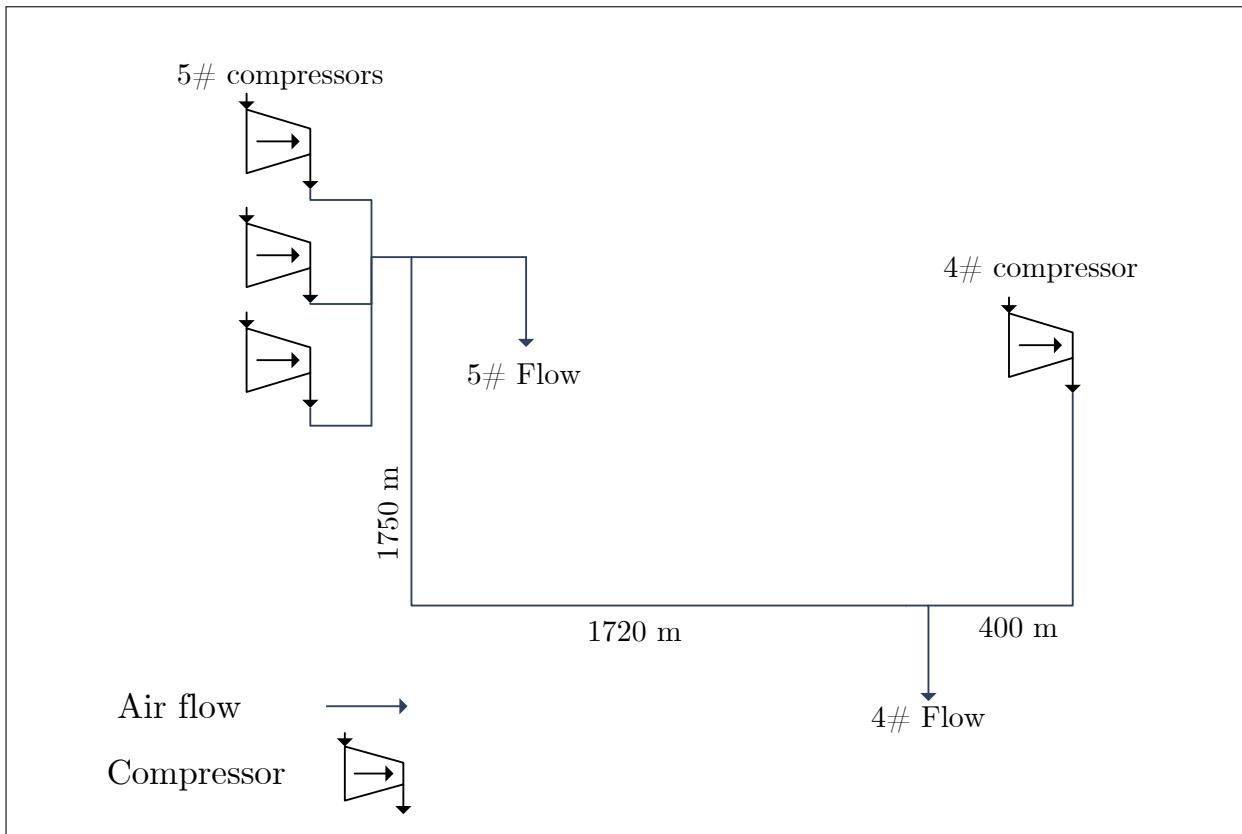


Figure 2.10: Compressor relocation simulation model. Adapted from Bredenkamp [29]

were also identified from the simulation.

Bredenkamp did not follow a clear simulation methodology for developing the compressed air simulation. Bredenkamp utilised the residual difference to calibrate and verify the model parameters. Therefore, the simulation precision and could be improved by following a detailed methodology.

Increasing the detail of the demand components would provide valuable results, showing more particular effects that result from the simulated scenarios. For example, the specific delivery pressures at mining level as a result of the intervention. Additionally, more scenarios or scenario combinations could be simulated with potentially greater savings.

2.6.6 Compressed air ring

Pascoe [32] developed a simulation model for a compressed air ring. The purpose of the simulation was to identify the benefits of reducing pressure during the blasting shift period through surface valve control. In a second case study, Pascoe used simulation to investigate the control benefits of exchanging a large compressor with two smaller compressors.

The model required complex supply side detail including modelling of the individual com-

pressor, locations of compressors and pipe lengths and control valves. The demand aspect of the system was simplified to a single flow per shaft, decline or processing plant. A schematic of the simulation model is shown in Figure 2.11.

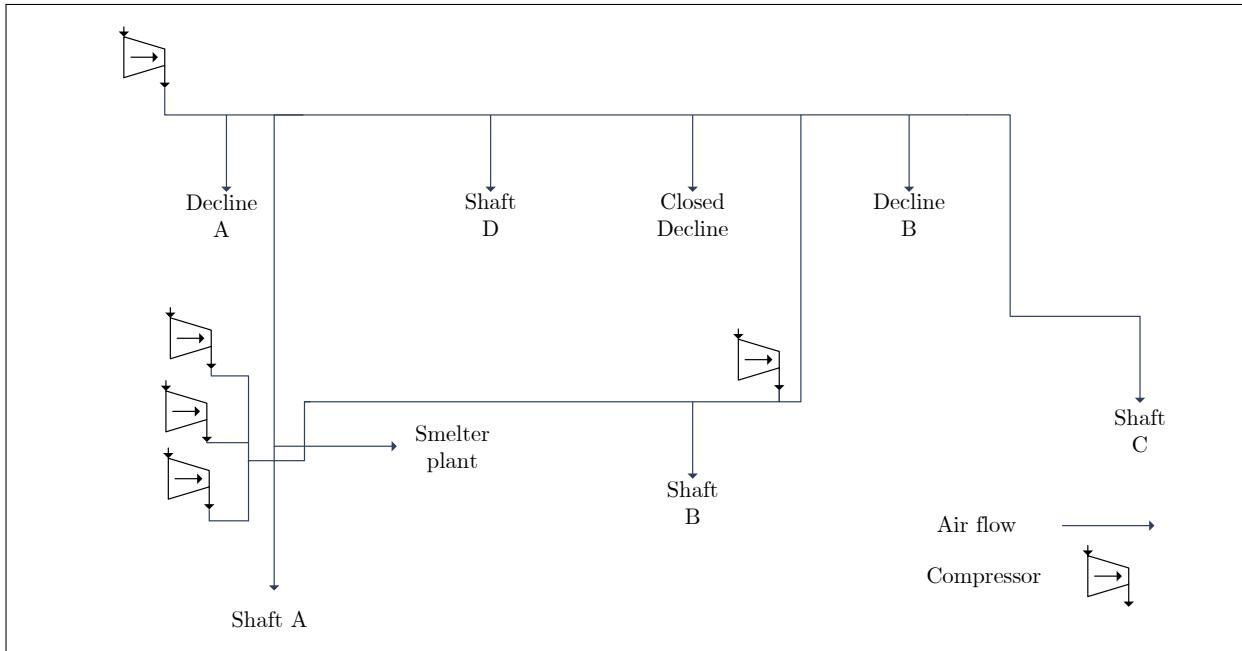


Figure 2.11: Simplified compressed air ring model. Adapted from Pascoe [32]

The methodology for simulation development used by Pascoe can be improved. For example, as was the case with Bredenkamp [29], [32] utilised the residual difference to calibrate and verify the transient simulation output variables. The simulation precision could therefore be improved through the use of a suitable transient error calculation method as discussed in Section 2.5.1. Increasing the detail of the demand components could also add value.

2.6.7 Mine complex

Maré *et al.* [19] developed a compressed air simulation for a mining complex. In the study [19] simulated and prioritised several scenarios with the goal of reducing energy and other operational costs.

Maré accurately modelled the individual compressors at each shaft in the mining complex. The detailed flow consumption at individual mining levels was found to be inaccurate. Therefore, the process boundaries for the model were selected to include only the air flow consumption at the shaft. A process schematic for the simulation mode is shown in Figure 2.12.

Eight scenarios were simulated using the model. The combined results showed that there was a potential energy saving of R 1.5m as well as an additional pressure improvement of 51 kPa. The infrastructure and resource costs were estimated to prioritise the interventions for

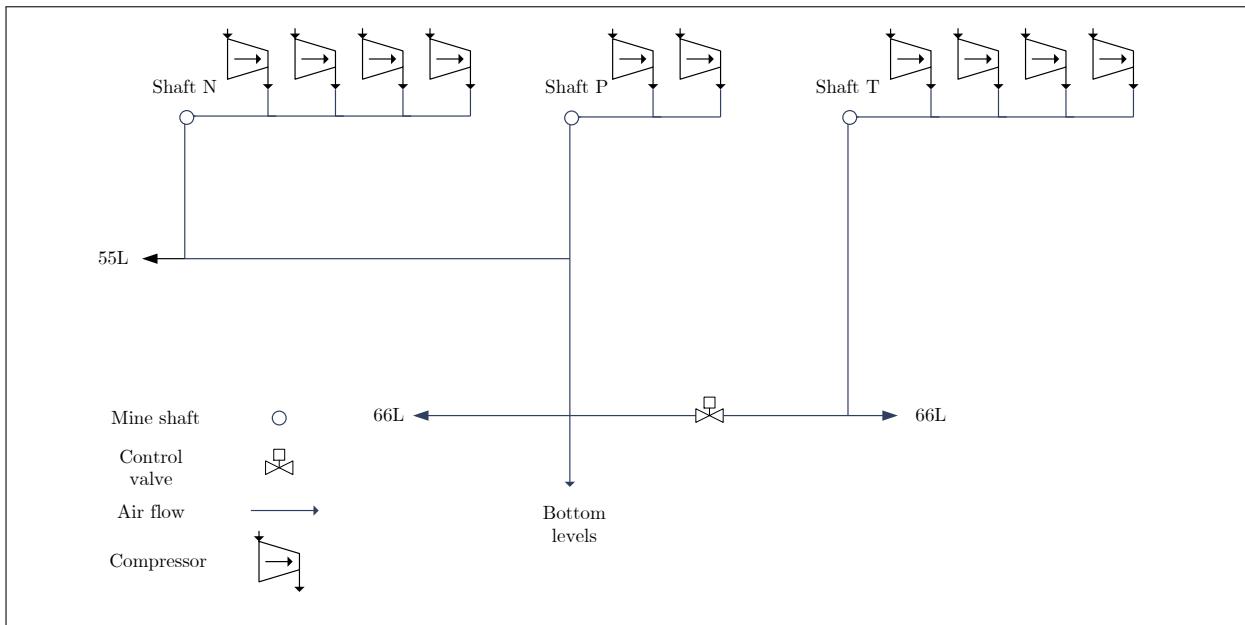


Figure 2.12: Simulation model for a complex air network. Adapted from Maré *et al.* [19]

the greatest financial benefit to the mines.

Maré used the residual difference to calibrate and verify the accuracy of the transient simulations. Therefore, the actual estimation error is greater than what is shown in the study. Through on site inspections and strategic measurements, the detail of the simulation model could be improved this could allow for more accurate simulations and more simulation scenarios.

2.6.8 Summary

Previous work in compressed air simulation was summarised and reviewed in this section. From the review of the literature, shortcomings and areas for improvement were identified. In general increasing the detail of simulation models could increase the accuracy and allow for more simulation scenarios.

2.7 Conclusion

In Chapter 2, a comprehensive study of relevant literature was performed. The literature study aimed to:

- Provide background on mining compressed air networks
- Review compressed air energy interventions in industry
- Review the usage of simulation in the mining industry
- Review the usage of simulation in the mining compressed air

Compressed air background was provided. This included background regarding various compressed air sub-components, the operational schedules and the effect on compressed air requirements and finally, the typical instrumentation found in a mining compressed air system.

A review of previously achieved energy improvements in the compressed air was performed. The literature was divided into studies that focus on improving the supply of compressed air and those that optimised the compressed air demand.

The use of simulation and estimation in the mining industry was then reviewed. Estimation and benchmark procedures in literature were first discussed. The discussion was followed by a review of simulation usage in mining DSM. Simulation procedures from the literature were then reviewed. Various simulation software tools used in industry were compared. Finally, an analysis and comparison of verification procedures were performed.

Finally, the use of compressed air simulation was reviewed. The shortcomings, success and potential improvements of previous compressed air simulation studies were discussed.

CHAPTER 3

Developing a simulation methodology

“Great Design is iteration of good design.” - Dr M. Cobanli

3.1 Introduction

This chapter details the development and implementation methodology of simulations to optimise mining compressed air systems. The method developed uses insights gathered from the literature reviewed in Table 4.2.

Implementation of a simulation is divided into three steps as shown in the flow diagram, Figure 3.1. Firstly, an investigation on the specific air network is performed. The data acquired from the system survey is then utilised to develop and verify a simulation model. In the final step, scenarios are tested using simulations. The results are then quantified and prioritised. After the process has been reviewed, a simulation report is then produced and given to the responsible mine personnel. Each step will be discussed in more detail in the section that follow.

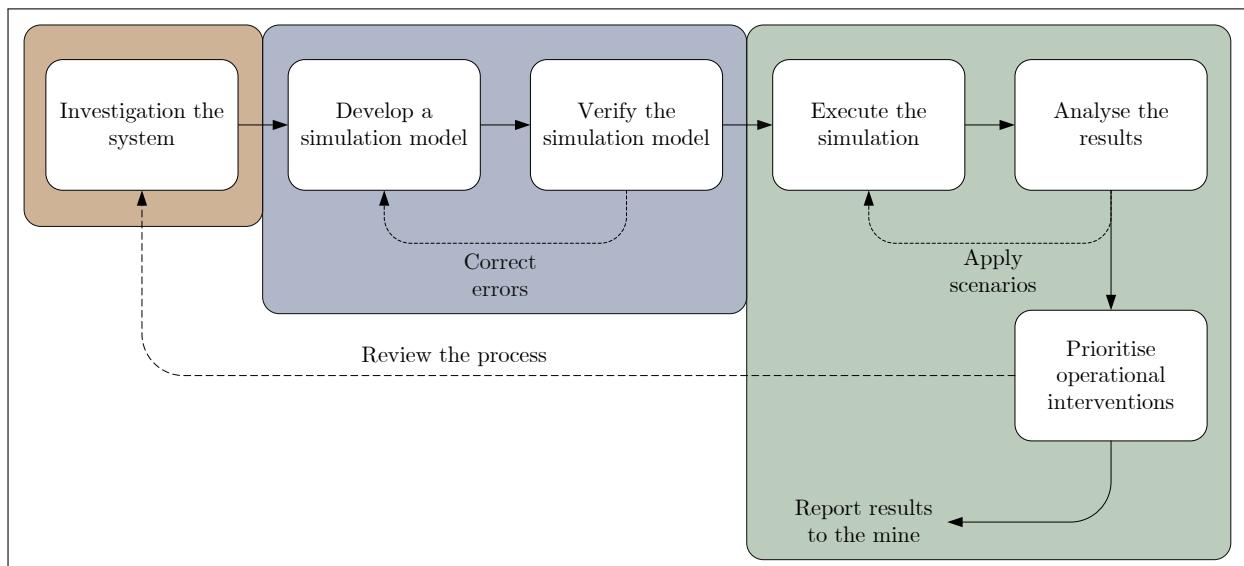


Figure 3.1: Flow diagram of the methodology for this study

3.2 Investigate the system

3.2.1 Preamble

Developing a detailed simulation model of a compressed air network requires thorough comprehension of the inner workings of the system. This section will discuss the investigations needed to obtain the required understanding.

3.2.2 Acquire data

The first step of the system investigation is to acquire the data and understanding that will be required to model compressed air system's function. This system survey will need access to resources such as data storage systems, instrumentation, and communication with relevant

engineers and personnel.

Comprehensive and up to date process layouts illustrate a compressed air network's unique set-up, scale and location of instrumentation. More detailed layouts can provide per-level air consumption breakdowns of the network, refuge bays areas, mining cross-sections and identified inefficiencies. The layouts are vital to understand the system process and identify what data parameters will be required for the model.

A baseline period¹ that best represents the typical operation of the mine is selected. Additionally, availability of data should be considered. The length of the baseline period is selected based on the scenarios that are to be tested; this can be changed later. For calibrating a compressed air system a 24 hour period of normal operation is usually sufficient. A longer period may be needed to verify the model.

3.2.3 Investigate mining schedules

A critical aspect of developing an accurate model of a mining compressed air system is the apprehension of the operational philosophy of the mine. The schedule for operations such as drilling, blasting or cleaning can have a major impact on compressed air requirements at different times of the day. By utilising the operational schedule, simulation scenarios can be optimised for the air requirements throughout the day.

3.2.4 Verify data accuracy

Data verification is the process where data is evaluated to ensure accuracy. It is important to verify data that is used for model development as an accurate representation of the operation of a system can only be achieved utilising data of high quality [60]. The factors that influence a dataset's quality, accuracy and integrity summarised as follows:

- Conversion of measurement value [61]
- Storage and collection of the system [62],[63]
- Traceability of measurement sources [63]
- Measurement equipment accuracy and malfunctions [60]
- Data abnormalities [60]

Therefore, a data verification methodology is utilised to ensure datasets are of high quality.

¹ A period that best reflects the typical operation before implementation of an energy intervention. This period is then compared with a period post-intervention implementation to calculate determine the improvements.

3.2.5 Resolve missing data

Data that is required to develop the simulation model, such as flows, pressures, may not be actively logged by mine systems. It is often necessary to investigate alternative sources and methods to obtain the data. For example, for process elements where instrumentation is absent, estimations can be made from assumptions made using instrumentation on the network or spot inspections.

Air network specifications such as piping sizes, technical layouts, major leak locations or specifications are often outdated or not recorded. Critical data should be obtained through audits and inspections of the system. If a manual inspection is not possible, estimations should be made from the available data.

3.2.6 Summary

This section discussed the method to investigate a compressed air system. The procedure described the processes for acquiring data and information regarding the specific compressed air network, process to evaluate and authenticate data accuracy as well as procedures to deal with situations where no data is available.

3.3 Develop and verify a simulation model

3.3.1 Preamble

Compressed air networks are comprised of components such as compressors, valves, pipes and other components. This section will discuss the development, calibration and verification of component models that make up a compressed air simulation.

3.3.2 Select the process boundaries and simulation parameters

The simulation boundaries determine the detail that the system process is modelled. For a simple compressed air model, the boundaries can be set around the compressor house. This model would then only include the compressor components, inlet and outlet air flows. Alternatively, a more complex model can be developed by choosing boundaries to include more aspects of the process such as specific flows on mining levels.

The boundaries should be selected based on the input data available, accuracy targets and available time and resources. A more detailed model will lead to more accurate simulation. However, it may take more time and resources to obtain the data required to calibrate the

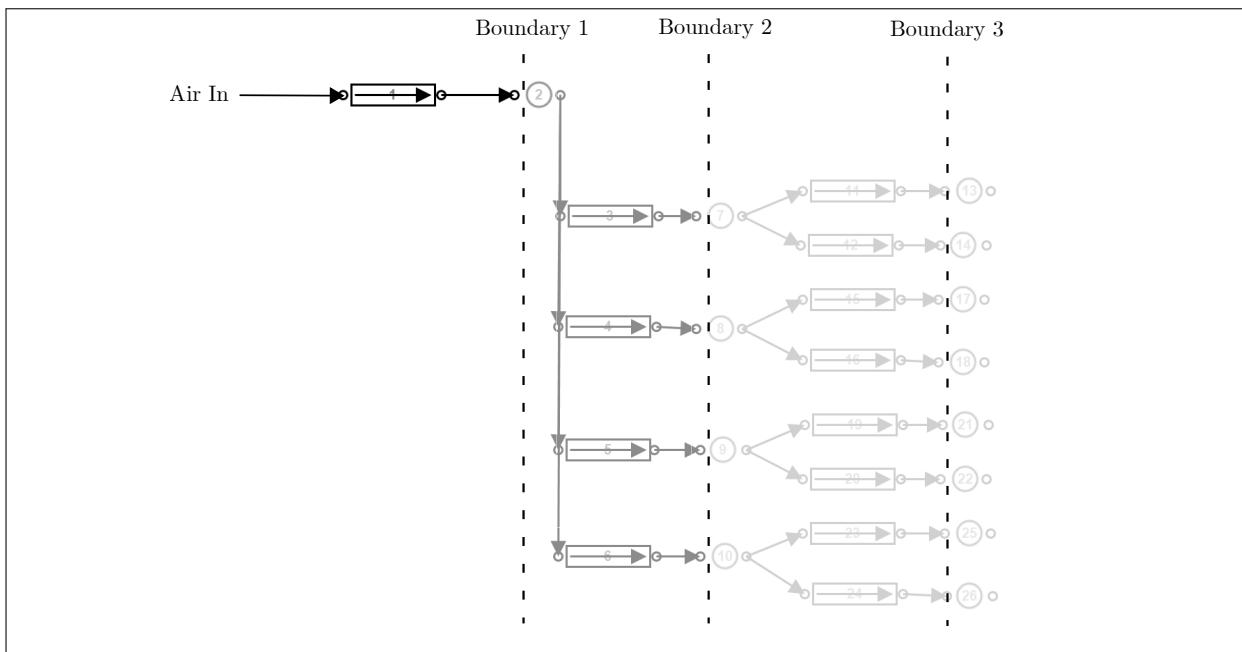


Figure 3.2: Selecting boundaries for the simulation model

model. Figure 3.2 shows an example of different boundary selection for the same system.

Period and data step sizes of the simulation is just as important. The period or duration of the simulation should be determined to ensure that the effects of a scenario are fully tested. Most commonly, previous studies have simulated a "typical" 24 hour period of operation. Most mining compressed air systems follow daily trends and patterns. Therefore, there is no need to simulate a longer period as day to day results would be very similar. A longer simulation may be required for cases where the system operation varies from day to day.

The simulation step size indicates the data resolution. A lower step size will result in a more accurate simulation of the system. However, the processing and data analysis is effected. In this study, the smallest available step size¹ is selected to ensure that the simulated results achieve the desired precision.

Compressed air processes such as opening/closing valves or compressors stopping or starting may occur within minutes or seconds. Therefore, Higher step sizes (30+ min.) sizes may delay process changes. This delay makes replication of the system control more difficult which reduces simulation accuracy.

A higher time-step resolution allows for more precise tuning of controllers and dynamic com-

¹ The minimum step size is determined from the logging interval of the input data instrumentation. For example, if all input data is logged at 10-minute intervals, the minimum step size would be 10 minutes.

ponents. If input data is not available at the desired resolution, the data can be interpolated using the appropriate method. An example of the application of linear interpolation to increase the time-step resolution as shown in Figure 3.3. However, incorrectly estimating the “in-between” data value may adversely affect the simulation accuracy.

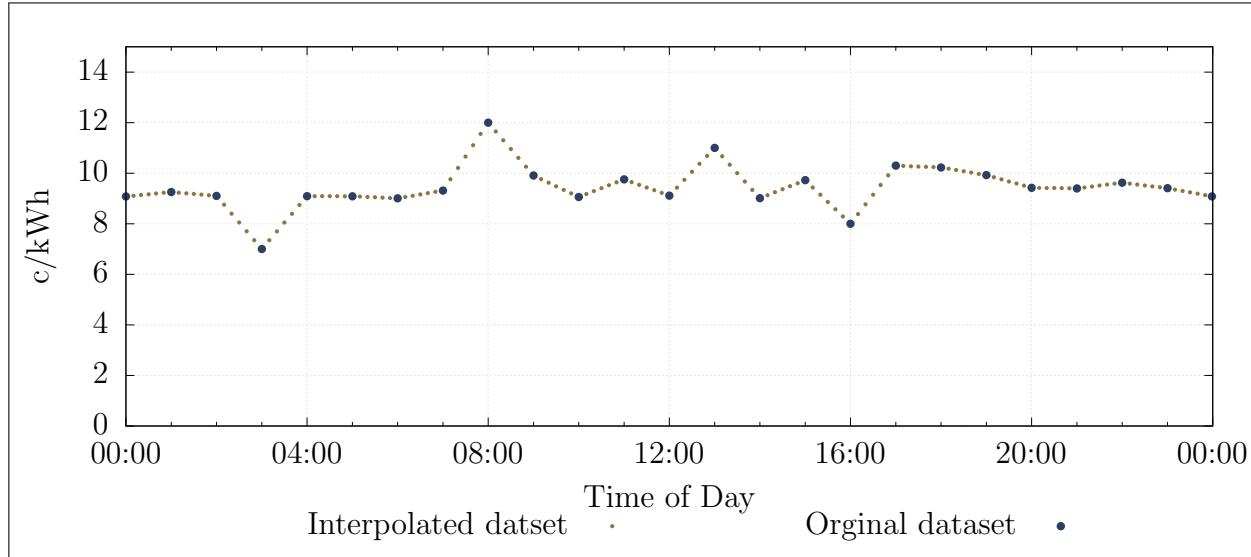


Figure 3.3: Interpolating input data to increase time-step resolution

3.3.3 Model compressed air network components

Ambient conditions

Ambient air condition underground and on the surface change the characteristics of the air, affecting the operation of the system. Figure 3.4 shows the average summer air conditions. If no data is available for the specific simulation period, the conditions can be estimated by scaling this profile.

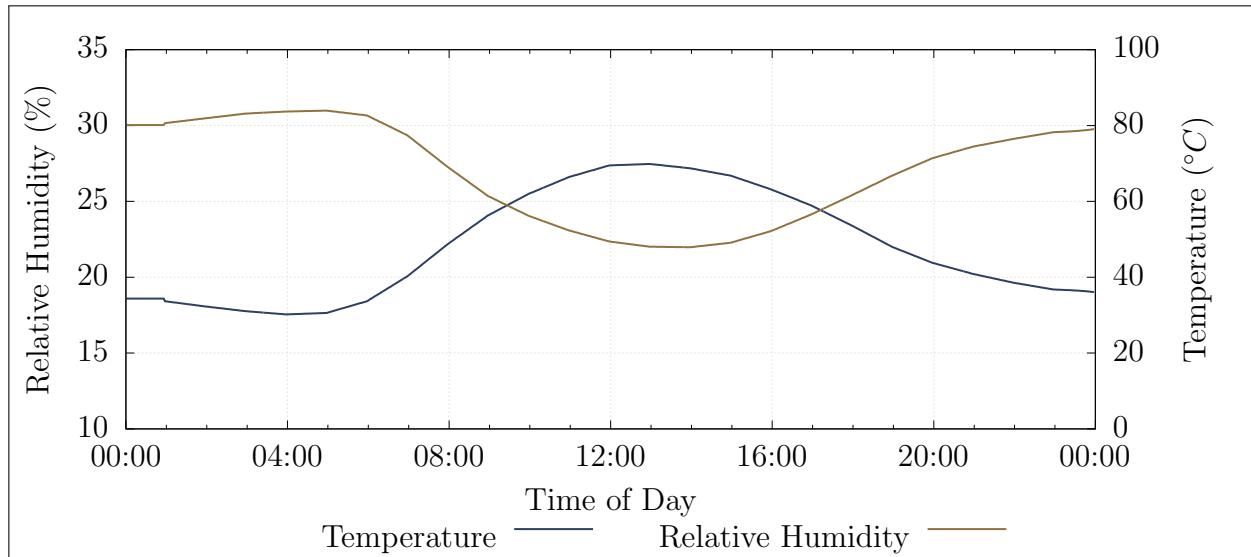


Figure 3.4: Average summer ambient air conditions at a South African gold mine

The assumption is made that underground conditions remain constant at each mining level. Pressure and temperature increase with depth as a result of auto compression and rock face temperature. Therefore, the conditions can be estimated using only the depth at each level.

Air pipes

Pressure losses occur over compressed air networks due to friction within the pipe; these losses should be taken into account in the simulation for large piping networks. A pipe model is used to account for these losses which are defined by the *Darcy-Weisbach equation*¹:

$$\Delta P = \frac{f L \rho V^2}{2D}$$

Where the pressure difference ΔP is a function of:

Parameter	Definition
f	Friction coefficient
L	Pipe length (m)
D	Pipe diameter (m)
ρ	Air density (kg/m^3)
V	Average velocity (m/s)

Table 3.1: Air pipe component model parameters

The pipe component may be used as a valve by controlling the open fraction between 0 and 1. Modelling the valve flow characteristics is discussed in Section 3.3.3 *Controllers*.

Compressors

Three compressor models were investigated, each with varying complexity. The models are:

- Air compressor
- Dynamic compressor
- Positive displacement compressor

The air compressor is a general, simplified model. It requires minimal user inputs by making several assumptions. This model is useful when parameters for a compressor are not available. Alternatively, the air compressor model is ideal when doing a quick or preliminary simulation. However, it is not ideal for detailed simulations which require more precision.

¹ B. Glenn, ‘The Darcy–Weisbach Equation,’ [Online] <https://bae.okstate.edu/faculty-sites/Darcy/DarcyWeisbach/Darcy-WeisbachEq.htm>, [Accessed 20-05-2017]

The dynamic compressor components are more complex, taking into account factors such as heat generated by the polytropic process and mechanical inefficiencies. Hence, the model can be used more accurately and for more complex simulations than the general compressor model. However, it should be noted that the dynamic compressor is simplified by several assumptions, for example, a constant efficiency at varying loads.

For most scenarios, the dynamic compressor model is most suitable. This component is modelled by fitting a quadratic curve through three points of operation to obtain an equation for corrected mass flow as a function of the pressure ratio. This characteristic curve of a compressor is shown in Figure 3.5 can be accurately estimated even when only one data point is available by making approximations for the zero flow and pressure points on the curve.

Once the flow characteristics of the compressors are set, the efficiency and Polytropic coefficient parameters are calibrated such that the output power and air temperature match the actual or estimated outputs of the compressor.

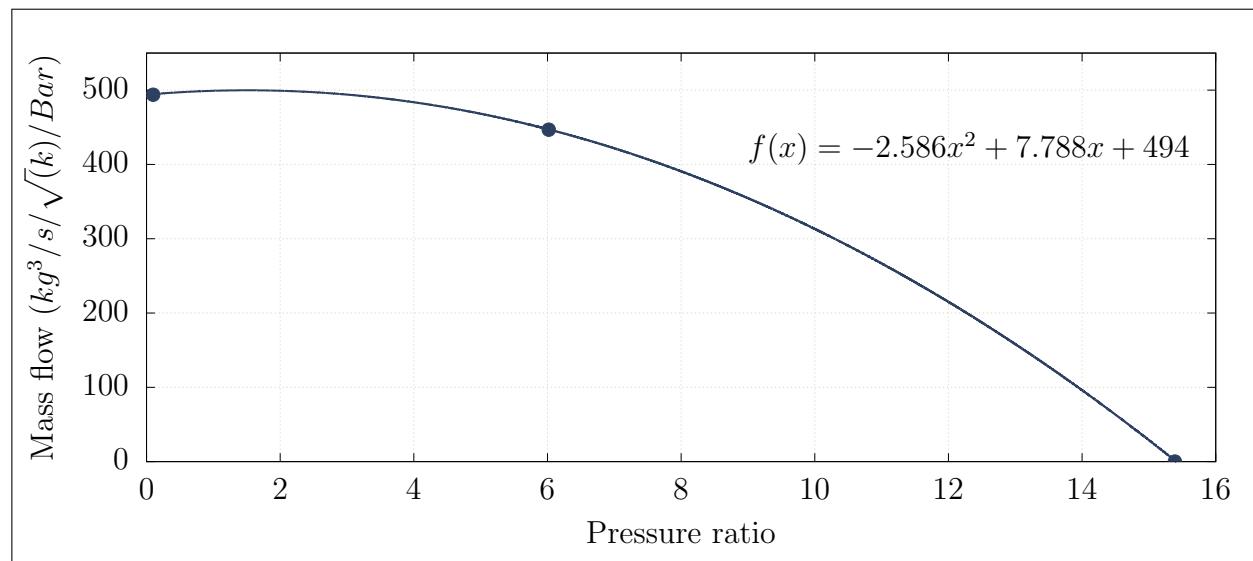


Figure 3.5: Estimating the characteristic curve of a compressor by fitting a quadratic function to points of operation

Once the models are accurately calibrated, the compressor component integrates into the air network in the arrangement shown in Figure 3.6. The Compressor is connected to the inlet air source via an inlet pipe and air node and the rest of the network via an air node and outlet pipe. The additional pipe components allow the inlet and outlet conditions to be monitored and controlled in the simulation.

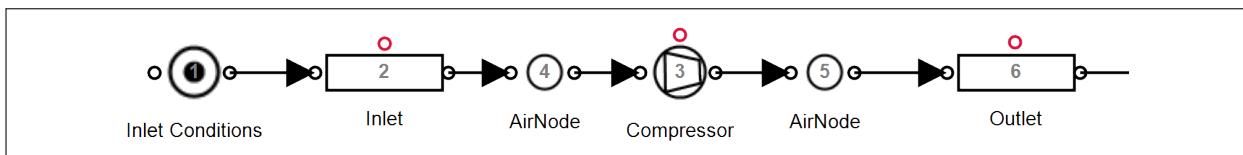


Figure 3.6: Integrating the compressor component into the simulation

Demand

A flow demand represents any air flow leaving the network. Flows leaving the network include any air consuming equipment such as drills and agitators as well as losses like air leaks and open pipes. The air flow is dependent on pressure and the specific resistance to flow of the outlet.

The resistance of the flow demand can be obtained using the inlet pressure, outlet pressure and flow. If the flow is not known, a reasonably accurate estimation can be made by calculating the expected flow from the size of the outlet. This estimation will affect the accuracy however.

The air demand may vary throughout the day. For example, a mining section may utilise more machines during certain periods of the day. A schedule and flow profile is used to replicate this in the simulation. Figure 3.7 shows how a calibrated air demand or leak is integrated into the simulation model on PTB.

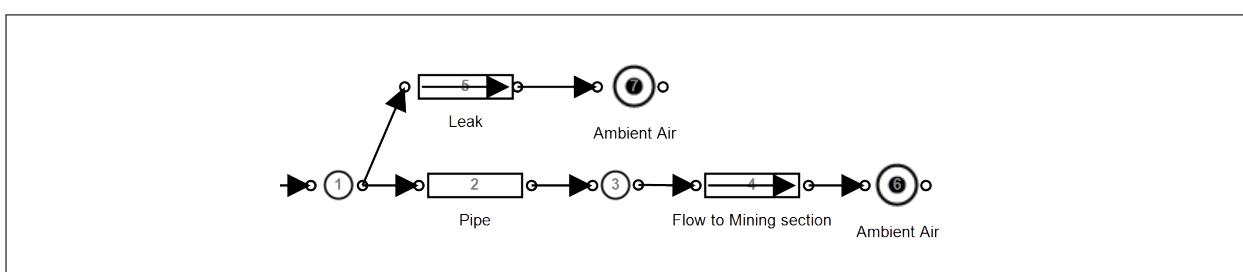


Figure 3.7: Implementing flow demands and leaks into the simulation

Compressed air control

Simulation components require dynamic control to replicate the operation of the actual air network. Control is typically implemented on compressors and valves throughout the network to follow setpoints and schedules. It is important to not only include the controllers in the simulation but to replicate any nonlinearities, limitations and response delays related to specific types of control. Implementing these control factors will ensure the model reacts in the same way the actual network would, improving accuracy.

On a typical mine, a compressor's power output is controlled to ensure that the discharge

pressure matches a specified setpoint. This control is achieved through either VSDs (or Variable Frequency Drives (VFDs)) and guide vane control. On PTB valve or compressor control can be replicated using a Proportional-Integral (PI) controller as shown in Figure 3.8. For the control system models in Figure 3.8, outlet pressure is used as feedback to the compressor and valve controller.

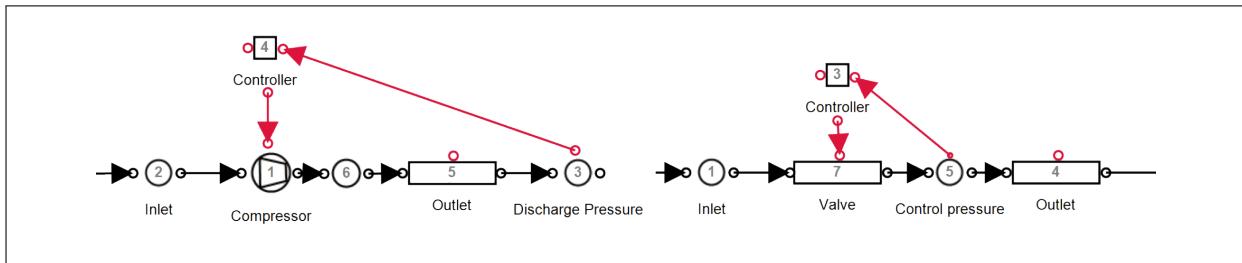


Figure 3.8: Control components in Process Toolbox

Guide vanes are most commonly used in mining air systems to control compressors. Guide vane control entails controlling the position of the inlet guide vane. The guide vane is opened or closed to control the compressors discharge pressure. Manipulating the guide vane position will affect the output power the compressor inputs into the system.

Figure 3.9 shows the relationship between power and guide vane position. A linear relation between guide vane position and compressor output can be used to estimate effect of guide vane control. The model should take into account the minimum guide vane position limit that is typically set at around 40% open. As illustrated in Figure 3.9 this control position maps to an output power for the compressor of about 60% of the maximum power. When more pressure is required than can be obtained with the guide vanes fully opened, another compressor is needed to operate.

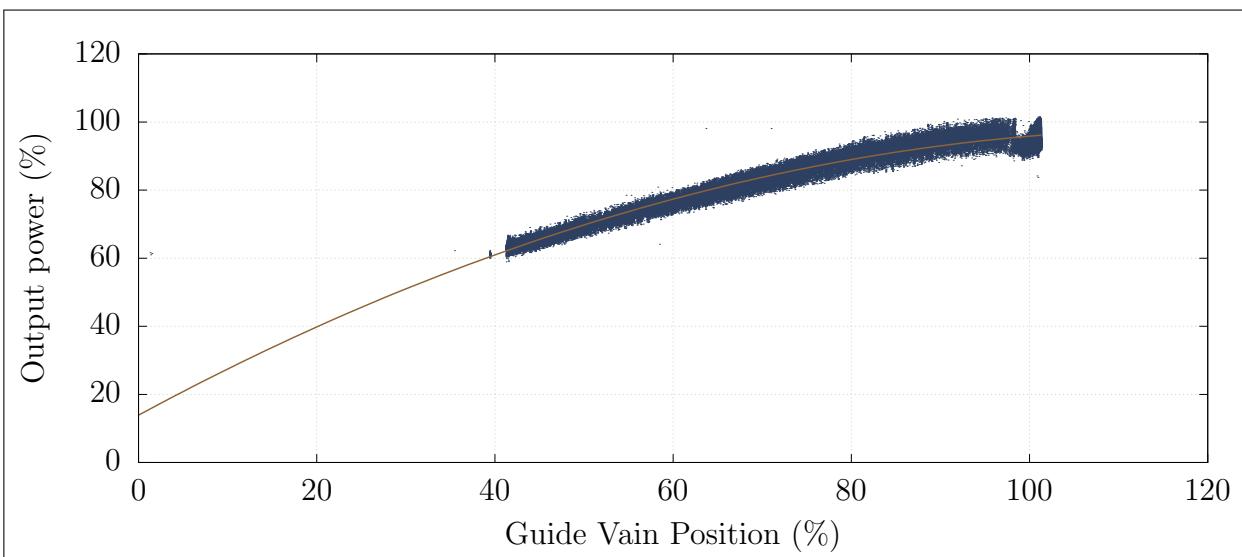


Figure 3.9: Modelling the compressor control from a guide vane¹

In PTB, the guide vane controller is modelled using a PI controller. The non-linear limitations of guide vane control must be implemented in the controller. The control limitation is applied in the model by using a minimum control output limit that matches the minimum power reduction achieved by closing the guide vane to its minimum position.

Mines utilise control valves at underground sections to adjust the pressure at individual mining stations independently [64]. Controlling of valve components is performed similarly as control of the compressor components. Figure 3.8 shows the outlet pressure is used as feedback for a PI controller. The control output is mapped to the valve fraction of a pipe component.



Figure 3.10: An example of a compressed air control valve[65]

Compressed air after-cooling

The air compression process generates significant heat. Compressed air at high temperatures contains a significant amount of water vapour. After-coolers are installed on compressed air systems to prevent condensation in the air network, improve the system capacity and to protect equipment from excessive heat [25].

After-cooling reduces the compressed air temperature out of the compressors. This cooling can affect the operation of the network. Hence, including after-cooling to the simulation model should improve accuracy.

Modelling the after-cooling is achieved in PTB using a heat transfer node at the outlet of the compressor component model. The heat transfer parameters shown in Table 3.2 should be

¹ Data recorded from a guide vane controlled compressor on a mine over a period of six months

Parameter	Definition	Unit
A	The heat transfer area	m^2
UA	Heat transfer coefficient	$kW/\text{°}C$
T_{amb}	Ambient air temperature	$\text{°}C$

Table 3.2: The input parameters for the after-cooling simulation model

calibrated such that the air temperature matches after-cooled air temperature measurements. An assumption of 40 °Celcius can be used if no measurements are available. Depending on the accuracy requirement, after-cooling can be excluded from the simulation. Post after-cooling, compressed air is usually still warmer than ambient conditions. Air temperatures underground can be accurately matched by including heat transfer for compressed air pipelines.

3.3.4 Verify the simulation model

From the review of literature in Section 2.5.1, it was determined that MAE and the coefficient of determination are the most effective methods of measuring model accuracy. Therefore, for this study, both measures are utilised in the model verification. These measures are obtained by comparing the major simulation outputs (Total system power, flow and pressure) to actual data from the system. R-squared and MAE metrics are calculated by applying the applicable methodologies discussed in Section 2.5.1.

For this study, the selected verification constraints were selected as :

$$r^2 > 0.9 \text{ and } Err\% < 5\%$$

If these limits are met for the power, flow and pressure of the system, the model is considered accurate. As an extra measure relative error of the output for the minor model components should be $> 85\%$ of the actual data. To obtain the true error of the model, instrumentation measurement error should combined with the calculated simulation error.

Additionally, periodically repeated simulations could be used to verify simulations more definitively. To perform the repeated simulation verification, the input variables should be updated for each new period. The output values of the simulation should then be compared with actual measurements. For each simulation, the verification constraints should be met.

3.3.5 Select simulation inputs

The inputs of a simulation are any parameters that do not remain static or follow the same profile in day to day operation of the system. Examples of such parameters in a compressed air simulation are:

- Surface ambient conditions
- Machine operation schedules
- Air demands
- Operational changes

Changing the simulation baseline period for a calibrated simulation should only require the updating of the input parameters. Figure 3.11 shows an example of a changing compressor schedule where an input parameter would need to be updated in the simulation.

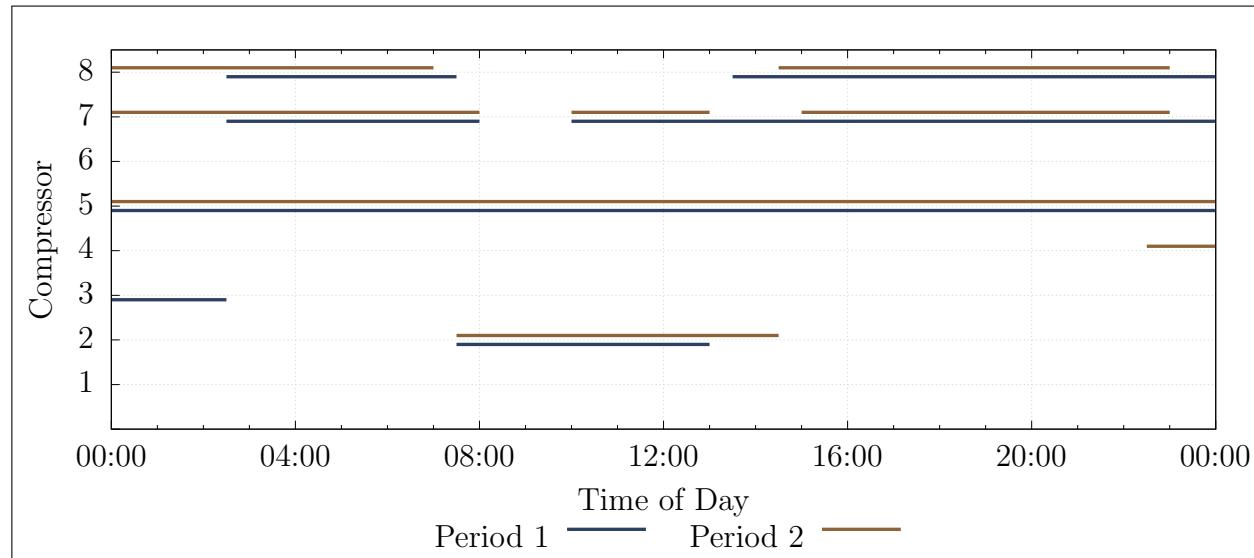


Figure 3.11: An example of two baseline periods, showing a changed compressor schedule

3.3.6 Periodic simulation

Periodic simulation refers to the repetition of simulations over subsequent periods to determine the sequential accuracy of the model. Sequential accuracy is important to verify the model is valid in general and not just a single period. This simulation will also indicate where operational changes have occurred as the simulation accuracy will be reduced.

The following process, as illustrated in Figure 3.12, was followed to implement periodic simulation:

- Collect simulation input data periodically for each simulation interval
- Import input data series into simulation model
- Execute/solve the simulation
- Export the output simulation values
- Compare output data with the systems actual operation and identify major discrepancies
- Trigger the process periodically

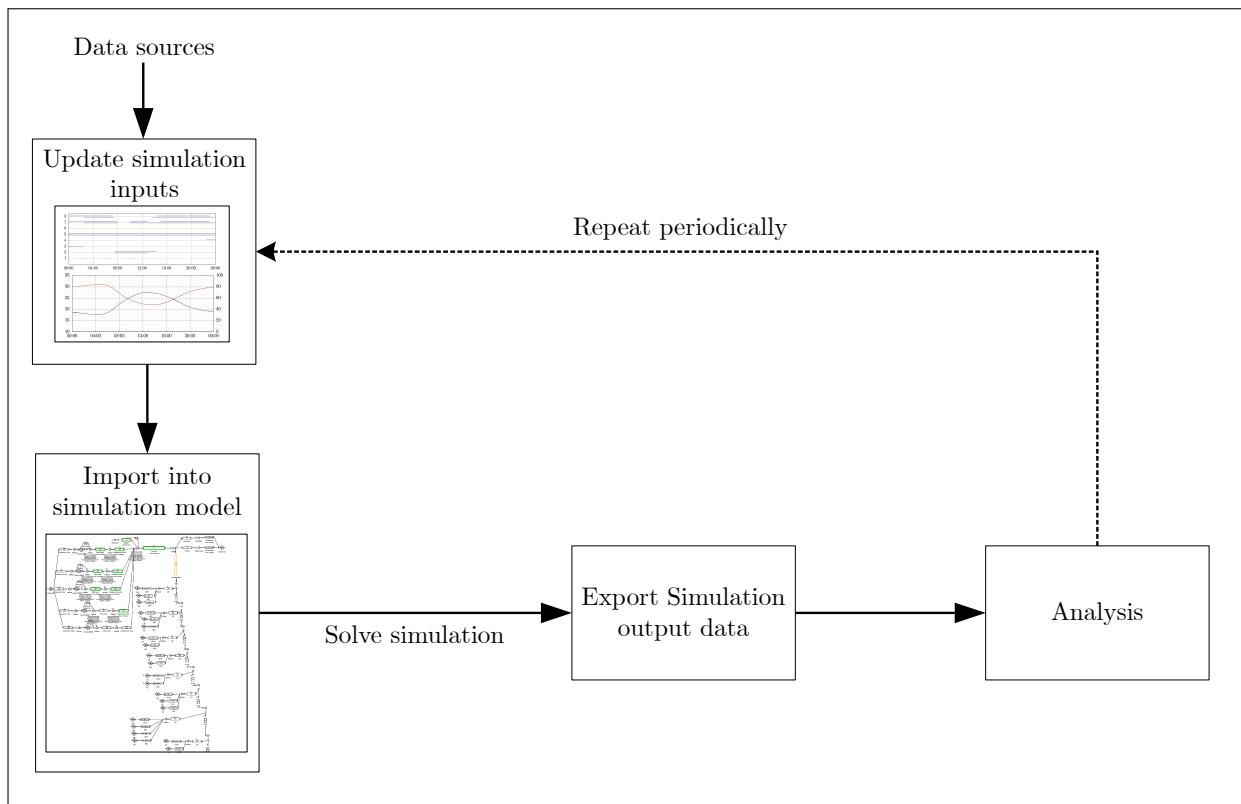


Figure 3.12: The periodic simulation process that was followed in this analysis

3.3.7 Summary

In this section, the subprocesses required for the simulation model development and verification were discussed. First, a method for selection of the model boundaries and parameters were discussed. A discussion on modelling procedure for compressed air sub components then followed. The verification process and the selection of accuracy limits were then discussed concerning the analysis performed in section Section 2.5.1. The simulation input selection procedure was then reviewed. Finally, a procedure for repeated/periodic simulation was provided.

3.4 Implement the simulation

3.4.1 Preamble

Once a simulation has been developed and verified, the implementation of interventions and scenarios then follows. In this section, the approach of implementing the simulation methodology, and analysis of interventions will be discussed.

3.4.2 Execute simulation scenarios

At this point, the simulation model has been verified using historical data. The verified output data series is now used as a baseline upon which interventions can be quantified.

The simulation inputs of the model are now adjusted to create the desired scenario. For example to create a scenario where a specific compressor is shut down over a period, the input schedule of the compressor is adjusted in the simulation model.

The simulation is then executed, and this process is repeated for each of the scenarios. For each scenario, the desired output parameters must be exported for further analysis.

3.4.3 Quantify operational benefit

With the data for each of the simulated scenarios exported, the relative improvement compared with the baseline should now be quantified. This comparison is achieved by analysing the differences between the baseline and optimised data series as shown visually in Figure 3.13.

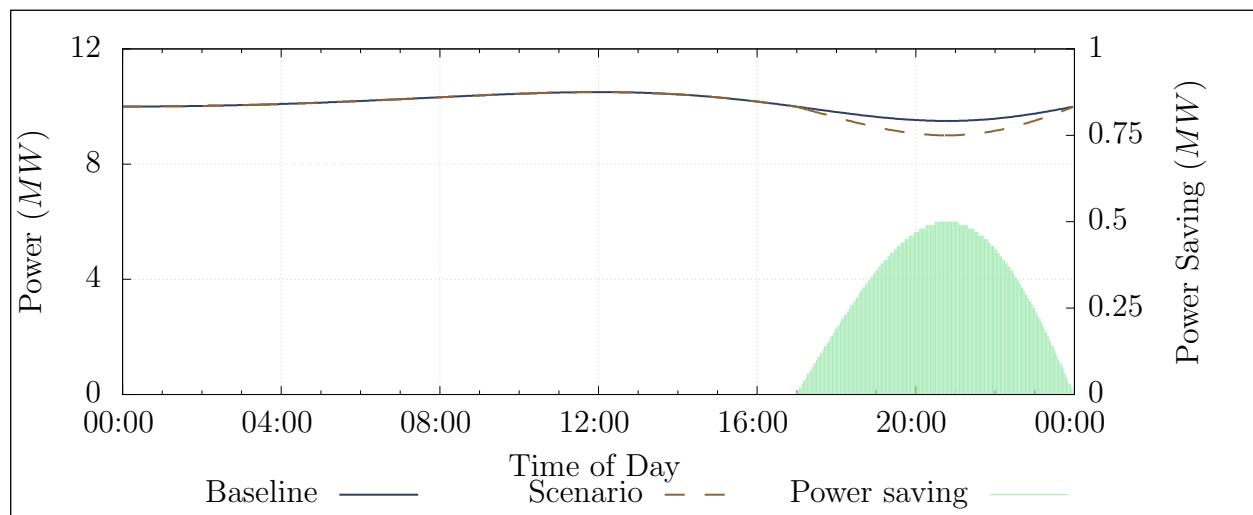
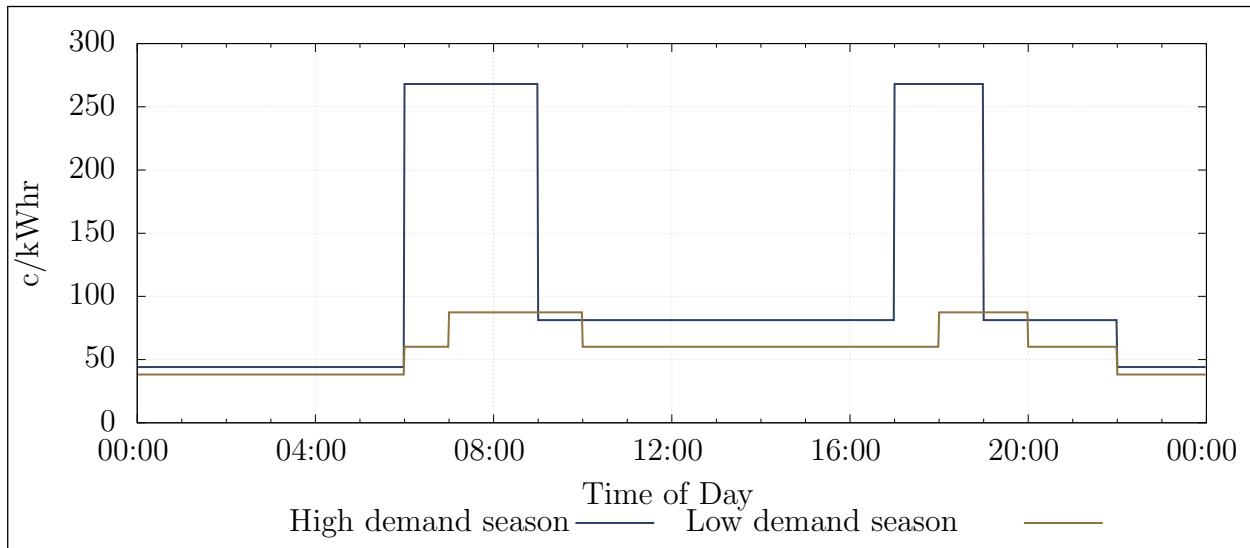


Figure 3.13: An example of a baseline vs. optimised power comparison

For power data, the expected annual energy cost saving can be calculated using average weekday energy saving and the tariff structure provided by Eskom. The energy supplier's weekday Time Of Use (TOU) tariffs for both the high demand (Jun - Aug) and low demand (Sep - May) seasons are shown in Figure 3.14.

Estimating the cost benefit for improvements in pressure delivery is harder to quantify. Instead, the average pressure benefit for a period should be provided in kPa. For example, “The simulation indicated an x MW saving with an additional pressure improvement of y kPa during the drilling shift”.

¹ Eskom, "2017/18 Tariffs and charges" [Online] http://www.eskom.co.za/CustomerCare/TariffsAndCharges/Pages/Tariffs_And_Charges.aspx, [Accessed 28 June 2017]

Figure 3.14: Eskom's weekday TOU tariff structure.¹

3.4.4 Report results to the mine

Once the benefits for each simulated scenario has been calculated and quantified, the interventions should be prioritised in the order of the greatest benefit for the mine. The implementation costs and pay back periods of the interventions can also be considered in this process.

The results and recommendations should be submitted to responsible mine personnel in the form of a report, an example report is shown in Figure 3.15. At this point, the process of implementation becomes the mine's responsibility. The mine may require further validation of the results through practical testing.

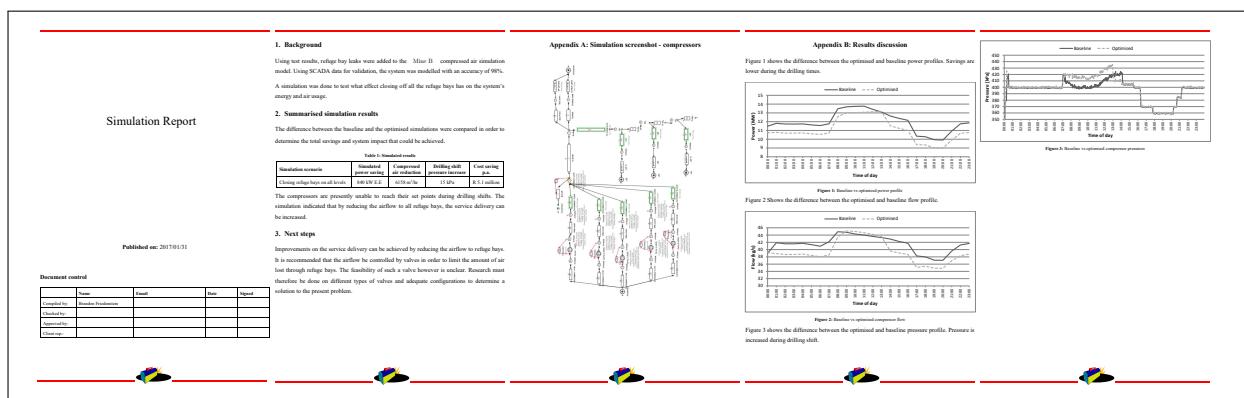


Figure 3.15: An example of a simulation report that is submitted to the mine personnel

3.4.5 Summary

This section discussed the of implantation of the simulation procedure. Implementation involves execution of the simulation scenarios, followed by the numerical calculation and

quantification of energy cost savings and other benefits. Finally, the procedure to report findings to the mine is given.

3.5 Conclusion

The aim of chapter 3 was to provide a methodology to develop compressed air simulations. The method was broken into three steps:

1. Investigate the system
2. Develop and verify a simulation model
3. Execute the simulated scenarios and quantify the benefits

This investigation step involved obtaining and verifying data and information regarding the compressed air network. Processes to resolve scenarios where data can not be obtained were also provided.

In the next step, a simulation model development and verification procedure were provided. This procedure also described the selection of model and simulation parameters, the development of subcomponent models as well as a verification procedure. A methodology for repeated/periodic simulation was also provided.

The final step involved the execution of the simulation followed by the methods to calculate, quantify and report potential benefits of the simulated scenarios.

CHAPTER 4

Results and validation

“It doesn’t matter how beautiful your theory is, it doesn’t matter how smart you are. If it doesn’t agree with experiment, it’s wrong.” - Richard P. Feynman

4.1 Introduction

This section will validate the developed simulation methodology through case studies. Mining compressed air systems were chosen as case studies based on the availability of data and the scope for energy and operational improvement. Two different mines were selected for the studies.

Three case studies were performed. In case study 1 and 2, improvements were simulated on mine A and mine B respectively. In Case study three periodic simulation analysis is implemented using the simulation developed for case study 2. From the results of the case studies, the potential benefits compressed air simulations for the South Africa mining industry is estimated and discussed.

4.2 Case study 1. Simulated improvements on mine A

4.2.1 System investigation

Mine A represents a group of three gold mining shafts and a gold processing plant in the Free state province. The mine shafts and gold plant share a compressed network. Before this study, efforts had been made to optimise the system through DSM energy projects. However, there may still be potential for further optimisation. An investigation was performed to gather data and understanding of the system and to identify potential energy and operation improvement strategies.

An air-flow distribution layout was developed for the system; a simplified layout is shown in Figure 4.1. From the schematic along with information and data, an understanding of the air network's operation was obtained. The system typically utilises 5 MW of instantaneous power. During the drilling shift, the demand increases to 6 MW. Figure 4.2 shows the average weekday power profile between January and May 2016.

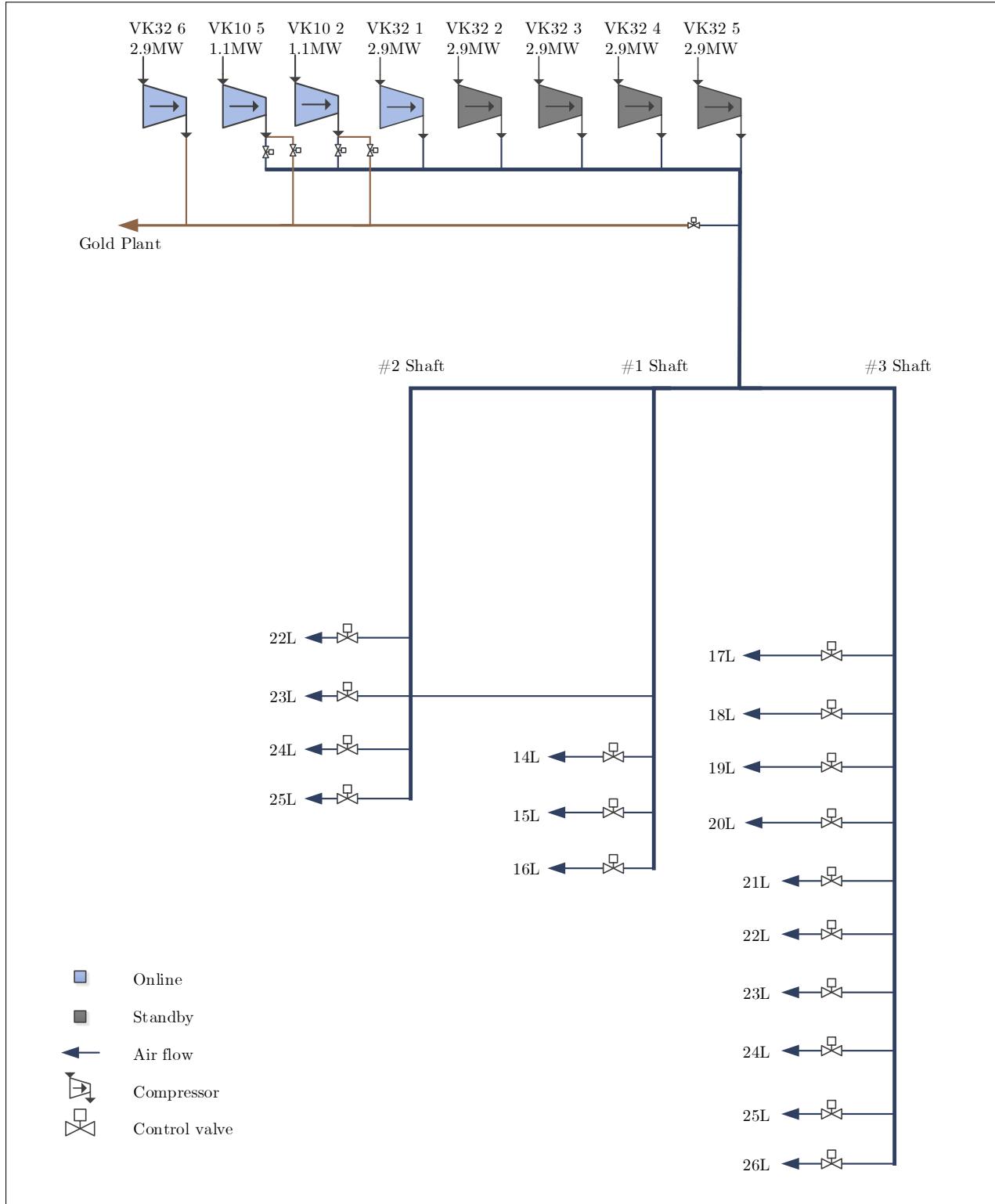


Figure 4.1: Simplified process flow chart of the compressed air network

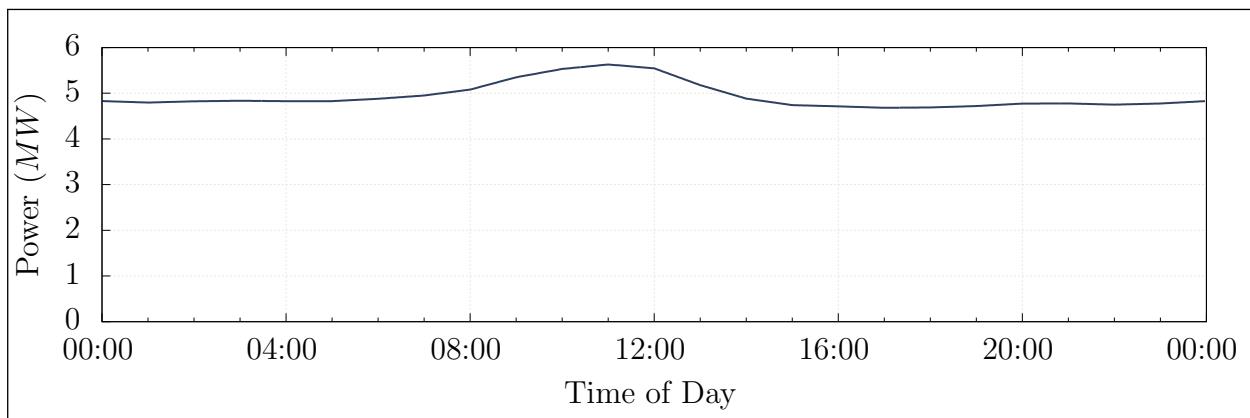


Figure 4.2: Average power profile

Seven compressors are available in the system. Five large compressors (VK32) rated at 2.9 MW each and two smaller compressors (VK10) with a power rating of 1.1 MW each. No more than four compressors are required at any time; the other compressors are therefore on standby. Air is supplied to the sections in three mining shafts as well as a gold processing plant on the surface.

The mine normally operates the compressors with a constant pressure setpoint of 500 kPa . The setpoint is kept high for the gold processing plant, which requires constant high pressure throughout the day. This constraint makes it difficult to reduce the setpoints of the compressors. It is possible to control the air supply pressure to the gold plant independently from the rest of the network. The supply to the gold plant is controlled by the surface valves.

The evening Eskom energy peak time was identified as a period where savings could be obtained. During this time, air is not required underground as blasting is scheduled. Due to the energy tariff structure, interventions during the energy peak also maximise the financial benefit for the mine. Compressor setpoint control and underground valve control were identified as strategies to achieve these savings. Due to the risk of loss of production, the mine would not allow practically testing the scenario on the actual system. The simulation was therefore required to accurately calculate and analyse the benefits of the compressor set points. A model was developed to test these scenarios.

4.2.2 Model development

With the data and understanding gathered from the investigation, a model was developed using the PTB software tool and the methodology discussed in Chapter 3. First, the simulation boundary was selected to include the measured flows to each level underground, as well as the surface processing plant. For highest accuracy, the simulation step size was set to 30

minutes to match the available data resolution.

The simulation component models were developed and calibrated using the respective methods discussed in chapter 3. The following assumptions were made to simplify the model development:

- The effect of compressed air after-cooling is negligible
- Heat transfer over the pipe length is negligible
- Typical underground air conditions
- Surface ambient air conditions followed normal summer trends

The model components were calibrated so that the simulated outputs matched data from the real system. The process flow diagram for the simulation is shown in Figure I.1(Appendix I). The model data inputs and outputs are described in Table 4.1.

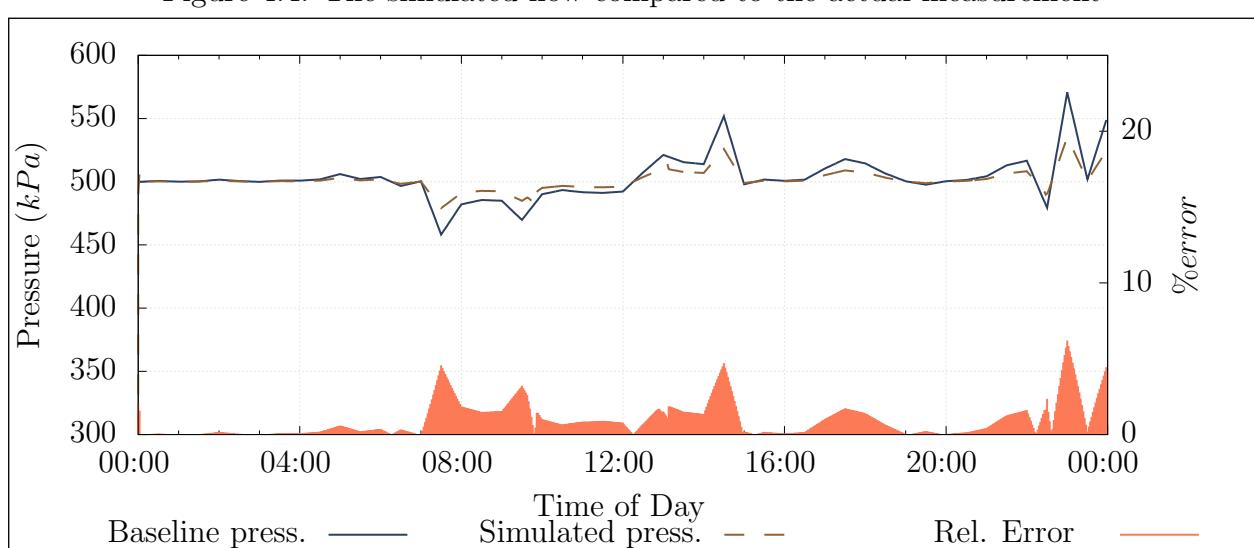
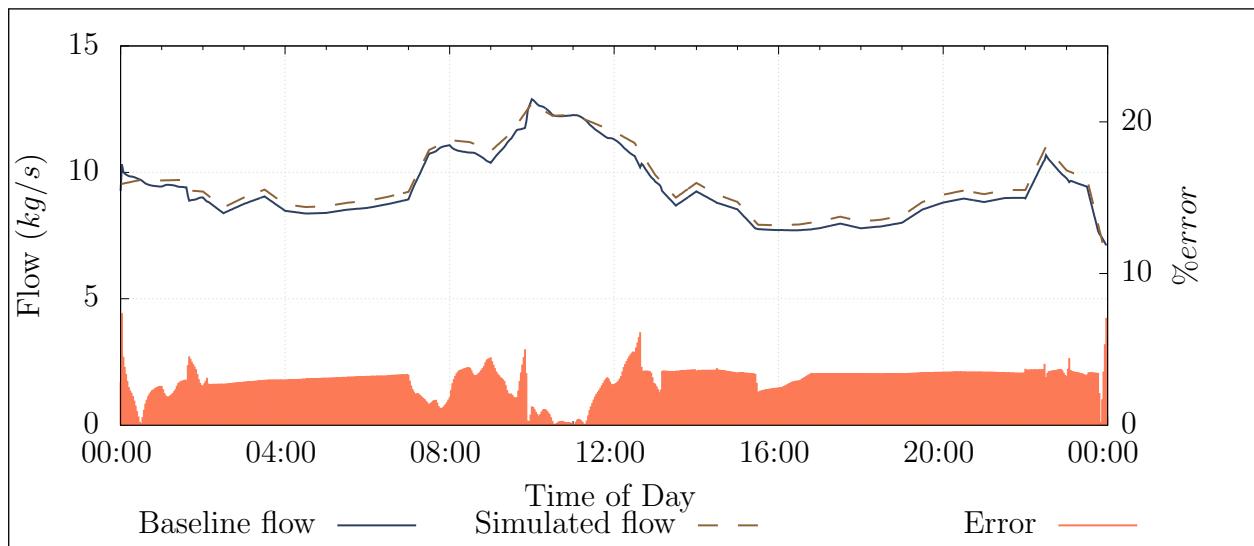
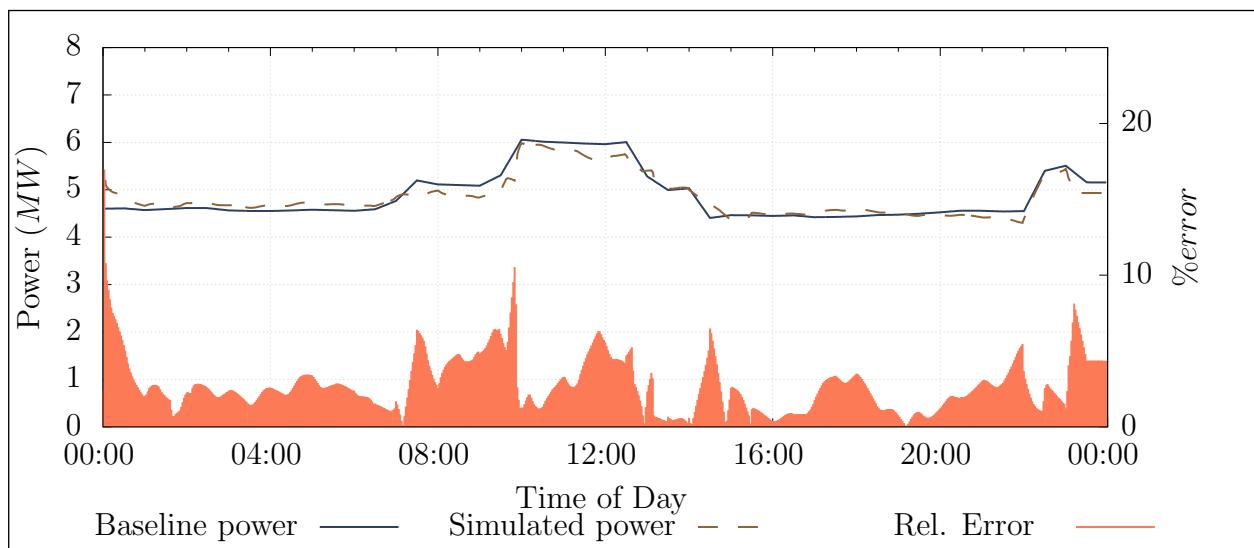
Inputs	Outputs
Level measured flows	Compressor powers
Compressor schedules	Network flows
Compressor setpoints	Network pressures
Underground valve setpoints	

Table 4.1: Data inputs and outputs for the Case study 1 simulation model

4.2.3 Verification of the simulation model

Verification was performed to ensure that simulated output accuracy was $> 95\%$. Figure 4.3, Figure 4.4 and Figure 4.5 show the total simulated power, flow and outlet pressure of the compressors compared to the system. The calculated error metrics for the key process parameters are shown in Table 4.2. The average relative percentage error for the total power and flow was 2.78% and 3.0% respectively. The $Err\%$ of the outlet pressure was 0. All the process parameters were well within the target relative error of $\pm 5\%$ percentage error and $r^2 \geq 0.9$.

The accuracy of the simulation was checked in more detail to ensure that the sub-component outputs matched the actual measurement with high accuracy. Table II.1(Appendix II) shows the precision each measured simulation output in the model.



Verification method	Result	<i>Err%</i>
Total Flow		
Residual Difference	0.25 kg/s	2.73%
MAE	0.27 kg/s error	3.0%
Coefficient of determination	$r^2 = 0.99$	-
Total power		
Residual Difference	0.02 MW	0.57%
MAE	0.14 MW error	2.78%
Coefficient of determination	$r^2 = 0.91$	-
Compressor outlet pressure		
Residual Difference	0.03 kPa	0.01%
MAE	0.21 kPa error	0.03%
Coefficient of determination	$r^2 = 0.99$	-

Table 4.2: Case study 1: Verification of simulation model

4.2.4 Execution of simulation scenarios

Scenario 1. Compressor set points

The Eskom evening peak tariff time occurs during the blasting shift. During this time, the pressure requirements underground are lower than the rest of the day. Reducing pressure in the network reduces power as less work is required from the compressors. Additionally, losses caused by air leaks are reduced. However, lowering the pressure setpoint of the compressors requires independently controlling the air to the gold plant.

The compressor set-points were reduced for the simulation model to 420 kPa , the minimum allowed compressor set-point during the drilling shift. The compressor schedule was changed to allow independent control of the gold plant pressure. Gold plant pressure was maintained at 490 kPa .

The results of the simulation, shown, in Figure 4.6, indicated an average power reduction of 0.46MW Peak-Clip (P.C.). This energy optimisation relates to R 0.37m Per Annum (p.a) energy cost saving to the mine.

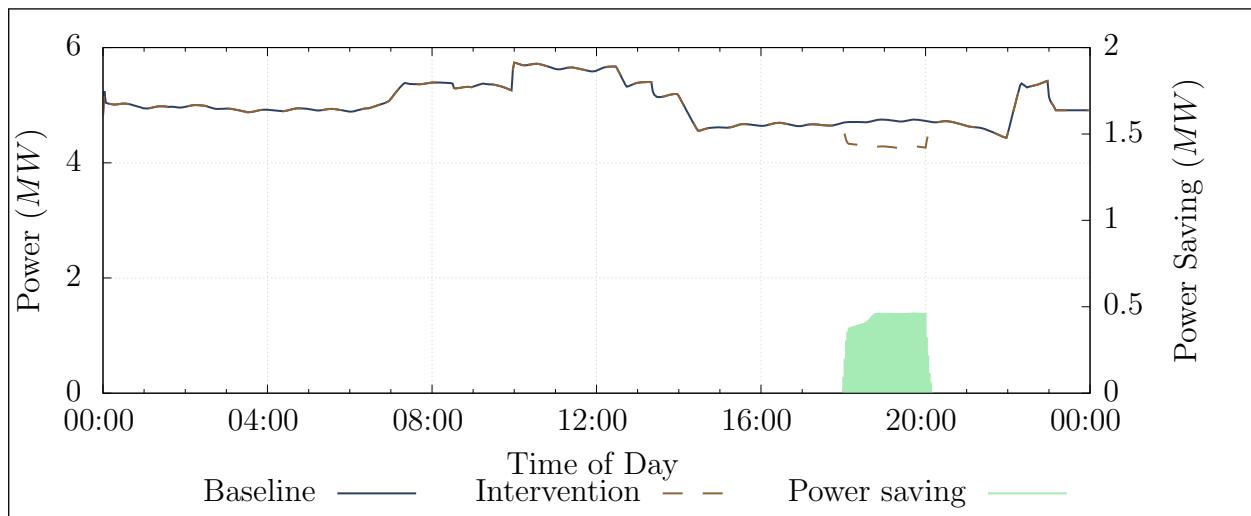


Figure 4.6: Energy savings by reducing compressor setpoints

Scenario 2. Control valves set points

An alternative scenario is to reduce the pressure at the control valves at each level. By reducing pressure at the control, setpoints can be lowered to the minimum requirement per level. This reduction can lead to higher savings than could be achieved through compressor setpoint reduction. This scenario would be relatively easy to implement as it does not require any changes to the compressor control schedule.

Air pressure setpoints were reduced to 300 kPa at the underground control valves during the evening peak period. Analysis of the simulation results showed a 1MW average P.C. saving. The intervention would lead to an annual cost saving of R 0.91m.

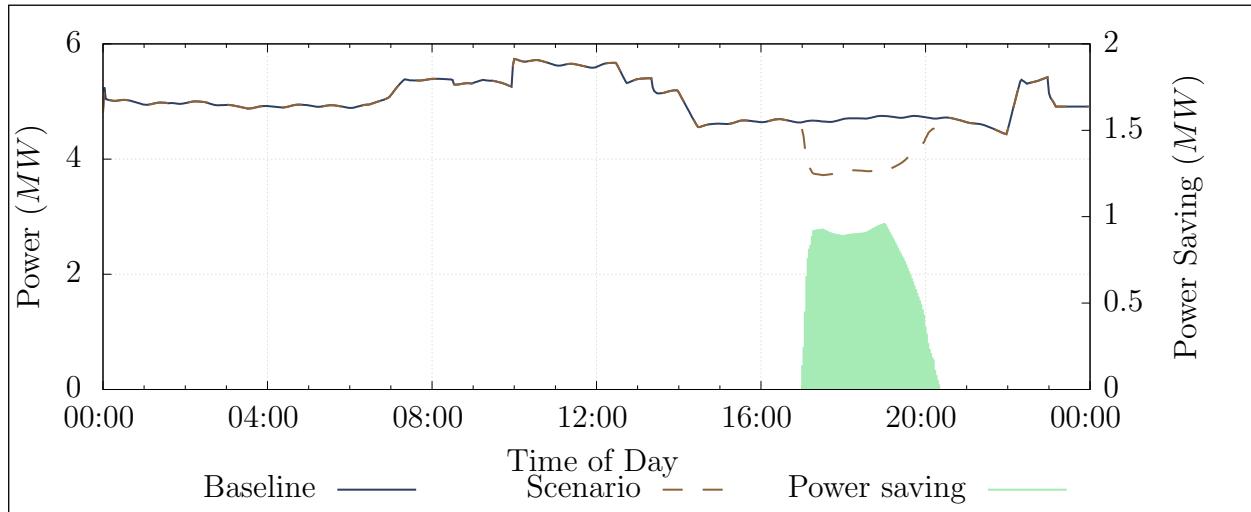


Figure 4.7: Scenario 2 simulated peak time power reduction

4.2.5 Comparison of scenario results

Comparing the scenarios in Table 4.3 showed that Scenario 2 had a larger peak energy impact than scenario 1. Further savings could be achieved through a combination of the two scenarios as well as investigating setpoint reductions during other periods of the day.

Scenario	Power saving	Cost saving p.a
Scenario 1 results		
Reducing compressor setpoints	0.46 MW P.C.	R 0.37m
Scenario 2 results		
Reducing underground pressure during evening peak	1.0 MW P.C.	R 0.91m

Table 4.3: Comparison of Mine A's simulated scenarios

4.2.6 Validation of results

Scenario 2 was implemented on the actual compressed air system. An energy saving of just under 1 MW P.C. was recorded when compared with the 2016 power baseline profile. These results matched the simulated scenario closely. Figure 4.8 shows the practical result compared with the simulated and baseline power profiles. The average evening peak power clip achieved on the actual system was 0.95 MW. This energy improvement was within 5% error of the simulated result.

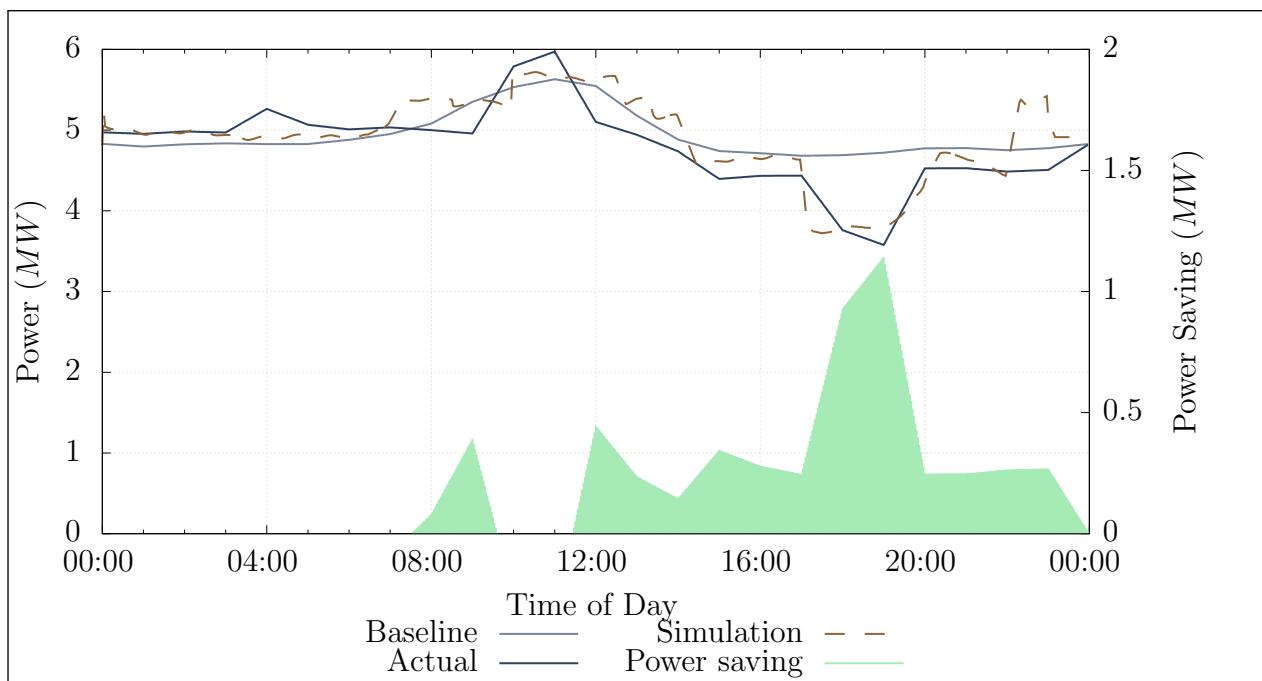


Figure 4.8: Actual power savings achieved on the system

4.2.7 Summary

A case study was implemented on a mine compressed air network in the Freestate. Following the simulation methodology. An investigation was performed to gather data and identify potential interventions. A simulation model was developed to test scenarios. The tested interventions showed a P.C. saving of 1 MW which would result in a cost saving of R 0.9m. The simulation was validated with results from implementation on the actual system. The.

4.3 Case study 2: Simulated improvements on mine B

4.3.1 System investigation

Case study B was performed on a large South African gold mine. The mine utilises five compressors supply compressed air to various surface and underground operations. An investigation was carried out to gather the data and information required to build a simulation model of the network as well as to identify potential cost-saving simulation scenarios.

A basic air distribution layout was developed for the system. Figure 4.9 Illustrates the system process in detail, indicating the flow distribution to the three mining shafts and gold processing plant.

Data related to the mines scheduling as well as critical limits and setpoints of the compressed air system was obtained from various mine personnel. From this, a general understanding of the operation was obtained. Critical data parameters such as Power, pressures and flows of the system were gathered from the SCADA as well other data measurement sources. This information will be used to develop and calibrate the simulation model.

Strategic level investigations were performed on the significant mining levels to map and measure the locations and air usage for the cross-sections, refuge bays, major leaks and other compressed air consumers on each level. An example of a resultant schematic from the underground investigation is shown in Figure III.1. The information gathered from the system investigations was then utilised to develop and calibrate a simulation model.

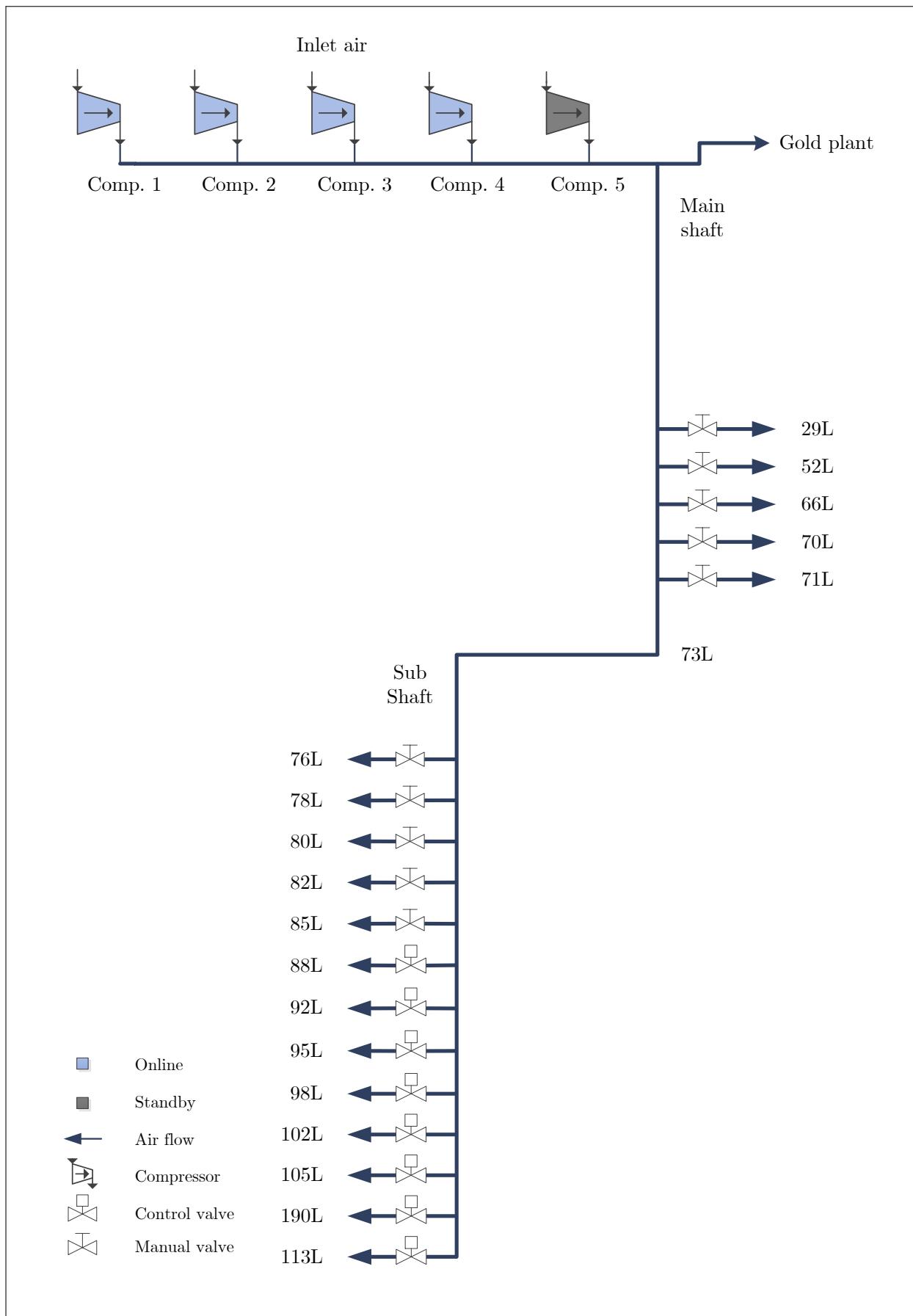


Figure 4.9: Process schematic of the compressed air network

4.3.2 Model development

From the investigation, the compressed air network was modelled in PTB. The methodology described in Chapter 3 was utilised in this process. The following assumptions made to simplify the model development:

- After cooling reduced compressed air temperature from 100° Celcius to 40° Celcius
- Typical underground air conditions

The boundaries of the baseline simulation were selected based on the available data for the system. The developed simulation model is shown in Figure I.2. For maximum accuracy, the simulation step size was set to the two minutes to match the resolution available from the data source. The model components were calibrated so that the simulated outputs matched data from the real system. The process flow diagram for the simulation is shown in Figure I.2. The model data inputs and outputs are described in Table 4.1.

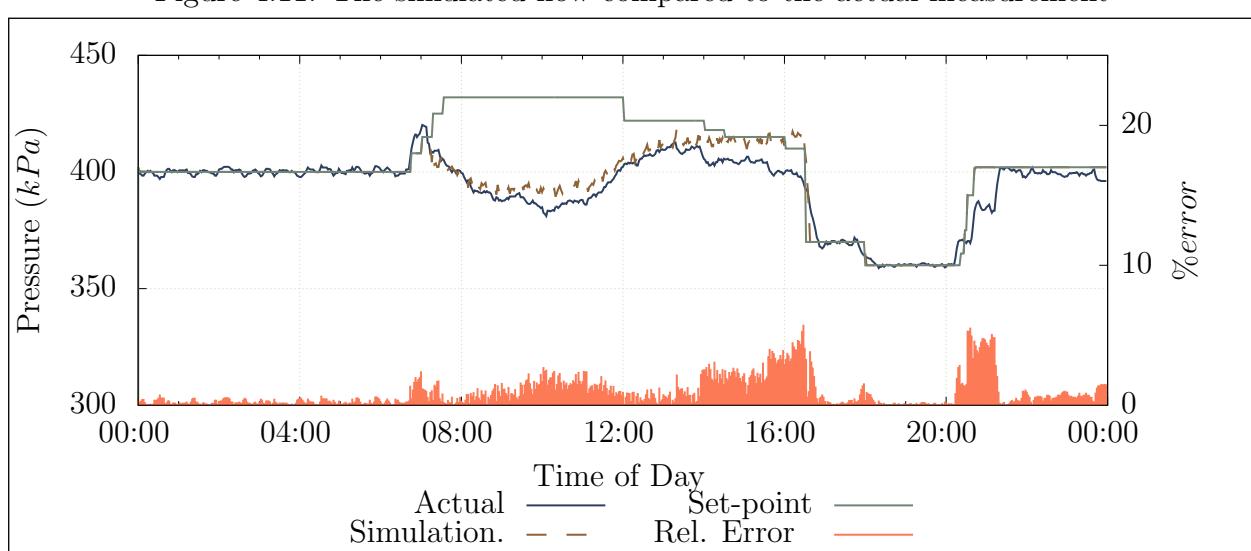
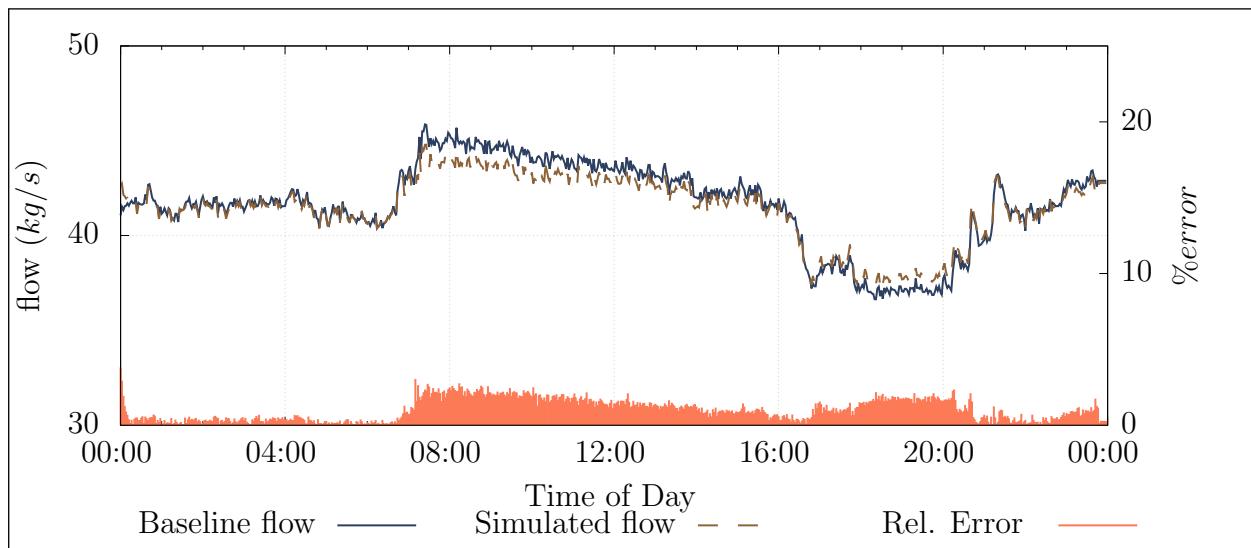
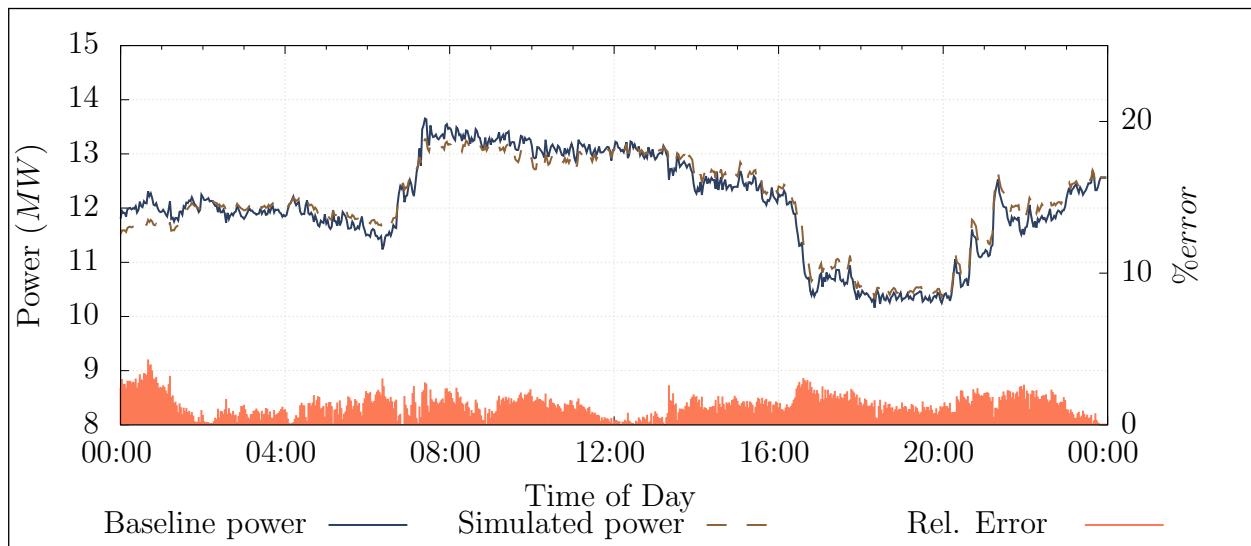
Inputs	Outputs
Level measured flows	Compressor powers
Compressor schedules	Network flows
Set-points	Network pressures

Table 4.4: Simulation inputs and outputs

Verification of the simulation model

Using the verification methodology, the simulation model was verified by comparing the simulation outputs to actual measured values. The compressor's outlet pressure often does not match the set-point. The measured outlet pressure was used as set points for the compressors to verify the power and flow outputs. The setpoints ensured that the pressure in the network is identical to that of the actual.

With the simulated network pressure almost identical to the actual, the power and air-flow outputs were compared with outputs from the physical system. Figure 4.10 and Figure 4.11 show the comparison of the total power and flow of the system with the physically measured values for that same period. The relative error of these process parameters compared to the real network was 1.02% and 1.36% respectively. This simulation error was within the acceptable error limits. A summary of the verification metric for the major process totals is provided in Table 4.5.



Once the power and flow parameters were verified with an acceptable error, the actual pressure set-point profile was imported to the compressor controllers. The simulated outlet pressure was then compared to the actual measured pressure and setpoint. This comparison is shown in Figure 4.12. The error of the compressor outlet pressure was acceptable at 0.98%. The measured flows for all measured subcomponents were independently verified to ensure system accuracy. A comparison between simulation outputs and physical measurements is shown in Table II.3(Appendix II).

Verification method	Result	<i>Err%</i>
Total Flow		
Residual Difference	0.22 kg/s	0.53%
MAE	0.43 kg/s error	1.02%
Coefficient of determination	$r^2 = 0.99$	-
Total power		
Residual Difference	0.05 MW	0.39%
MAE	0.16 MW error	1.36%
Coefficient of determination	$r^2 = 0.91$	-
Compressor outlet pressure		
Residual Difference	2.79 kPa	0.71%
MAE	3.85 kPa error	0.98%
Coefficient of determination	$r^2 = 0.890$	-

Table 4.5: Verification of simulation model

4.3.3 Scenario 1. Refuge bay optimisation

After an underground investigation, unnecessary refuge bay leaks were identified as a significant inefficiency that can be reduced. A test on a single mining level was performed to measure the potential flow saving of reducing refuge bay leaks. The test showed that by reducing refuge bay leaks, by closing the valves, would lead to an average air saving of 0.05 kg/s per refuge bay at normal operational pressures. This measurement was conservative as it was not possible to close all the refuge bays on the level for the test.

Due to the size of the mine, extending these tests to include the rest of the mining sections was not practical. Therefore, the benefits of an intervention on the entire mine could not accurately be determined from practical tests. Using simulation the typical operation with can be accurately compared with the intervention scenario to quantify the potential financial and operational benefits throughout a given period.

The simulation model boundaries were updated to include refuge bay leaks on each level. For each refuge chamber, an air leak was added to the model by utilising per level layouts indicating locations of refuge bays. These leaks were modelled as flow demands using the data from the initial refuge bay tests. The overall mass flow of the system was maintained to ensure model accuracy.

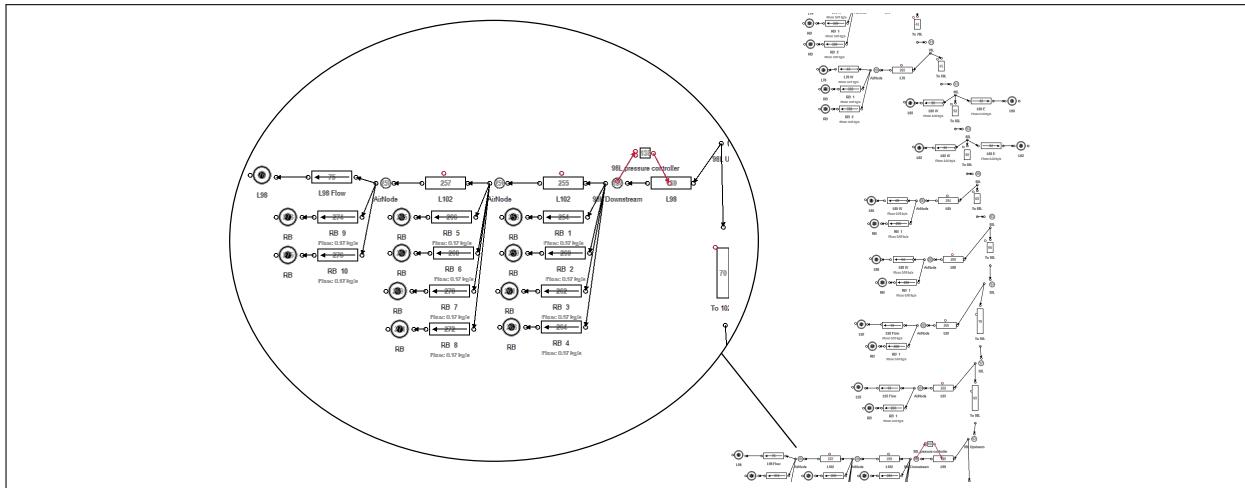


Figure 4.13: Addition of the refuge bay flow demands in the simulation model.

By adding the flow components in the same location in the actual network, the pressure in each chamber is correctly modelled. The full simulation model schematic is shown in Figure 4.13. The updated simulation was re-checked with actual data from the network. This model was then used as a baseline to quantify saving for the scenario.

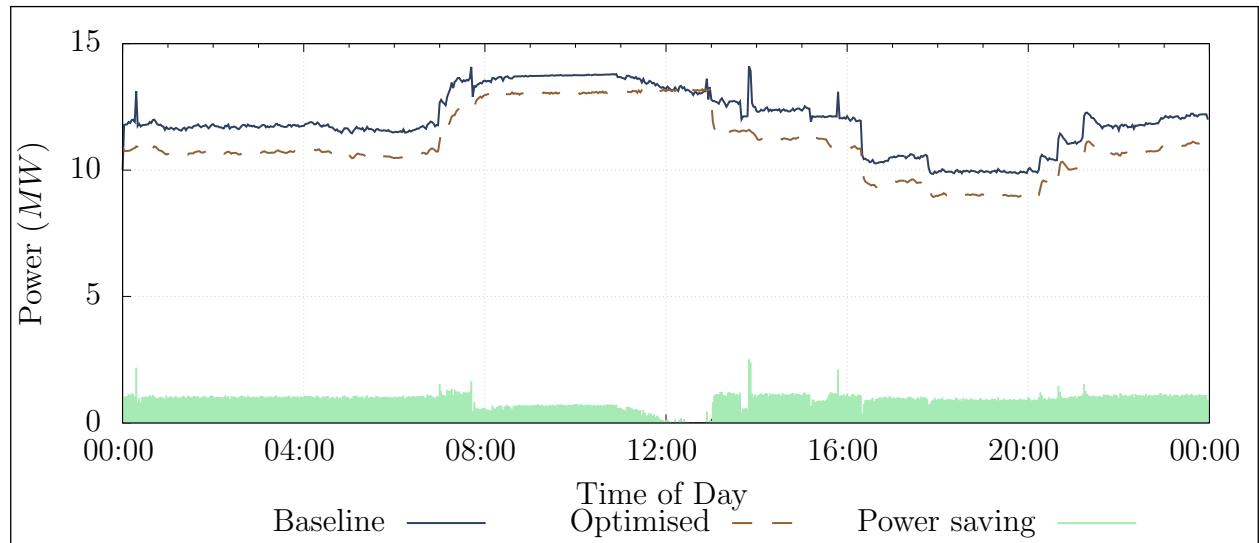


Figure 4.14: The baseline system power compared to the system power when refuge bay leaks are reduced

Recreating the optimised scenario was modelled by setting the refuge bay flow components to 0 kg/s . The simulation was performed and the output data compared to the baseline.

Figure 4.14 shows the baseline power compared to the optimised scenario. The comparison showed a potential 0.92 MW improvement in Energy Efficiency (E.E.) through optimisation of refuge bay leaks. The optimised scenario would lead to R 5.13m in energy cost saving for the mine.

An additional pressure benefit was identified during the drilling shift, shown in Figure 4.15. The reduced flow lead to an average pressure increase of about 15 kPa during the drilling period. The pressure increase could lead to a general improvement in drilling efficiency and potentially increase the rate of production. This improvement may have a greater benefit to the mine than the energy cost savings.

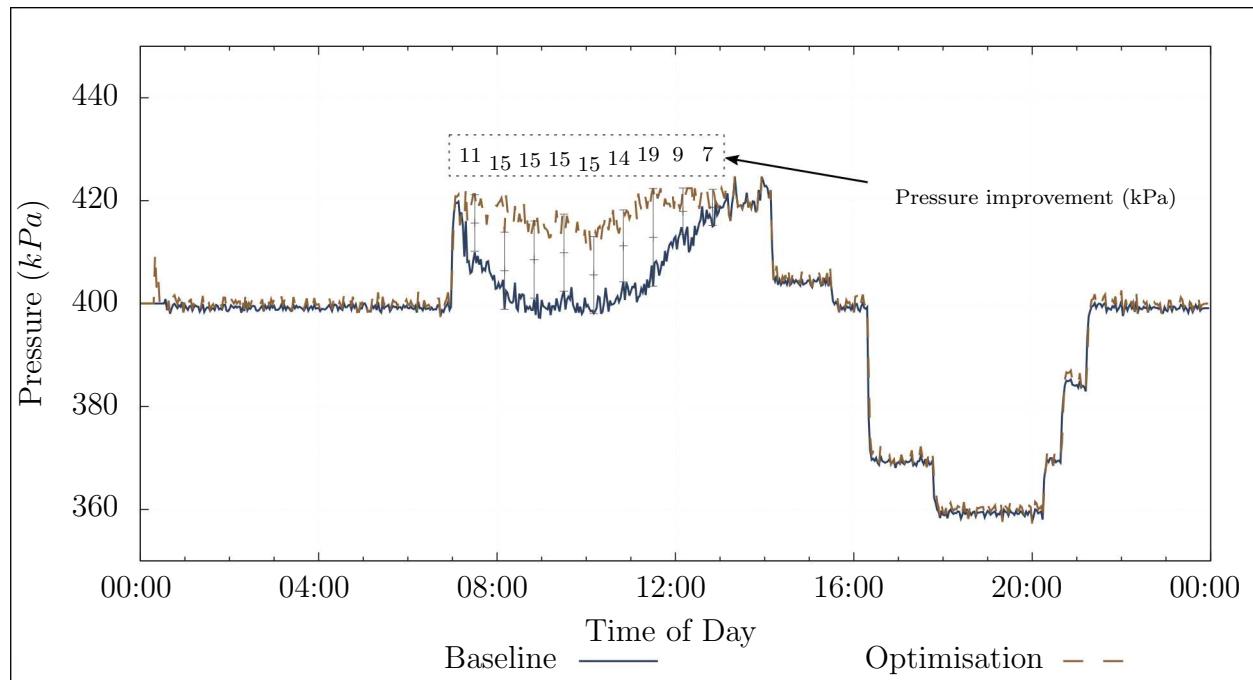


Figure 4.15: The baseline system pressure relative to the system pressure when refuge bay leaks are reduced

4.3.4 Scenario 2. Closing off levels and inactive work areas

Scenario background

The largest impact on energy costs can be achieved during the energy high demand periods. At mine B, blasting coincides with the evening energy demand peak. Typically, the air requirement during this period is lowered. However, due to compressed air misuse, leaks and open valves, significant amounts of air is still used during this time. Reducing pressure to areas during these times may lead to a major power and cost saving.

Closing off valves at inactive work areas and reducing station pressure were identified as two

strategies to reduce airflow during the evening peak. The station cannot be closed completely as some services require the air supply at all times.

Simulations were performed to identify the most suitable strategy and to quantify the effect of each on a single mining level. The level was modelled to include all the main leaks, refuge bays and drilling sections.

Underground investigation

105L is the most active production level at Mine B. An investigation was therefore performed to identify all the air users, leaks and inefficiencies at the level. A walkthrough manual inspection of the level was performed. Air users, leaks and inefficiencies were identified, quantified and mapped; the resultant schematic is provided in Appendix III. The results were then used to develop a detailed model of the level in PTB.

The simulation process schematic of the level is shown in Figure 4.16. This model was then used to simulate various station and working area air optimisation scenarios. When compared to the baseline the effect of each scenario for a single mining level can be determined. The results may then be used to estimate the benefit for a generalised control philosophy applied to all active mining levels.

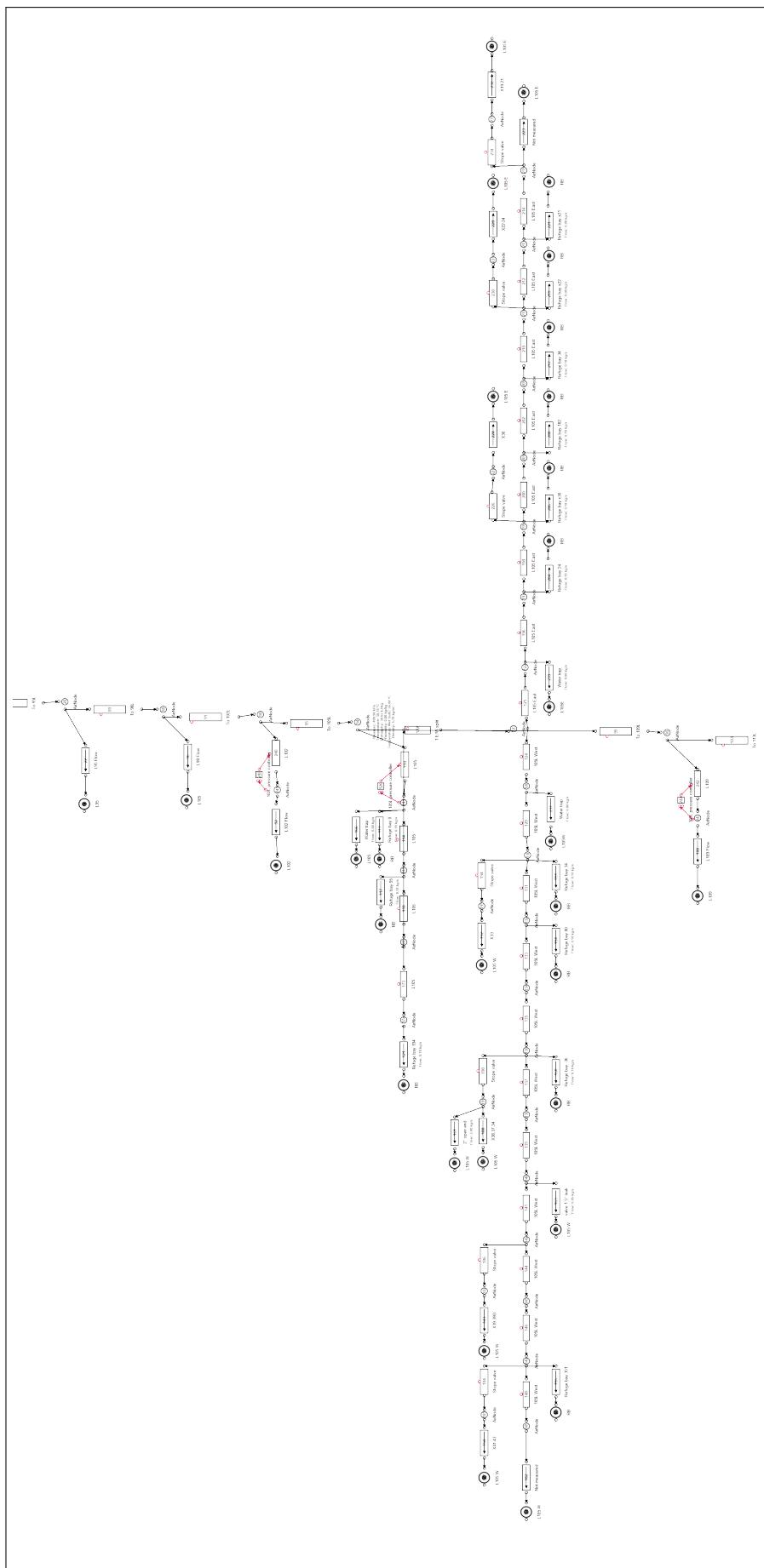


Figure 4.16: Underground level simulation layout

Simulated results

Station control, in-stope control and a combination were simulated were all simulated for 105L. Station control means control of the pressure of at the station of the level. In-stope control is control of the is cut the airflow to the mining section during certain periods. Figure 4.17 shows the effect the various interventions have on the flow for 105L compared to the baseline.

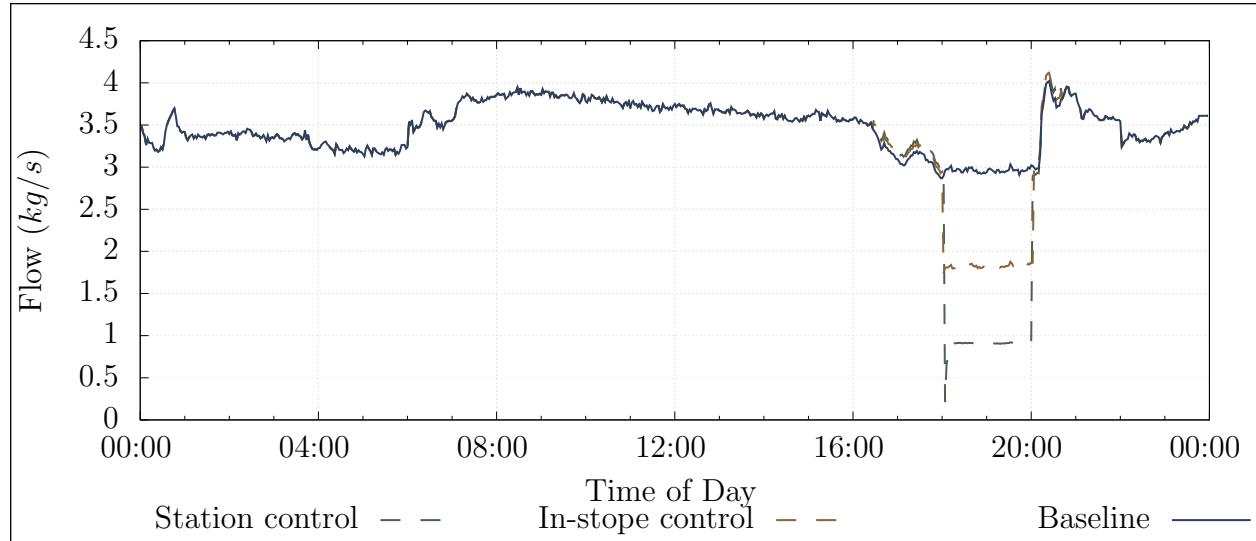


Figure 4.17: Flow reduction during blasting period for 105 level

Station control had the largest impact on the flow usage for the level. This result reflects the power reduction achieved by each intervention shown in Figure 4.18. The impact was 0.4 and 0.7 MW for the stope and station intervention respectively.

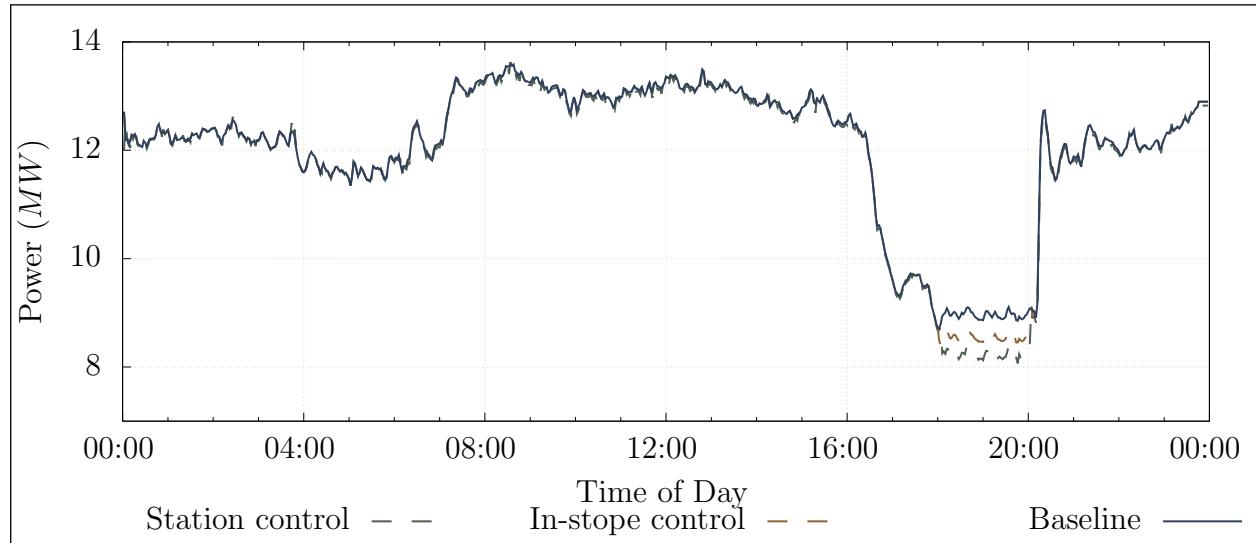


Figure 4.18: Comparing simulated flow interventions on 105L

A simulation was done to estimate the potential savings of extending the evening station control to other levels. The flow demands for levels 95—115 were updated to match the flow saving achieved in the 105L simulation. The savings obtained in generalised station control

simulation was 2.0 MW P.C., shown in Figure 4.19. It was calculated that this would lead to an annual energy cost saving of R 2.5m.

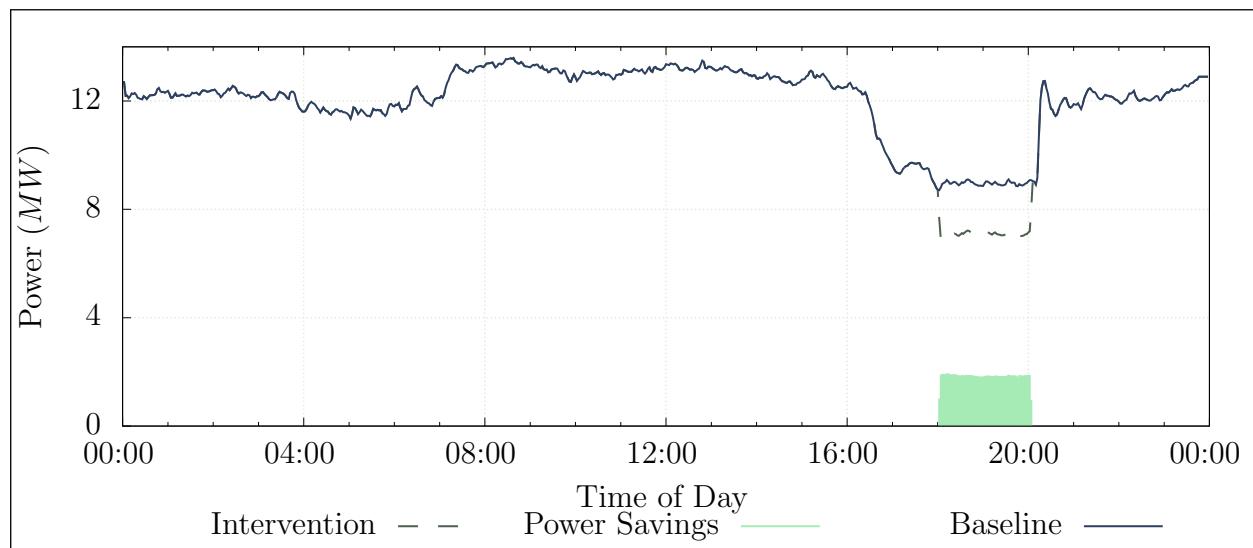


Figure 4.19: Energy saving achieved by general peak time station control

4.3.5 Comparison of interventions

The interventions were then compared to feedback to the mine. The refuge bay intervention is recommended as it will achieve the highest energy cost saving. These results were used to produce a report that was sent to the responsible personnel at Mine B.

Scenario	Power saving	Cost saving (p.a)	Additional ben- efit
Scenario 1 results			
Refuge bay leakage reduction	0.92 MW E.E.	R 5.17m	Increase in drilling pressure
Scenario 2 results			
105L peak time in-Stope control	0.4 MW P.C.	R 0.3m	-
105L peak time Station control	0.7 MW P.C.	R 2.5m	-
General peak time Station control	2.0 MW P.C.	R 2.5m	-

Table 4.6: Comparison of the simulated scenarios

4.3.6 Summary

A second case study was implemented on a mine compressed air network. An investigation was performed to gather data and identify potential interventions. This investigation included underground spot checks and measurements.

A simulation model was developed to test scenarios. Two scenarios (refuge bay optimisations and per level flow optimisations) were simulated. The simulated interventions showed an E.E. saving of up to 0.92 MW which would result in a cost saving of R 5.17m with an additional pressure improvement for scenario 1. The results for scenario 2 showed a P.C. impact of up to 2 MW with an annual energy cost saving of R 2.5m. The results of the simulations were reported to the mine.

4.4 Case study 3: Periodic simulation analysis

4.4.1 Preamble

Periodically updating the inputs of a simulation could be used to verify the compressed air model accuracy. If the precision of simulation outputs remains within constraints for subsequent days, this would indicate that the model is correctly calibrated. This analysis can therefore ensure that the model is up to date.

Additionally, this process could be used to identify significant operational changes that occur within the system. The operational changes would cause the simulation outputs to differ from the actual measured parameters. This information can be used to make improvements to the system.

A daily periodic simulation analysis was developed for the periods between 2016/11/01 and 2016/11/30 using the periodic simulation methodology discussed in Chapter 3. The simulation model that was developed in case study 2 was used for the analysis. The simulation receives the following data inputs, as shown in Table 4.7.

Inputs	Outputs
Ambient air conditions	Compressor power
Measured flows	Flows
Compressor schedules	Pressures

Table 4.7: Data inputs and outputs for the simulation

4.4.2 Results

The simulation was triggered for 24 hour periods. For each period, the data inputs shown in Table 4.7 were imported into the model, the simulation was then processed, and the outputs are compared with the real system parameters. NO human interference in the model to calibrate or modify the model was performed. Figure 4.20 shows the average daily accuracy

of the simulated total system power, flow and the shaft pressure per period.

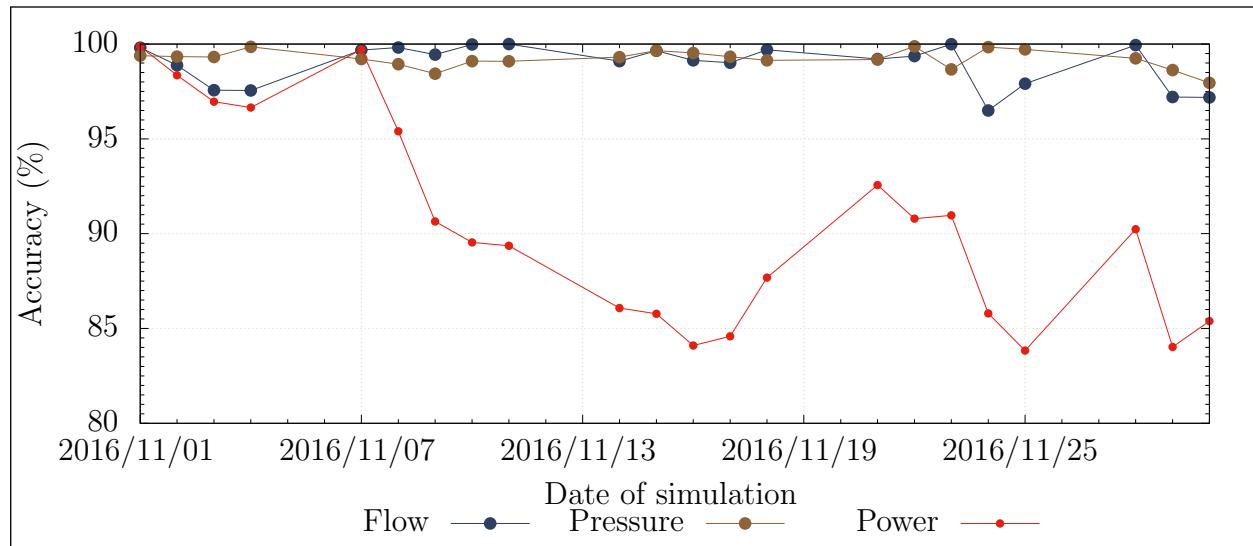


Figure 4.20: The flow, pressure and power error percentages for daily periodic simulations over a month

The accuracy of the process parameters of the simulation was within 5% for the duration of the periodic simulation. However, From the 2016/11/07, the accuracy of the simulated power dropped by between 10 and 15%. The average daily power of the system up to that point was approximately 12.5 MW. A 15% simulation error then relates to 1.9 MW difference between simulated and actual values. This error suggests a major shift in operation of the system.

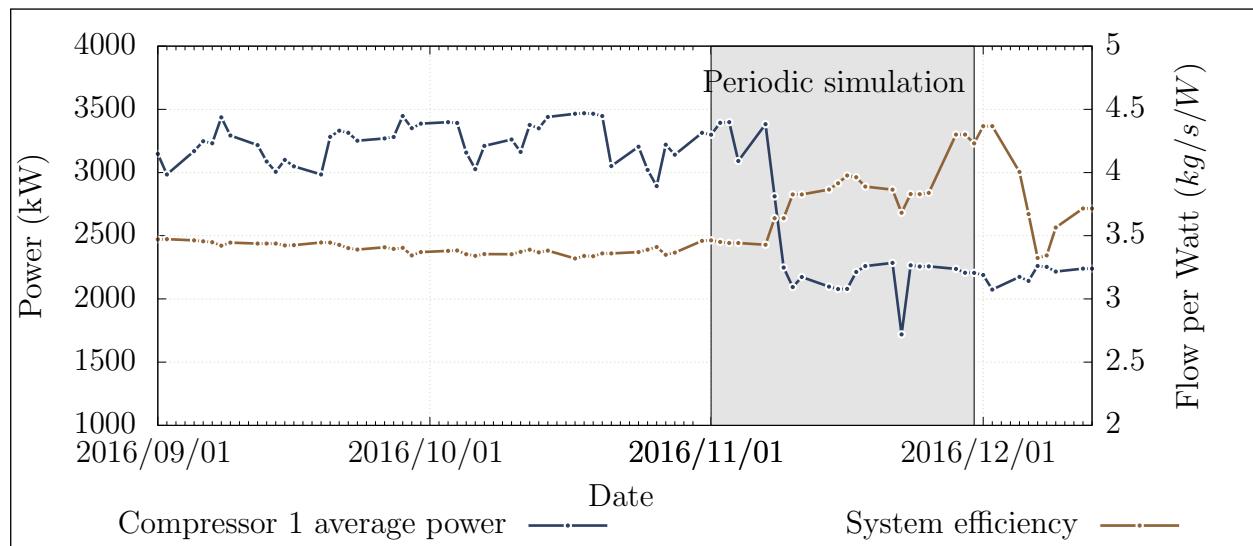


Figure 4.21: Supply efficiency and compressor 1's average power output over the time of the periodic analysis

An analysis was done to try and determine the source of the discrepancy. From the data, it was identified that the simulated power for compressor 1 was the source of the difference. Checking the actual power measurement for compressor 1 showed almost a 2 MW drop in power compared to compressor's output in the initial period. At the same time, the specific

power lowered. This drop in specific power is illustrated in Figure 4.21 which shows the average daily power for compressor 1 (blue), compared to the specific power(yellow).

A 2 MW shift in power is not likely; the results likely indicate that there was a fault in the power metering starting from 2016/11/07. A measurement error explains the perceived increase in efficiency over the same period as less power is measure then is being provided. In this situation, the simulated power measurement is a more accurate metric for compressor 1's power over this period.

4.4.3 Validation

Data from independent power meters was compared to the measurements used in the analysis. Figure 4.22 compares the measured power from the two data sources. By comparing the compressor power to the independent data, it is clear that there is an error in measurement as suggested by the periodic simulation analysis.

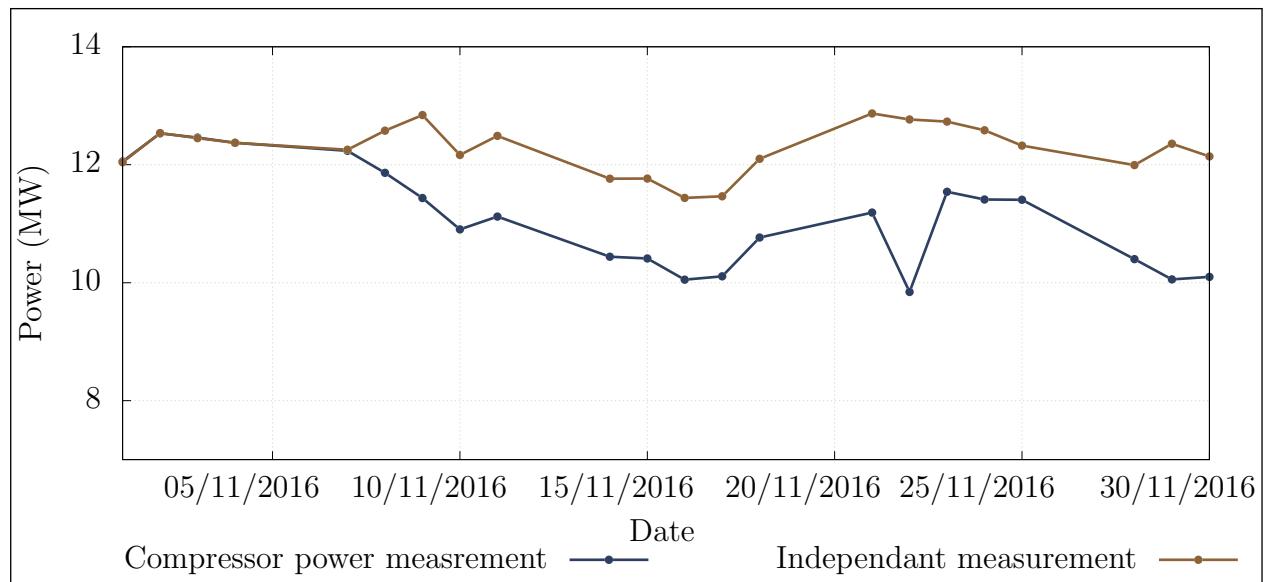


Figure 4.22: Comparison using alternative power source.

4.4.4 Summary

An investigation into periodic or repeated simulation was performed. Using the simulation model that was developed for Mine B, the periodic simulation methodology was implemented for the duration of a month. The results indicated that the model's process parameters, with the exception of power, were accurate for the duration of the analysis.

A shift in output power for a compressor for power was identified as a result of error in power measurement. Further analysis is recommended to determine how long a simulation model

could be expected to remain valid before re-calibration is required.

4.5 Potential benefit for SA mines

There are approximately 75 operational gold and Platinum Group Metal (PGM) mines in South Africa, as illustrated in Figure 4.23^{1,2}. Whilst there are some mines that utilise hydraulic systems, most mine use compressed air for underground processes. By utilising the compressed air simulation methodology in this study, these mines could collectively achieve significant energy and cost savings for the industry.

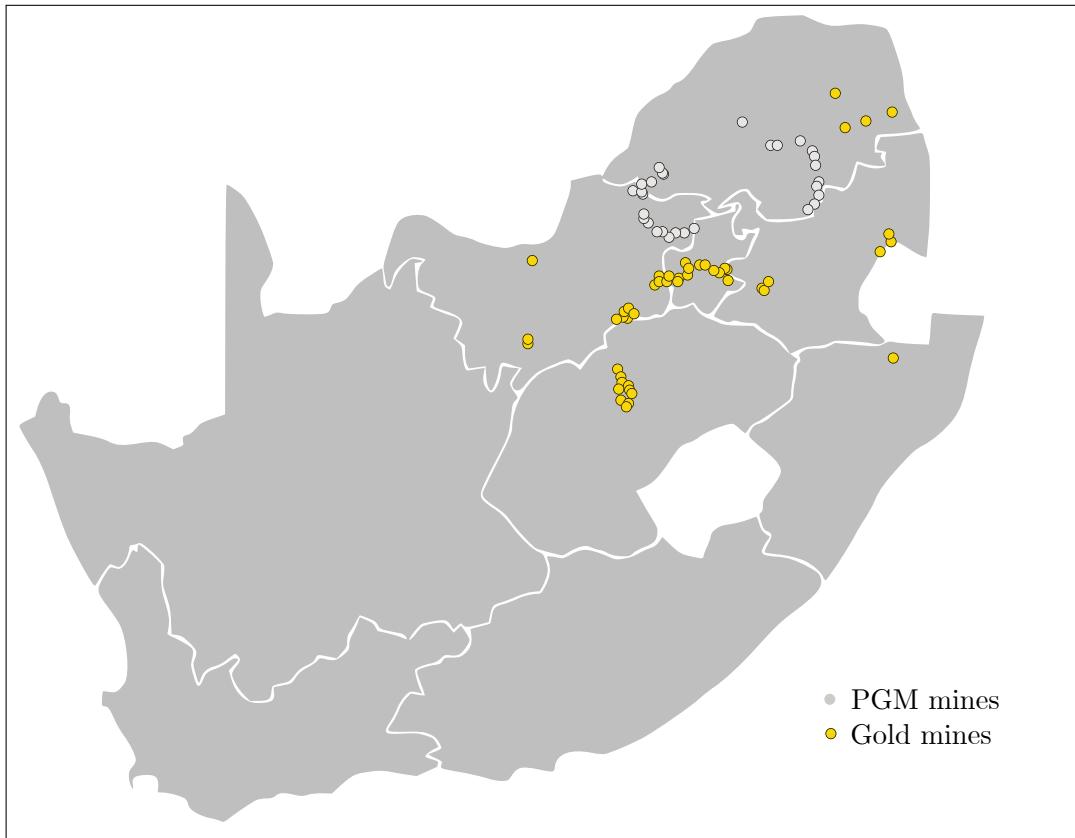


Figure 4.23: Gold and PGM mines in South Africa^{1,2}

In this study, the simulated interventions resulted in savings of (on average) 0.69 MW E.E. or 1.025 MW P.C.. Using the following assumptions:

- There are 75 operational mines
- Similar savings could be identified for compressed air at 60% of the operational mines

A potential energy saving of approximately 30 MW E.E. or 45 MW P.C. could be achieved. The combined cost saving for these interventions would amount to up to R 240m p.a.

¹ Chamber of Mines, [Online] <http://www.chamberofmines.org.za/sa-mining/gold>, [Accessed 16-06-2017]

² Chamber of Mines, [Online] <http://www.chamberofmines.org.za/sa-mining/platinum>, [Accessed 16-06-2017]

4.6 Conclusion

Chapter 4 aimed to validate the methodology developed in Chapter 3. This validation was achieved through the implementation of two case studies South African gold mines. A third case study was performed to analyse and validate the value repeated/periodic simulation.

In case study 1, the compressed air network for gold mining complex consisting of three mining shafts was investigated. A simulation model was then developed, and the accuracy was verified using typical operation from historical data. Two scenarios were simulated, Optimising compressor set-points and reducing pressure during the evening energy peak.

The results showed that there was a potential P.C. power reduction of 0.46 and 1.0 MW respectively. The reduction in power during the energy peak times could lead to a potential cost reduction of up to R 0.91m. The result was validated using results of actual tests on the system.

In case study 2, the same procedure was applied to another gold mine. The mine consisted of a main and sub-shaft. An investigation was done to gather data and identify potential interventions. The investigation included underground spot checks and measurements.

A simulation model was developed for the system. Two scenarios were simulated using the developed model. The simulations showed an E.E. saving of up to 0.92 MW for scenario 1 and a P.C. impact of up to 2 MW for scenario 2. The intervention could lead to an energy cost saving of R 5.17m for the mine. Additional pressure improvements were identified in case study 1. The results were reported to the mine.

A third case study was performed to validate the periodic or repeated simulation methodology. The results showed that the simulation model remained valid for consecutive days in the tested date range. Faulty power measurements were also identified from the results. Further analysis was recommended to determine how long a simulation model can remain valid before calibration is required.

Finally, the result of large-scale implementation of compressed air simulation in South African gold and PGM mines. A conservative estimation showed that a potential power reduction of 50 MW E.E. or 75 MW P.C. could be obtained. This energy improvement would lead to a significant total energy cost saving of R 400m p.a.

CHAPTER 5

Conclusion

5.1 Preamble

Chapter 5 serves as a conclusion for this dissertation. An overview of the complete dissertation will be provided. This overview will summarise the work done in the proceeding chapters. The limitations of the study will then be discussed. This overview will lead to a discussion of recommendations for future work that can be done in the field.

5.2 Dissertation overview

The South African mining sector is currently facing significant challenges that pose a risk to the profitability of the industry. A central challenge that faces the industry is that of rising operational costs. Energy costs contribute a significant portion of the cost increases as energy tariff increases have constantly surpassed inflation over the past ten years.

Compressed air systems consume the largest portion of energy used in a mine. Compressed air has also been shown to be largely inefficient. It is therefore reasoned that the greatest energy impact can be achieved through compressed air interventions.

Energy interventions in mining compressed air have been performed in the past. However, compressed air simulation has not been used to its full potential. The specific effects of interventions can be identified with minimal risk using new computer modelling and simulation tools for compressed air systems; This will lead to further energy and cost reductions as well as other potential improvements for the operation of a mine.

A review of background and literature was performed. The purpose of the review was to provide background on mining compressed air networks and to evaluate literature in pertaining to compressed air energy interventions, simulation usage in the mining industry and specifically simulation usage in compressed air systems.

A methodology was developed for compressed air simulation using the findings in from the literature review. The method describes the simulation procedure in three steps summarised as follows: investigate the system, develop a simulation model and execute the simulation scenarios.

The simulation methodology implemented with case studies. The studies were performed on two different mines compressed air systems. A full system investigation was conducted in each study. From the data and information obtained from the investigation, simulation models were developed for both systems. The models were verified using measurement data

from the physical system. The models were used to simulate various operation scenarios.

In Case study 1, two scenarios were simulated. The results first scenario, reducing compressor setpoints, showed potential power reduction of 0.46 MW P.C. which would result in a yearly energy cost saving of R0.37M. Scenario 2, reducing underground control valve pressure, showed a potential power reduction of 1.0 MW P.C. which would result in a yearly energy cost saving of R0.91M. The result of the simulation was validated by comparison with the actual results from the test results from the physical system.

In Case study 2, two scenarios were simulated. The first scenario looked at reducing refuge bay leaks. The results showed a potential energy efficiency improvement of 0.92 MW, which would result in an annual energy cost saving of R5.17M. A further pressure improvement of 15 kPa during the drilling time was identified. In Scenario 2, optimising peak time demand through station and stope control was investigated. The results showed a potential power reduction of up to 2.0 MW P.C. could be obtained. The optimisation would lead to energy cost saving of R0.91M p.a.

In case study 3, an analysis into repeated, periodic simulations was done. The investigation aimed to check the validity of simulation models and to identify when major shifts in a compressed air system's operation occur. A faulty power meter was identified. However, from the pressure and flow process parameters, the analysis showed that the model accurately represented the system over the period of repeated simulations.

Implementation of the simulation methodology developed in this study to other gold and PGM mine compressed air systems would significant energy cost and operation improvement for the industry. It was estimated that up to a 50 MW energy efficiency improvements could be achieved. This increase in efficiency would lead to cost savings up to R400M p.a for the industry.

5.3 Recommendations for future studies

- Simulation methodology for other mining systems, ventilation, water reticulation
- A sensitivity analysis to analyse the accuracy of simulations with different boundary choices.
- Further analysis of error calculations and verification methods.
- Quantify financial benefits of pressure improvements for drilling
- more...

Unfinished

Bibliography

- [1] P. Neingo and T. Tholana, “Trends in productivity in the South African gold mining industry,” *Journal of the Southern African Institute of Mining and Metallurgy*, vol. 116, no. 3, pp. 283–290, 2016.
- [2] G. M. Mudd, “Global trends in gold mining: Towards quantifying environmental and resource sustainability,” *Resources Policy*, vol. 32, no. 1, pp. 42–56, 2007.
- [3] J. Müller and H. Frimmel, “Numerical analysis of historic gold production cycles and implications for future sub-cycles,” *The Open Geology Journal*, vol. 4, no. 1, pp. 29–34, 2010.
- [4] J. Vosloo, L. Liebenberg, and D. Velleman, “Case study: Energy savings for a deep-mine water reticulation system,” *Applied Energy*, vol. 92, pp. 328–335, 2012.
- [5] H. Le Roux, “Energy consumption reduction challenge,” *Mining Weekly, PO Box*, vol. 75316, p. 30, 2005.
- [6] J. Marais, “An integrated approach to optimise energy consumption of mine compressed air systems,” Ph.D. thesis, North-West University, 2012.
- [7] L. Padachi, G. Bolt, M. Kleingeld, J. Marais, and E. Mathews, “Energy efficiency through the integration of compressed air supply control with air network demand control,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2009, pp. 77–80.
- [8] J. Du Plessis and R. Pelzer, “Development of an intelligent control system for mine compressor systems,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2011, pp. 59–63.
- [9] P. Fraser, “Saving energy by replacing compressed air with localized hydropower systems: a ‘half level’model approach,” *Platinum in Transformation*, 2008.
- [10] J. Neale and P. Kamp, “Compressed air system best practice programmes: What needs to change to secure long-term energy savings for New Zealand?” *Energy Policy*, vol. 37, no. 9, pp. 3400–3408, 2009.

- [11] J. Marais and J. v. M Kleingeld, “Simplification of mine compressed air systems,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2013, pp. 1–8.
- [12] L. B. N. Laboratory, *Improving Compressed Air System Performance: A Sourcebook for Industry*, 2nd ed. Washington DC: US Department of Energy, 2004.
- [13] A. van Tonder, “Sustaining compressed air DSM project savings using an air leakage management system,” Masters dissertation, North-West University, 2011.
- [14] J. Snyman, “Integrating various energy saving initiatives on compressed air systems of typical South African gold mines,” Masters dissertation, North-West University, 2010.
- [15] C. Kocsis, R. Hall, and S. Hardcastle, “The integration of mine simulation and ventilation simulation to develop a ‘life cycle’ mine ventilation system,” *Application of Computers and Operations Research in the Minerals Industries, South African Institute of Mining and Metallurgy*, pp. 223–230, 2003.
- [16] J. Banks, *Handbook of simulation: principles, methodology, advances, applications, and practice*, 1st ed. John Wiley & Sons, 1998, ch. 1, pp. 3–30.
- [17] C. Kriel, “Modernising underground compressed air DSM projects to reduce operating costs,” Masters dissertation, North-West University, 2014.
- [18] J. Bredenkamp, A. Schutte, and J. F. van Rensburg, “Challenges faced during implementation of a compressed air energy savings project on a gold mine,” in *Proceedings of the 12th Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2015.
- [19] P. Maré, J. Bredenkamp, and J. Marais, “Evaluating compressed air operational improvements on a deep-level mine through simulations,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2017.
- [20] S. Fouché, “Improving efficiency of a mine compressed air system,” Masters dissertation, North-West University, 2016.
- [21] W. Booysen, “Reducing energy on RSA mines through optimised compressor control,” Masters dissertation, North-West University, 2010.
- [22] M. Yang, “Air compressor efficiency in a vietnamese enterprise,” *Energy Policy*, vol. 37, no. 6, pp. 2327–2337, 2009.

- [23] A. van Jaarsveld, S. Ebben, and A. Ashanti, “Development and implementation of an electrically powered stope rockdrill for Tau Tona mine,” in *SAIMM Narrow Vein and Reef Conference, Sun City, South Africa*, 2008.
- [24] D. Brake and G. Bates, “Criteria for the design of emergency refuge stations for an underground metal mine,” in *AusIMM Proceedings*, vol. 304, no. 2. The Australian Institute of Mining and Metallurgy, 1999, pp. 1–8.
- [25] F. Schroeder, “Energy efficiency opportunities in mine compressed air systems,” Masters dissertation, North-West University, 2009.
- [26] M. Kleingeld and J. Marais, “A high level strategy plan for reducing a mine group’s dependence on compressed air,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2010.
- [27] J. Marais and M. Kleingeld, “An expert control system to achieve energy savings on mine compressed air systems,” in *Proceedings of the Industrial and Commercial Use of Energy Conference(ICUE)*. IEEE, 2010.
- [28] W. Booysen, M. Kleingeld, and J. F. van Rensburg, “Optimising compressor control strategies for maximum energy savings,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2009, pp. 10–12.
- [29] J. Bredenkamp, “Reconfiguring mining compressed air networks for cost savings,” Masters dissertation, North-West University, 2013.
- [30] R. Saidur, N. Rahim, and M. Hasanuzzaman, “A review on compressed-air energy use and energy savings,” *Renewable and Sustainable Energy Reviews*, vol. 14, no. 4, pp. 1135 – 1153, 2010.
- [31] A. J. M. van Tonder, “Sustaining compressed air DSM project savings using an air leakage management system,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2011, pp. 133–137.
- [32] B. Pascoe, “Improving mine compressed air network efficiency through demand and supply control,” Masters dissertation, North-West University, 2016.
- [33] J. Marais, M. Kleingeld, and R. Pelzer, “Increased energy savings through a compressed air leakage documentation system,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2009.

- [34] H. Nester, R. Pelzer, and M. Kleingeld, “Energy savings through the automatic control of underground air demand,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2008.
- [35] S. Bester, D. le Roux, and D. Adams, “The effect of compressed air pressure on mining production and energy demand,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2013, pp. 1–4.
- [36] C. Cilliers, “Benchmarking electricity use of deep-level mines,” Ph.D. thesis, North-West University, 2015.
- [37] W. van Niekerk *et al.*, “The value of simulation models when implementing mine DSM projects,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2013, pp. 1–4.
- [38] W. van Niekerk, “The value of simulation models for mine DSM projects,” Masters dissertation, North-West University, 2012.
- [39] A. Holman, “Benefits of improved performance monitoring of mine cooling systems,” Masters dissertation, North-West University, 2014.
- [40] P. Maré, “Novel simulations for energy management of mine cooling systems,” Ph.D. thesis, North-West University, 2016.
- [41] W. Bouwer, “Designing a dynamic thermal and energy system simulation scheme for cross industry applications,” Ph.D. thesis, North-West University, 2004.
- [42] Y. Hollander and R. Liu, “Estimation of the distribution of travel times by repeated simulation,” *Transportation Research Part C*, vol. 16, pp. 212–231, 2008.
- [43] S. W. van Heerden, R. Pelzer, and J. Marais, “Developing a dynamic control system for mine compressed air networks,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2014, pp. 1–8.
- [44] A. van Tonder, “Automation of compressor networks through a dynamic control system,” Ph.D. thesis, North-West University, 2014.
- [45] H. Sarin *et al.*, “Comparing time histories for validation of simulation models: error measures and metrics,” *Journal of dynamic systems, measurement, and control*, vol. 132, no. 6, p. 061401, 2010.

- [46] Y.-j. Xu, S.-j. Zhang, and Y.-h. Xiao, “Modeling the dynamic simulation and control of a single effect LiBr–H₂O absorption chiller,” *Applied Thermal Engineering*, vol. 107, pp. 1183–1191, 2016.
- [47] J. Kurnia, A. P. Sasmito, and A. S. Mujumdar, “Simulation of a novel intermittent ventilation system for underground mines,” *Tunnelling and Underground Space Technology*, vol. 42, pp. 206–215, 2014.
- [48] J. C. Kurnia, A. P. Sasmito, and A. S. Mujumdar, “Dust dispersion and management in underground mining faces,” *International Journal of Mining Science and Technology*, vol. 24, no. 1, pp. 39–44, 2014.
- [49] D. C. Arndt, “Integrated dynamic simulation of large thermal systems,” Ph.D. thesis, University of Pretoria, 2000.
- [50] M. Sprague and T. Geers, “Objective error measure for the comparison of calculated and measured transient response histories,” *Shock and Vibration Bulletin*, pp. 99–107, 1984.
- [51] M. Sprague and T. Geers, “Spectral element method for modelling cavitation in transient fluid-structure interaction,” *International Journal for Numerical Methods in Engineering*, p. 2467–2499, 2004.
- [52] D. Russell, “Error measures for comparing transient data: Part I, development of a comprehensive error measure,” *Proceedings of the 68th Shock and Vibration Symposium*, pp. 175–184, 1997.
- [53] D. Russell, “Error measures for comparing transient data: Part II, error measures case study,” *Proceedings of the 68th Shock and Vibration Symposium*, pp. 212–231, 1997.
- [54] C. J. Willmott and K. Matsuura, “Advantages of the mean absolute error (MAE) over the root mean square error (RMSE) in assessing average model performance,” *Climate research*, vol. 30, no. 1, pp. 79–82, 2005.
- [55] S. Dominic and Maier, “Dynamic modeling and simulation of compressor trains for an air separation unit,” *IFAC Proceedings Volumes*, vol. 47, no. 3, pp. 432–437, 2014.
- [56] G. E. D. Plessis, D. C. Arndt, and E. H. Mathews, “The development and integrated simulation of a variable water flow energy saving strategy for deep-mine cooling systems,” *Sustainable Energy Technologies and Assessments*, vol. 10, pp. 71–78, 2015.
- [57] P. Peach, “Optimising deep-level mine refrigeration control for sustainable cost savings,” Masters dissertation, North-West University, 2016.

- [58] S. Mousavi, S. Kara, and B. Kornfeld, “Energy efficiency of compressed air systems,” *Procedia CIRP*, vol. 15, pp. 313–318, 2014.
- [59] J. Zahlan and S. Asfour, “A multi-objective approach for determining optimal air compressor location in a manufacturing facility,” *Journal of Manufacturing Systems*, vol. 35, pp. 176–190, 2015.
- [60] A. Gous, W. Booysen, and W. Hamer, “Data quality evaluation for measurement and verification processes,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2016, pp. 9–15.
- [61] L. Meijzen, J. van Rensburg, and W. Booysen, “Verification procedures to ensure consistent energy metering,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2015, pp. 138–146.
- [62] S. van Niekerk, “Quantification of energy consumption and production drivers in steel manufacturing plants,” Masters dissertation, North-West University, 2016.
- [63] H. J. van Rensburg, “Structuring mining data for RSA section 12l EE tax incentives,” Masters dissertation, North-West University, 2016.
- [64] G. Heyns, “Challenges faced during implementation of a compressed air energy savings project on a gold mine,” Masters dissertation, North-West University, 2014.
- [65] W.van Niekerk *et al.*, “The implementation of a dynamic air compressor selector system in mines,” in *Proceedings of the Industrial and Commercial Use of Energy Conference (ICUE)*. IEEE, 2015, pp. 129–132.

APPENDIX I

Simulation process flow schematics

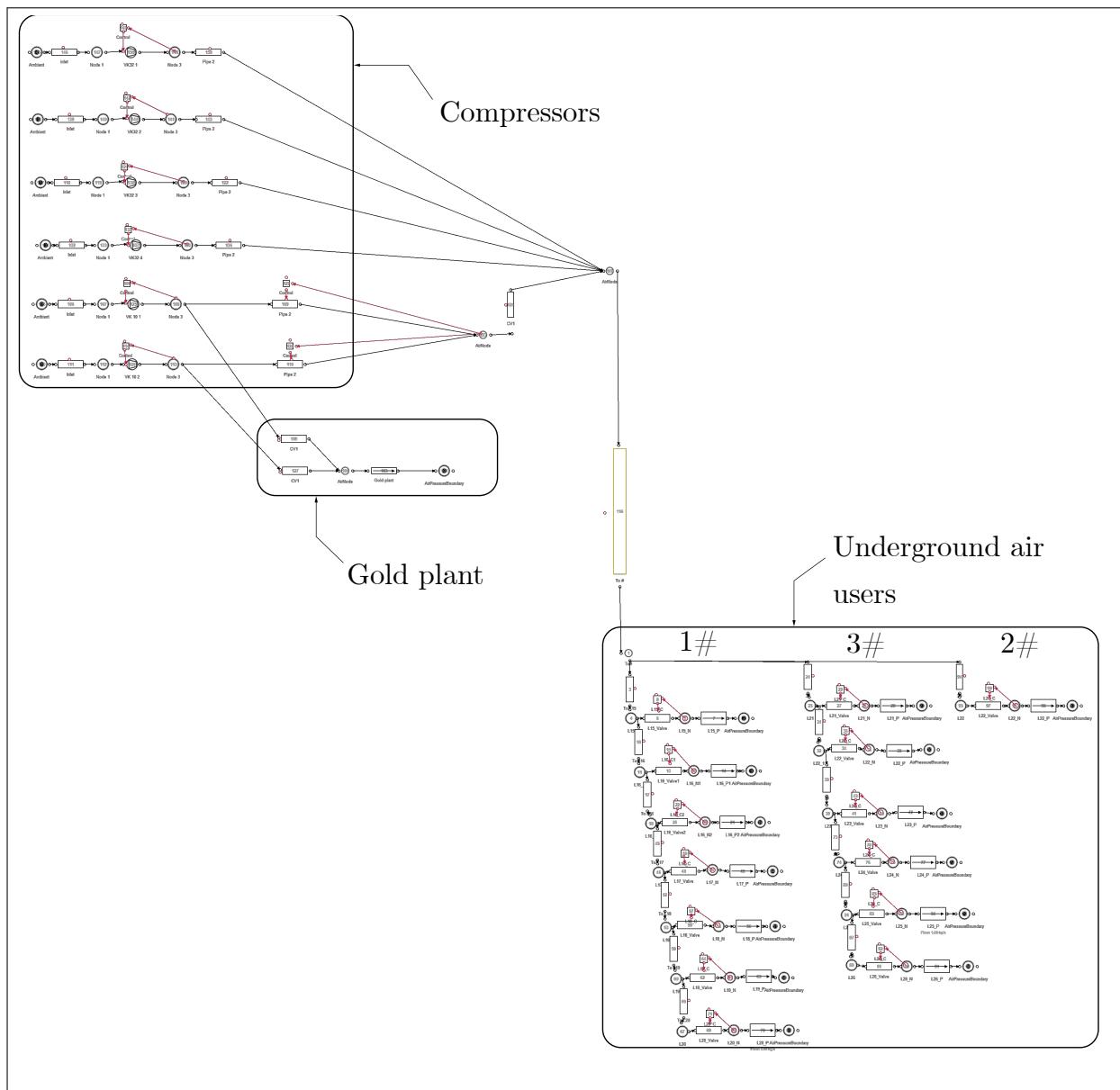


Figure I.1: Mine A: Simulation process flow schematic

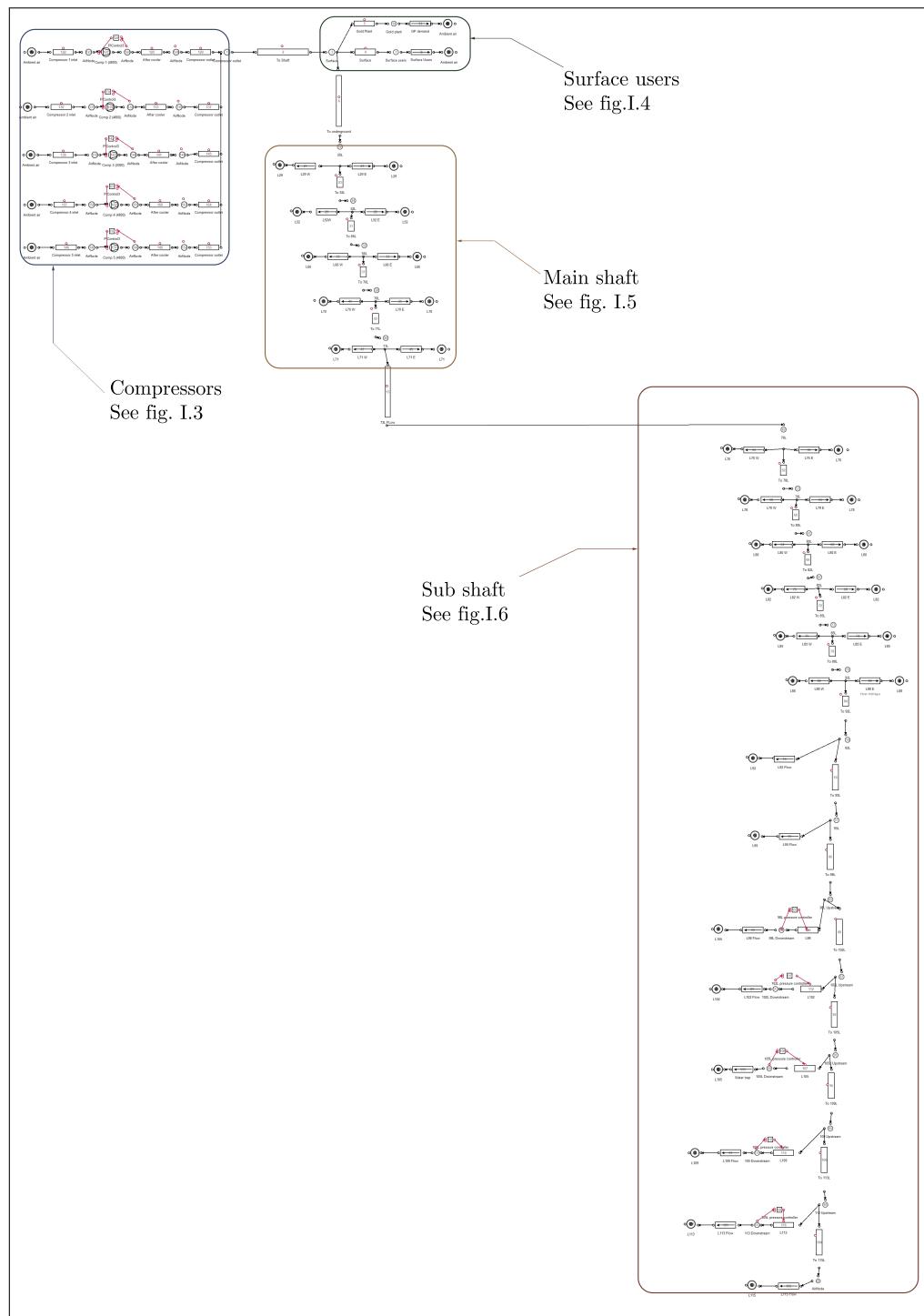


Figure I.2: Mine B: Baseline process flow schematic

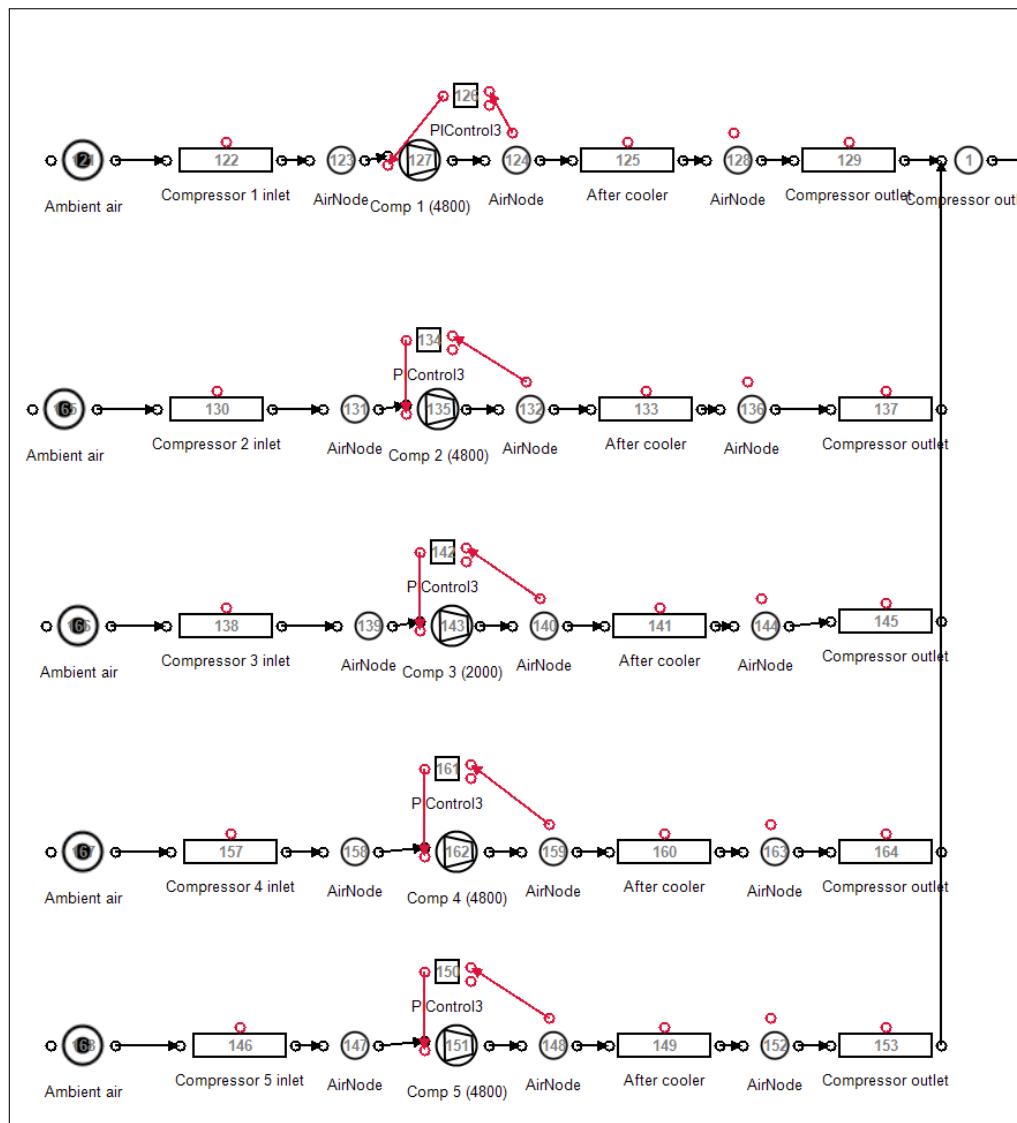


Figure I.3: Mine B: Compressors

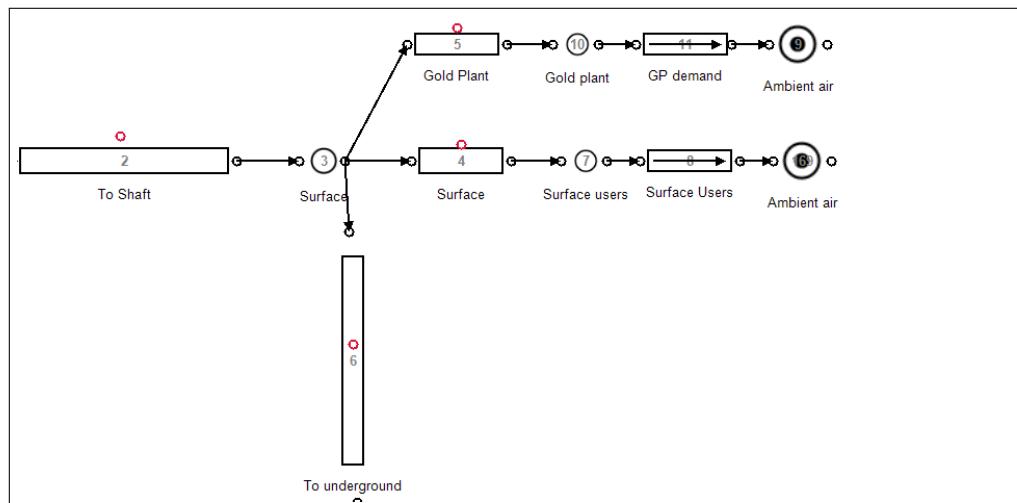


Figure I.4: Mine B: surface compressed air users

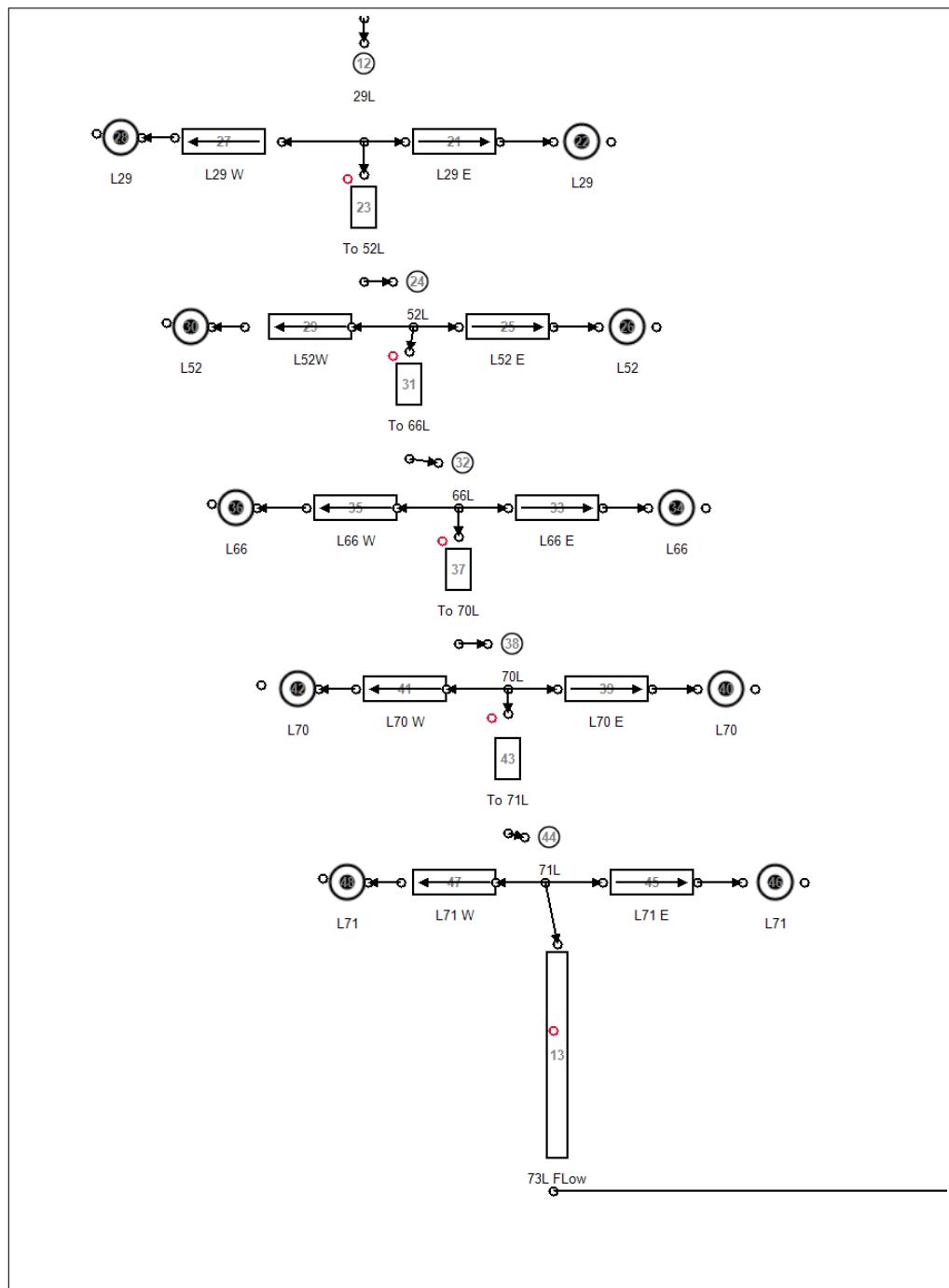


Figure I.5: Mine B: Main shaft compressed air users

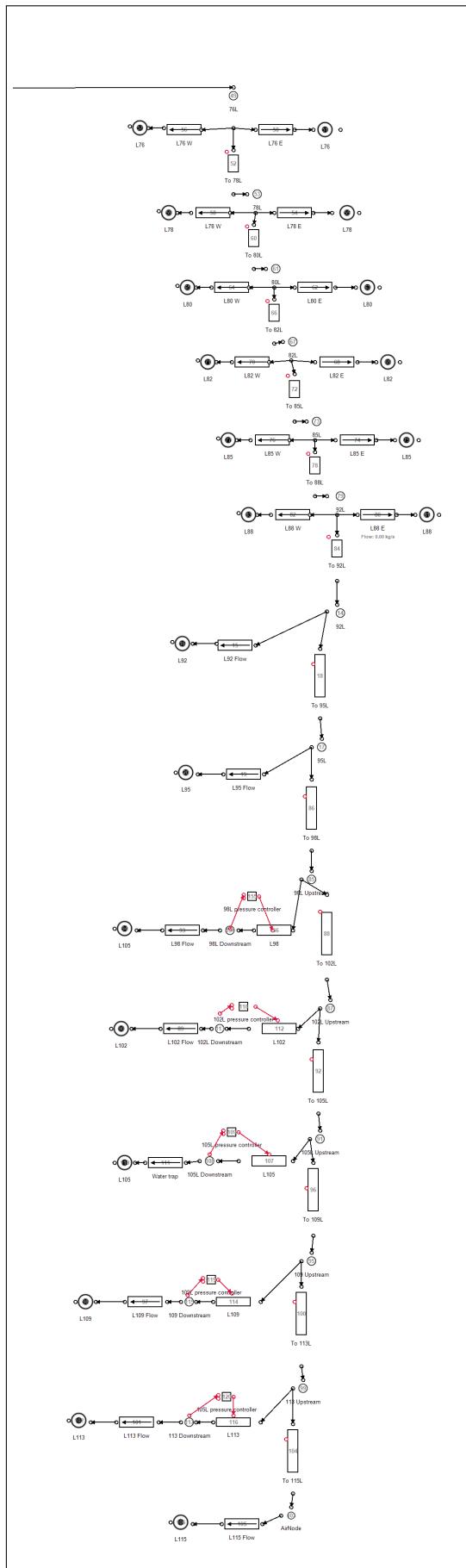


Figure I.6: Mine B: Sub-shaft compressed air users

APPENDIX II

Model components verification

Component	Actual Ave.	Simulated Ave.	Accuracy
Power (kW)			
VK-32 1	0	0	100%
VK-32 2	460	477	96.35%
VK-32 3	142	117	82.31%
VK-32 4	428	408	95.45%
VK-32 5	1813	1903	95.03%
VK-32 6	732	725	99.01%
VK-10 1	744	745	90.02%
VK-10 2	687	635	92.24%
System	4940	4911	97.34%
Flow (kg/s)			
1# 15L	1.45	1.42	97.87%
1# 16L	2.15	2.18	98.48%
1# 17L	0.34	0.35	97.30%
1# 18L	0.38	0.39	97.39%
1# 19L	0.31	0.31	98.77%
1# 20L	0.12	0.13	92.65%
2# 23L	1.31	1.35	96.69%
3# 21L	0.67	0.66	98.68%
3# 22L	0.74	0.67	89.54%
3# 23L	0.04	0.04	98.72%
3# 24L	0.11	0.11	98.76%
3# 25L	0.33	0.33	99.95%
3# 26L	0.45	0.47	95.07%
Gold Plant	2.59	2.48	95.72%
Total	9.32	9.51	97.01%

Table II.1: Case study 1: Model verification

Component	Actual Ave.	Simulated Ave.	Accuracy
Pressure (kPa)			
1# 15L	394	384	98.55%
1# 16L	421	419	99.61%
1# 17L	345	344	99.81%
1# 18L	336	336	99.99%
1# 19L	311	309	99.53%
1# 20L	368	368	99.90%
2# 23L	365	327	91.25%
3# 21L	303	302	99.92%
3# 22L	332	301	95.42%
3# 23L	332	332	99.98%
3# 24L	413	409	99.10%
3# 25L	413	409	99.01%
3# 26L	515	509	98.99%
Surface	502	501	99.95%

Table II.2: Case study 2: Model verification

Component	Actual Ave.	Simulated Ave.	Accuracy
Power (kW)			
Compressor 1	3406	3669	92.37%
Compressor 2	3911	3668	93.92%
Compressor 3	1440	1453	99.05%
Compressor 4	0	0	100%
Compressor 5	3299	3274	99.22%
System	12057	12103	98.73%
Flow (kg/s)			
95L	1.51	1.42	93.95%
98L	3.75	3.53	93.99%
102L	2.97	2.79	98.72%
105L	5.65	5.71	98.84%
109L	3.57	3.37	94.27%
113L	5.09	4.84	95.05%
Gold Plant	1.41	1.35	95.14%
Sub-shaft total	34.12	34.76	98.09%
Total	41.65	41.43	98.96%
Pressure (kPa)			
Surface	393	396	99.02%

Table II.3: Case study 2: Model verification continued

APPENDIX III

Detailed mining level investigation diagrams

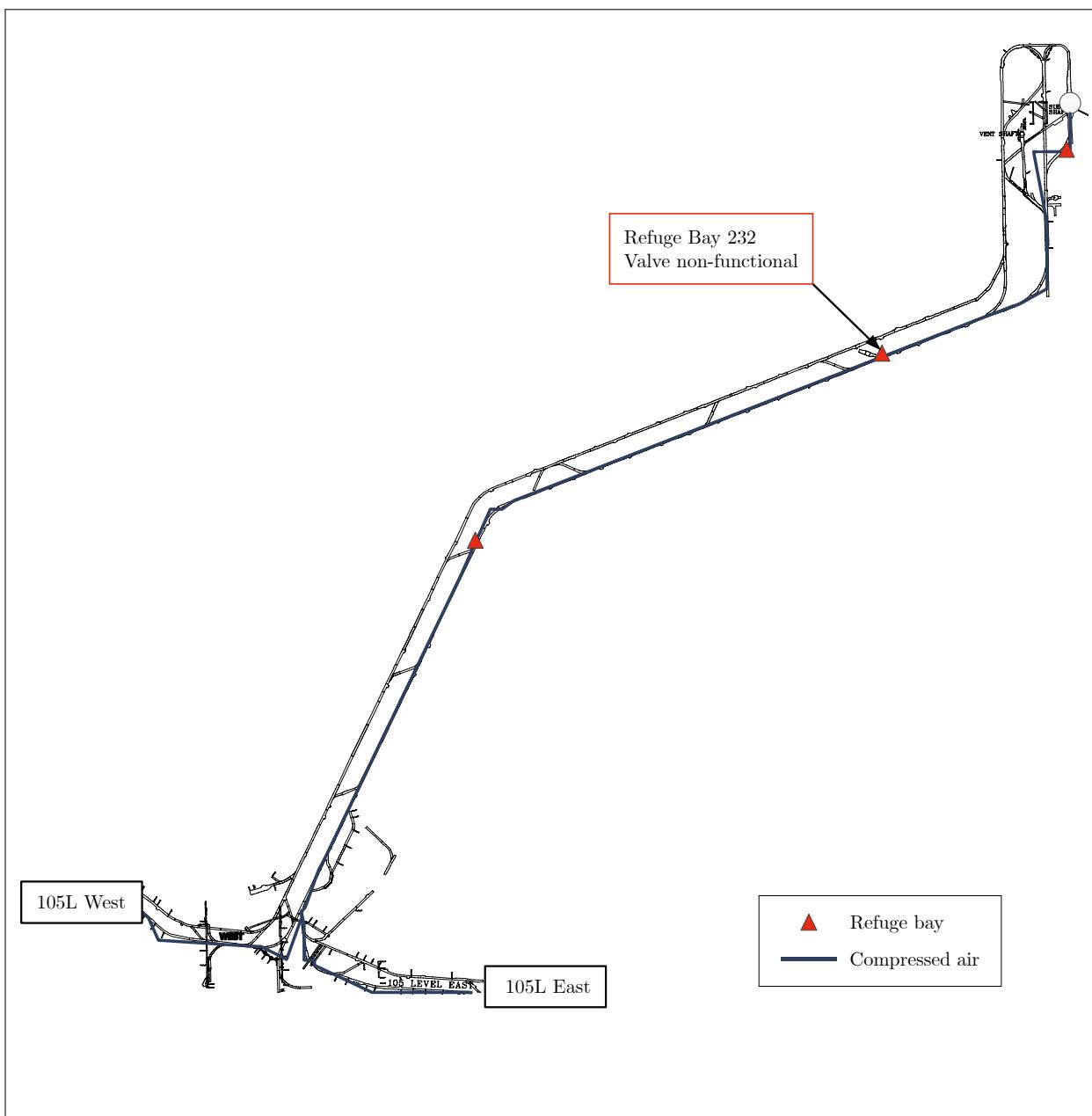


Figure III.1: 105L investigation: Station

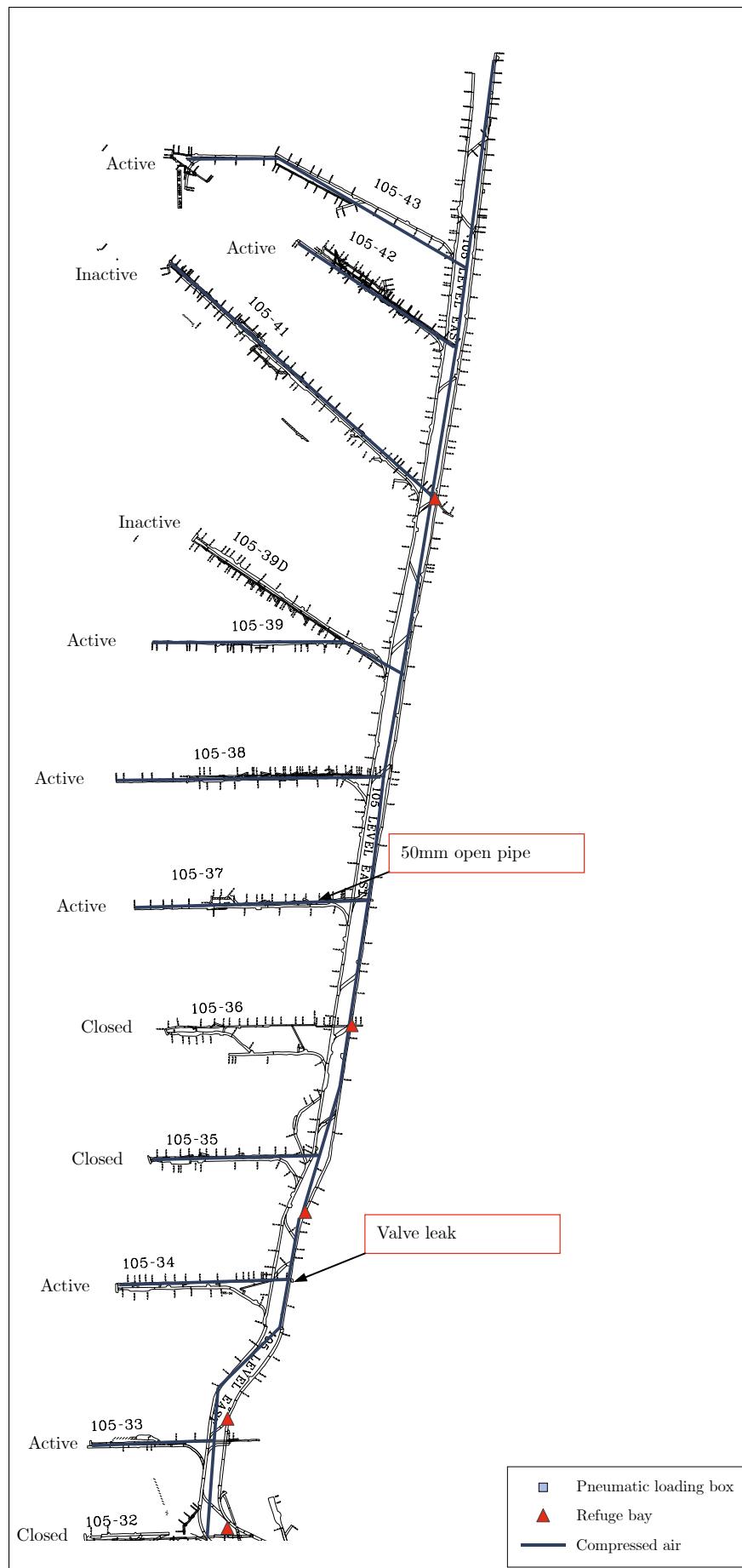


Figure III.2: 105L investigation: West

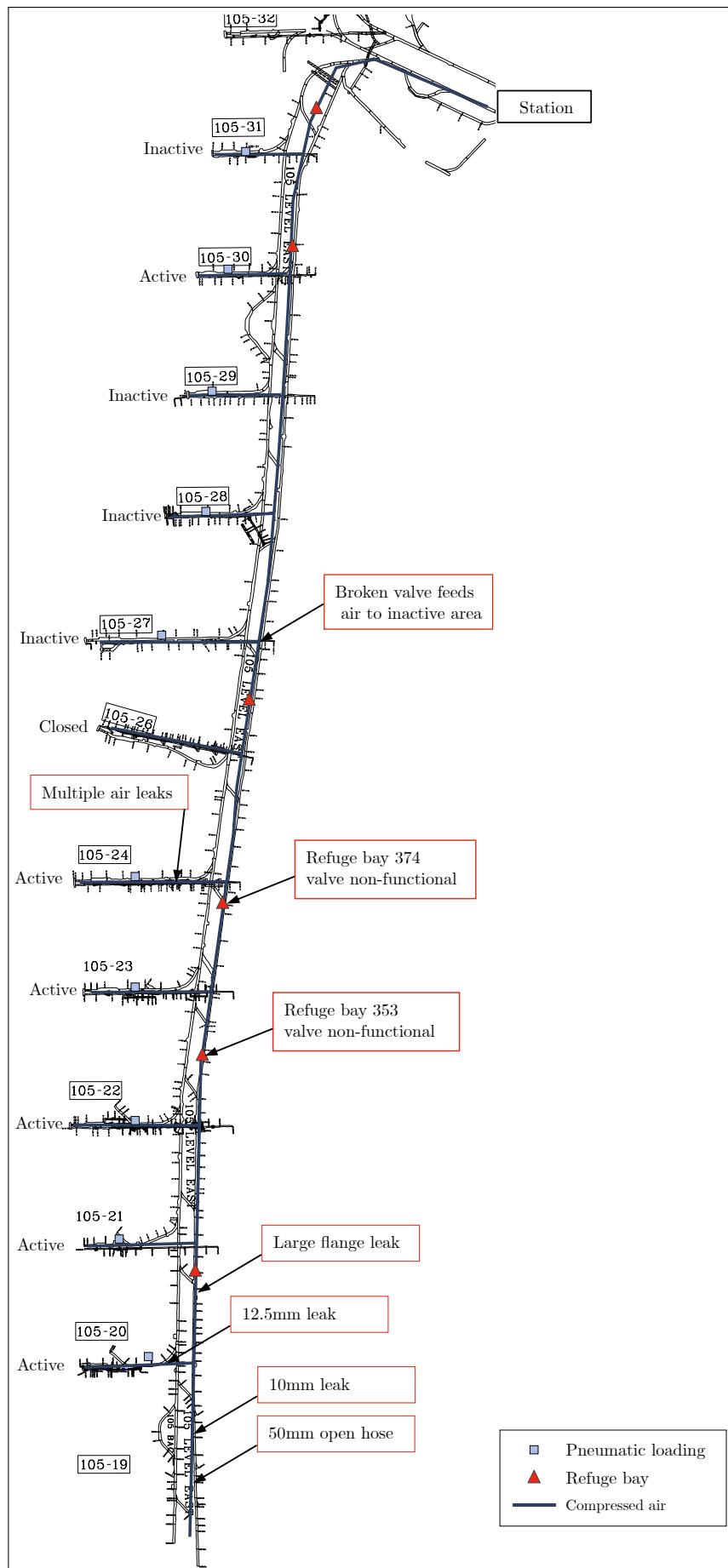


Figure III.3: 105L investigation: East