Using near-infrared spectroscopy (NIRS) technology for ore sorting is a new concept for the mineral processing and mining industry. NIRS has been used for decades in the laboratory for the identification of pure minerals, and has its roots in the recycling, food, and pharmaceutical industries (Salter and Wyatt, 2009; Death et al., 2009). Ore sorting in itself is not a new concept; it was introduced during the Stone Age with hand-sorting being the first method to separate valuable materials from gangue (Wills and Napier-Munn, 2006; Sassos, 1985). However, over the years the scale of mining has increased and hand-sorting has become challenging to implement (Jonsson, 2014).

Sensor-based sorting is an advanced technology that can help overcome problems such as ore dilution experienced in mineral processing operations. The benefits are lower energy and water costs, reduced ore dilution, lower environmental impact, and improved profit margin (Mathew, 1974; Jonsson, 2014; Salter and Wyatt, 2009; Death et al., 2009). In sensor-based sorting, materials are mechanically sorted based on certain physical properties that are detected by a sensor (Dalm, 2013). Research has found that with NIR scanners it is possible to obtain a ‘fingerprint’ of a sample that is directly related to its mineralogical composition. According to Woutruba et al. (2009), molecules absorb NIR radiation by means of electronic and vibrational transitions. Nowadays, there are many sensors available in the market that utilize different material properties; the choice of a sensor is dependent on the mineralogy of the ore.

Traditional metallurgical techniques for diamond winning from kimberlitic ore involve size reduction, dense medium separation (DMS), and final recovery by X-ray technology (Ketelhodt et al., 2013). These techniques have no ability to remove waste rocks that are often associated with the kimberlite. The kimberlite orebody at Cullinan Diamond Mine (CDM) is intruded by sills and dykes of waste rock, and the current mining method cannot separate the waste material (Bartlett, 1994). Waste hoisted from underground is mostly from development areas. This paper evaluates the viability of the new COLOR/NIR sorting technology to remove the waste rocks from the kimberlite ore stream, thus reducing the amount of waste going into the DMS process.
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Experimental procedures

Material characterization

In order to provide background information for sensor-based sorting, mineralogical characterization of a run of mine (ROM) sample from CDM was performed by X-ray diffraction (XRD). The primary use of XRD is for the identification and characterization of compounds based on their diffraction patterns. The sample, which had a top size of -70 mm, was reduced using a single toggle jaw crusher followed by ball milling. The XRD results are shown in Figure 1 and Figure 2.

Both the kimberlite (Figure 1) and the waste (Figure 2) were found to contain NIR-active minerals, and there is thus the possibility of using NIR technology to discriminate between the two streams. Images (Figure 3) extracted from the classifier also showed a very good possibility of separating kimberlite from the associated waste materials. Classifiers are part of the NIR technology programme and are used to classify materials (Korsten, 2014).

Preliminary ore sorting studies

Tests work was conducted with the Pro Secondary COLOR/NIR sorter illustrated in Figure 4. The sorter uses a combination of a high-resolution line camera and NIR scanner to provide accurate detection of the mineral fingerprint of each rock (Ketelhodt et al., 2013). The working width of the COLOR/NIR sorter is 1200 mm.

The technical set-up of the sorter is shown in Figure 5. The sorter used was a chute-type sorter, with the material fed by a vibratory feeder and accelerated down a chute under gravity. Scanning is done by means of an NIR and a colour sensor on the free-falling stream of particles. Prior to the sorter, a washing/sizing screen with a 10 mm aperture size was installed for the purpose of particle preparation. Sizing was done to obtained a size ratio of 1:3, because research on
Sensor-based technology has shown that the sorting results can deteriorate at a particle size ratio greater than three (Wotruba and Riedel, 2005).

The material that was supplied for the test work was a 12 t sample of Cullinan ROM material. The waste rocks at CDM comprise of gabbro, norite, felsite, and quartzite. The sample, which was collected by means of a belt cut, was within the size range of -70+35 mm. The sorter needs to be programmed with information obtained from a pre-classified sample. Thus, a 10 kg sample of the kimberlites and waste material was collected from underground at Cullinan. Groups were formed by statistical means and applied into the sorting algorithm, called the classifier, for bulk sorting. The process flow diagram configuration shown in Figure 6 was tested with the aim of maximizing overall kimberlite recovery with a target of 5% kimberlite loss to the waste stream.

Preliminary ore sorting results

Rougher stage

The overall results of the tests that were carried out only in the rougher stage are shown in Table I and Figure 7. The average kimberlite recovery was calculated to be 97.8%, while the average waste removal was calculated to be 90.2%.

Scavenger stage

The scavenger stage was used to recover the kimberlite misclassified into the waste stream from the rougher stage. Approximately 4.1% of the kimberlite is calculated to be lost at the rougher stage. Table II and Figure 8 show the results for the scavenger stage. The average kimberlite recovery was calculated to be 99.8% and the average waste removal was calculated to be 93.1%.

The results clearly indicate that the introduction of a scavenger stage improves the separation efficiency and minimizes the potential for diamond loss. It is also seen that the kimberlite recovery decreases at a low classifier sensitivity. This is due to the fact that low classifier sensitivity reduces the aggressiveness of the machine with respect to its ejecting potential. In both the rougher and scavenger stage the kimberlite recovery was calculated to be above 95%, which clearly indicates that the COLOR/NIR sorter is able to discriminate between the CDM kimberlitic and waste rocks.

Table I

| Test | Classifier sensitivity setting (%) | Feed (kg) | Kimberlite in feed (kg) | Waste in feed (kg) | Kimberlite lost | Kimberlite recovery | Waste removal |
|------|-----------------------------------|----------|------------------------|-------------------|----------------|---------------------|--------------|
|      |                                   |          |                        |                   | Roug. (%) | Scav. (%) | (kg) | (%) | (kg) | (%) |
| 1    | 50                                 | 958.0    | 457.1                  | 509.9             | 1.03       | -        | 452.0 | 98.89 | 480.0 | 94.13 |
| 3    | 55                                 | 1220.0   | 616.5                  | 603.5             | 5.64       | -        | 563.0 | 94.57 | 550.5 | 92.87 |
| 4    | 65                                 | 1149.0   | 600.0                  | 549.0             | 1.76       | -        | 591.5 | 98.56 | 474.5 | 86.43 |
| 5    | 60                                 | 1376.0   | 685.0                  | 691.0             | 2.93       | -        | 666.0 | 97.23 | 629.0 | 91.02 |
| 6    | 50                                 | 1189.0   | 581.1                  | 607.9             | 4.28       | -        | 579.1 | 99.66 | 525.5 | 86.44 |

Table II

| Test | Classifier sensitivity setting (kg) | Feed (kg) | Kimberlite in feed (kg) | Waste in feed (kg) | Kimberlite lost | Kimberlite recovery | Waste removal |
|------|-----------------------------------|----------|------------------------|-------------------|----------------|---------------------|--------------|
|      |                                   |          |                        |                   | Roug. (%) | Scav. (%) | (kg) | (%) | (kg) | (%) |
| 2    | 45                                 | 1268.0   | 589.0                  | 679.0             | 5.87       | 0.24     | 587.5 | 99.75 | 622.0 | 91.61 |
| 7    | 45                                 | 1047.0   | 535.2                  | 511.8             | 4.83       | 0.34     | 533.6 | 99.69 | 473.4 | 92.50 |
| 8    | 50                                 | 992.0    | 505.8                  | 486.2             | 6.34       | 0.22     | 504.8 | 99.80 | 463.0 | 95.23 |
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Figure 9 compares the kimberlite loss in the rougher and scavenger stages against the target kimberlite loss of 5%. The kimberlite lost to the waste stream was monitored to prevent the risk of diamond loss.

There is a relatively high loss of the kimberlite ore when using only a rougher stage. However, introducing the scavenger stage minimized the risk of kimberlite loss to an average of 0.3%. The kimberlite lost was above target for tests 2, 3, and 8 because the aggressiveness of the machine was kept low.

Pilot plant results

Figure 11 shows the average tons treated through the pilot plant for a period of 6 months (November 2014 to April 2015) when the plant was running at full production.

The total tonnage fed to the pilot plant was lower than the design tonnage of 50 000 t/month, due to the fact that the availability and utilization of the pilot plant was lower than expected. The availability of the pilot plant was critical for successful assessment of the sensor-based sorter and economic evaluation.

Throughout the duration of the project the tailings stream was sampled to monitor the average kimberlite loss. Kimberlite loss is an indication of potential diamond loss, and should be kept at a minimum. The acceptable kimberlite loss was set at a target of below 5%.

Figure 12 shows the kimberlite loss during the pilot plant campaign. It was possible to keep the kimberlite loss below 5% for several months, but in January 2015 a high loss of ore occurred. It was decided to stop the sorter until the problem was found and rectified. The sorter was found to have lost programming and had to be re-programmed with pre-classified samples. It was also realized that the kimberlite loss was above target when the screen sprays were blocked and the feed coated with mud. After the problem was rectified.
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the kimberlite loss was acceptable, as can be seen in Figure 12.

Conclusion
Based on the test work, the COLOR/NIR sorter is generally a good tool to separate Cullinan kimberlite from waste rocks. The introduction of a scavenger stage will improve the kimberlite recovery and also minimize diamond loss. The pilot plant test work results were not consistent since the sensing technology is based on the surface properties of the material, hence the feed needs to be clean and free of dust or mud prior to sorting. However, this is not always the case in a production environment, particularly for kimberlite ores, which have a high content of clay material. Due to the risk of diamond loss as a function of kimberlite lost, it is recommended that a scavenger stage be implemented in further kimberlite preconcentration processes using the same sorting system.

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