Abstract

The current developments with unmanned aerial vehicles (‘UAVs’) are revolutionizing many fields in civil applications, such as agriculture, environmental and visual inspections. The mining industry can also benefit from UAVs in many aspects, and the reconciliation through topographic control is an example. In comparison with traditional topography and maybe modern techniques such as laser scanning, aerial photogrammetry is cheaper, provides faster data acquisition and processing, and generates several high-quality products and impressive level of details in the outputs. However, despite the quality of the software currently available, there is an uncertainty intrinsic to the surfaces acquired by photogrammetry and this discrepancy needs to be assessed in order to validate the techniques applied. To understand the uncertainty, different surfaces were generated by UAVs for a small open pit quarry in southern Brazil. Well-established topographic surveying methodologies were used for geolocation support and comparison, namely the RTK (real-time kinetic) global navigation satellite system (GNSS) (here called conventional method) and laser scanning. The results showed consistency between the UAV surfaces with a few outliers in when vegetation, water and mobile objects during the flight missions. In comparison with the laser-scanned surface, the UAV results were less erratic surrounding the RTK points, showing that surfaces generated by photogrammetry can be a simpler and quicker alternative for mining reconciliation, presenting low uncertainty when compared to conventional methods.

Keywords: reconciliation, topography, UAV, photogrammetry, laser scanning.
1. Introduction

Unmanned aerial vehicles (‘UAV’s) are also known by many other names, such as, unmanned aircraft system (‘UAS’), remotely piloted aircraft (‘RPA’), micro air vehicle (‘MAV’). The market maturity development carries different names for similar aircrafts (Nonami, 2007), which have been intensely developed for civil applications during the last few decades (Valavanis, 2007). Recently, many commercial gadgets allowed the popularization of new technologies and the increasing number of operators and fields of use. The light weight, battery autonomy, sensors improvements and miniaturization made the UAV application not only possible, but accessible for different applications (Herbst et al., 2016). Mining industry benefits from several developments using photogrammetry, such as geomorphology identification (Hugenholtz et al., 2013; Westoby et al., 2012; Javernick et al., 2014), rock mechanics (Salvini et al., 2016), targeting (Nishar et al., 2016), among others. Topographic modelling is one of the applications for the mining industry and it is used for mineral resources and ore reserve estimation, mine planning and reconciliation. Reconciliation in mining practices involves comparing the amount of material moved from the mine to any destination against the quantity and quality predicted by the mine planning.

One of the main sources of uncertainty in mining reconciliation is the topographic model updating (Câmara et al., 2013), which demands that corrections be made on planning and/or operations. Nevertheless, due to elevated costs and time consumption, the surveying frequency might be sparse and, in some cases, can take months without a feedback from the operations. An alternative way to increase the frequency of topography updating is through use of photogrammetry-based surfaces, acquired by UAVs and processed by specific software. Wheatstone (1838) developed an optical methodology for three-dimensional interpretation of models created from 2-D images. A few years later, Laussedat presents the idea of photogrammetry and the multi-images model for larger areas (Guilhot, 2006), introducing the stereoscopy and the three-dimension reconstruction by using two or more observation points, similar to a moving pair of eyes. Eisenbeiss (2009) discusses the application of UAVs in photogrammetry and recent publications investigate geomorphological applications for UAVs (Niethammer et al., 2012; Eltner et al., 2016; Turner et al., 2014 and Lucieer et al., 2014) and others use the methodology in quarries (Shahbazi et al., 2015; Rossi et al., 2017). A certain degree of uncertainty associated with the image acquisition is expected (Colomina and Molina, 2014), such as lenses distortion, GPS imprecision, vegetation density, light glowing or shading and water reflection.

To quantify the uncertainty generated by the new technology, this article compares digital surface models (‘DSM’) generated by photogrammetry using UAVs with those generated using traditional techniques. The surface models were generated for a quarry area in southern Brazil. Well-known topography techniques were used to assess the error of the photogrammetric model by comparison.

2. Materials and methods

Two UAVs were used for the aerial surveying. The aircraft models were two quadcopters Phantom 3 and Phantom 4, both manufactured by DJI®, with original cameras and GPS devices. Their technical specifications are shown on Table 1.

| Specifications             | Phantom 4 | Phantom 3 |
|----------------------------|-----------|-----------|
| Weight (g)                 | 1380      | 1280      |
| Max speed (m/s)            | 20        | 16        |
| Gimbal stabilization       | 3-axis    | 3-axis    |
| Sensor                     | 1 / 2.3° (CMOS) 12.4 Mpx | 1 / 2.3° CMOS 12.4 Mpx |
| Lens                       | FOV 94°   | FOV 94°   |

During the flight, the images overlapping was set at 80% to maintain high definition and confidence levels for the photogrammetry process, allowing some images to be discarded when necessary. The software used for flight mission planning and control was Pix4Dcapture, a freeware available for mobile devices. The acquired images were processed using Agisoft PhotoScan® v 1.2.6. The software was used to orient the pictures and recreate the topography through the point cloud generation, based on the homologous points in two or more images oriented by known ground points. The area was divided to comprise the portions with higher images overlapping, and also to reduce the non-interesting areas with vegetation cover which could introduce differences between the surfaces.

To reference the mapping and to assess the quality of the model reconstruction, a number of ground control points (GCPs) were taken by a GNSS system based on an RTK technique, model Trimble R4, and the points were also used as input for correcting small imprecisions on the mesh created with the images. In total, 14 ground control points were located with the RTK and used as gecorrection benchmarks and for adjustment of the DSMs.

Due to the vertical precision of the equipment and the geoid model adopted, RTK was also used to survey hundreds of additional points, which could be used as real benchmarks for quality check. A Topcon’s laser scanner model GLS-1000 was used to survey points every 20 cm and generate a surface for comparisons, as its use in the mining industry has been popular for more than a decade.

In total, five DSMs were generated, two of which were obtained from Phantom 3 and two from Phantom 4. For comparisons between UAV systems and their shooting direction settings, each equipment produced two sets of photos, one vertical and one with 70 degrees’ oblique, which was the maximum achieved by the controlling application. The last DSM was created from the point cloud obtained from the laser scanning. One of the resulting surfaces is shown in Figure 1, with the control points and orientation in the mine.
The RTK points were also used for visual quality check, but no surface was generated as the number of points is considerably small and cannot be compared with DSMs generated from several thousands of points. The error among the RTK points, assumed as real, and the DSMs was measured by projecting these points on each surface, preserving the XY plan coordinates. The new elevations for the five surfaces were compared against the 680 RTK original points (Figure 2).

For each DSM, the absolute differences were computed and statistically analyzed. The average error, the mean absolute disturbance, which is the sum of the errors in moduli divided by the number of points, mean squared deviation and slope of regression between the original points and the projected elevations are shown in Table 2.

Figure 1
A processed DSM resulting from UAV photogrammetry showing the control points in red and the laser scanner base points in yellow.

Figure 2
RTK original points distribution in plan view.

Table 2
Statistical results for the deviations (error) between each surface against the surveyed hard data.

| Equipment       | Mission            | Aver. error (m) | Mean absolute deviation (m) | Mean squared deviation (m²) | Slope of regression |
|-----------------|--------------------|-----------------|-----------------------------|-----------------------------|---------------------|
| Phantom 3       | Vertical Images    | 0.007           | 0.064                       | 0.80                        | 0.9957              |
|                 | 70 degrees Images  | 0.061           | 0.065                       | 1.21                        | 0.9935              |
| Phantom 4       | Vertical Images    | 0.075           | 0.061                       | 1.02                        | 0.9945              |
|                 | 70 degrees Images  | 0.075           | 0.086                       | 1.21                        | 0.9936              |
| Laser Scan      | 3 shooting points  | 0.647           | 0.232                       | 3.03                        | 0.9859              |
The average error might hide further distribution discrepancies and it is used to check bias occurrences. Comparing similar flights, the discrepancies between Phantom 3 and Phantom 4 are related to different daylight reflectance, slightly different cameras, sampling grid and the equipment accuracy itself. The histograms of errors between the original elevations of the RTK points and the same points projected onto each resulting surface are presented in Figure 3.

### 3. Results and discussion

It is noteworthy that the surface built from the laser-scanned point cloud shows the poorest adherence when compared to the other results. Laser scanning has been widely used in the last decade and it is recommended for many precision applications with broadly validated method. But despite being made from three different shoot bases, the laser scan application suggests that additional shoot bases would attenuate the extreme discrepancies. The long distances and the differences in height possibly shaded some areas such as inaccessible bench berms. However, the process of adding point clouds is time and computational consuming.

A visual check was performed to validate Table 1 and the RTK points are clearly closer to the two DSMs with the lowest statistical deviations. It is also noticeable that the laser-scanned points produced a surface with high local variability due to small disturbances. Figure 4 shows a vertical section with the five analysed surfaces and their visual differences from the RTK hard data.

![Histogram of differences between the elevations of the RTK points and their equivalent location at the resulting surface.](image)
The biggest distortions for the UAV surfaces are observed in planar areas and the oblique photos potentiate this effect. The results suggest that vertical photogrammetry produces more accurate surfaces for the same equipment and conditions because every statistic analysed shows better fit for the DSMs generated with images acquired vertically, when compared with oblique photos taken at 70 degrees.

Therefore, the most adherent surfaces were originated from the vertical missions of both UAVs, which showed very similar statistical and visual results. The vertical photos from Phantom 3 show smaller average error and minimum mean square deviation and a higher slope of regression (Table 2). Hence, from those DSMs, the one built from Phantom 3 vertical photos was selected as the most accurate and it was used for further comparisons with all other surfaces, made in pairs with the one taken as the reference.

James et al. (2017) used Monte Carlo simulations to quantify the uncertainty related to the surface models. Although there are other publications that make suggestions to minimize systematic errors, the intention presented here is to show the divergencies among techniques for quick and low-cost topographical surveying to be applied by small mining companies. The objective is not to correct errors, but to assess the discrepancies and highlight the effects of settings.

To compare the elevations obtained with each flight mission, it was necessary to regularize the point clouds obtained. Using a script developed to compare pairs of grids (each of them representing a surface), the surfaces were sampled and compared at regularly spaced grids (1 m x 1 m), so that the elevation was the only variable for the sampled points from this regular mesh. The algorithm uses the XY plan to needle the surfaces and record the Z interception value. Figure 5 shows the sampled regular grid from the surface originated with Phantom 3 and vertical photos.
The comparison between the reference surface and the one built from the scanned points clearly shows the largest discrepancies. This fact is attributed to the laser scanner limitations at large distances, typical of open pit areas. Even scanning from different bases, many areas show a sparse number of points, due to eclipsing and/or longer distances from the emission bases. Areas with the largest discrepancies are the same ones with a lesser number of data for the laser scanner and, consequently, longer triangle edges.

Moving objects is another source of differences, even when considering the two closest results (P3 and P4 vertical photos). Flight missions made with a time-lapse difference detected vehicles and other moving equipment at different locations. The areas with these differences could be manually classified in the software and were not processed, but for the purpose of showing and justifying the impacts, no filter was applied. Vertical or sub-vertical walls also tend to result in local discrepancies for vertical distances between surfaces. This explains how buildings, poles and slope faces present higher divergences.

For the images previously shown in Figure 6, the statistical comparisons between the selected DSM (Phantom 3 Vertical) against all the others are shown in Figure 7, the histograms of the differences were also limited to show the most important deviations, as the outer values represent moving objects, sub-vertical walls, vegetation, etc.

With the mean of 0.03 m, the differences between both surfaces from vertical imagery showed the smallest standard deviation (0.30 m), which indicates that areas with a lesser pair of points have big discrepancies and those with higher values are in vegetation or shaded portions, as previously observed in Figure 6. For the oblique photos, the resulting DSMs show higher and more consistent biased errors. The charts were truncated to evidence the differences around zero. However, extreme discrepancies were
observed in all surfaces, mainly in vegetation, suspended objects, shaded and flooded areas that could not be distinguished from others.

4. Conclusions

The DSMs created from two aircrafts in two different configurations each, showed acceptable discrepancies between themselves. When compared against reliable data acquired from traditional topography surveying, the surfaces showed different levels of deviation in elevation and the photos acquired in missions with vertical cameras showed more consistent surfaces than the oblique ones. In addition, the surface built from points surveyed with laser scanner was more erratic and showed several extreme error values in comparison with the most accurate surface.

The laser-scanned points were affected by occlusions due to slope crests or other objects, water reflection and even suspended dust. To correct these problems, high performance computers, more shooting points and a huge number of cloud generated by photogrammetry was as dense as the LS and the points received their respective RGB colours from the images taken. The high quality outputs, such as orthophoto, point cloud, contours, DSM, photos itself and movies for thematic purposes can be quickly produced.

The small aircrafts and cameras used in this study showed better performance when compared to traditional methods and/or the more sophisticated techniques. The differences observed in the elevations were within the acceptable level of variation given that the mining location is an area of constant movement of material and there is a large amount of irregularities and loose material in the pit surface. For further work, it is suggested that the resulting quality, processing speed, operational implementation and products can be investigated using different techniques.

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