CALIBRATION OF A MULTI-CAMERA ROVER

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ABSTRACT:

Multi-Camera-Rover are recently coming up for usual terrestrial surveying tasks. This technique is new for the surveyors. Although photogrammetric specialists realize the benefits of such systems immediately, surveyors have difficulties to find efficient usages. To approach this new measurement systems the technique has to be understood and the confidence on the accuracy has to grow.

In this paper we analyze the accuracy of a Multi-Camera-Rover using an indoor testfield by photogrammetric algorithms. The results show that the knowledge of the interior orientation parameter of the cameras and the relative orientation of the cameras is essential for precise geometric reconstructions. Knowing these additional data, high accurate results become possible.

1. INTRODUCTION

1.1 Motivation

Close-range photogrammetry has been a specialised working field in surveying and geoinformatics offering only loose connections with traditional terrestrial measurement techniques like tacheometry, height levelling and terrestrial laserscanning. Mostly the combination of both fields is done by tacheometric measured 3d pass points which can be identified in the digital images. Also, some automatic matching between photogrammetric images and terrestrial laserscanning data has been done. The major difficulty in combination of these different methods is the usage of solitary off-the-shelf digital cameras in the past. The modelling of a constant offset between the perspective center of the digital camera and e.g. several fixed exterior marks on the camera body is only seldom possible when a fixed focus camera or a fixed camera rig is used (e.g. in industrial applications).

With the new multi-camera rover (MCR) systems consisting of several hard mounted fixed focus cameras a close integration of close-range photogrammetry in tacheometric applications becomes easily possible when a tacheometer tracks the position of the camera or a GNSS receiver is mounted on-top of the MCR. Working with MCR is a new category of measurement techniques. Therefore new methods for calibration are necessary to ensure precise results in day-to-day applications.

This paper focuses on three important accuracy aspects concerning the camera system of a MCR which are: i) the interior orientation of each mounted camera ii) the relative orientation of all mounted cameras and iii) the offset to the exterior target mark used for tracking and positioning of the MCR. These aspects have to be handled in a sound calibration. The calculation of the interior parameters of all cameras, the parameters of the relative orientation and the 3d-offset of the system relative to the exterior target is done by bundle-block adjustment.

The method will be presented and applied for a Trimble V10 MCR. All calibration measurements are performed in the 3d-calibration test field of the photogrammetric laboratory of the FHWS which consists of 93 coded targets with known precise 3d coordinates. The results of all tests will be shown and compared to the data given by the manufacturer. Additional sensors included in the MCR (e.g. a tilt sensor or a magnetic sensor) are beyond the presented work.

1.2 Multi-camera rover

3d visualization is the main task of recently presented MCR. A few years ago Fujifilm brought to market two 3d cameras with two lenses systems. It was a new early try to provide 3d image data and 3d-video data to a broader community. Multi-camera systems are often fixed on a ring. They differ in the direction of the cameras. In industrial application the viewing direction is to the inside (e.g. 3d scanning of persons (TEN24 MEDIA LTD, 2016)), in 3d animation and topographic applications to the outside.

Outside directed multi camera systems are the Trimble V10 (Trimble Navigation, 2016) and the GoPro Odyssey system (GoPro, 2015). The Trimble V10 system seems to be in a continuous production. This system has twelve cameras at two height levels and vertical directions. The GoPro Odyssey systems seems to be only available in a small number. It consists of sixteen synchronized HERO4 Black cameras. (VideoStitch, 2016) announced recently a system of five EOS M3 Canon cameras with custom Samyang Fisheye lenses.

GoPro and VideoStitched obviously concentrate on the 3d-visualisation and 3d-gaming market. High geometric precisions are not necessary. Trimble V10 build a measuring instrument which has in some parts the capabilities of basic tacheometers. This measuring task makes a precise knowledge on the sensor system necessary, on the interior parameters of the cameras and the spatial calibration of the cameras in the system, in combination to other sensors. Two typical measuring situations are shown in fig. 1.

The V10 is at the moment a closed Trimble solution in the working process, because exact information on the sensor geometry is not easily accessible. In this article we try to gain some of this information by photogrammetric methods. In parallel we analyse the inherent system precision of the V10. The presented method
The overall size of the sensor plane is

\[ w \times h = (1944s) \times (2592s) = 2.752mm \times 3.669mm \] (1)

It is a 4 : 3 image format. A full frame camera with a width of 36mm would have a 35mm lens (wide angle lens) with a pixel size of 13µm. NIKON D800 have 4.9µm pixel size.

(Köhler, 2014), (Shams, 2016) and (Whitehead, 2014) give background information how Trimble does the calibration of the cameras (with a tilting plane and collimator) and the determination of the interior orientation (checkerboard pattern room). Trimble itself calculates the principle point, the camera constant, the radial symmetric distortion up to 7th order and the non-symmetric distortion; magnitude and direction. All resulting residuals in image coordinates are below one pixel.

The V10 user manual (Trimble, 2016) names an accuracy of the V10 of 0.4mm at a distance of 1m, 4mm in 10m and 40mm in 100m.

1.4 Outline of the article

The main topic of this article is, how the V10 can be used as a MCR for different software packages than the Trimble Business Center. In this case, only the image data is available at first sight. Additional EXIF-data in the jpg-image files of the V10 is available, but just a few parameter which are quite useless.

There are five main items to address:

- the stability of each camera itself (test 1),
- the stability of each camera itself within movement (test 2),
- the 3d-reconstruction accuracy for 3d-points just using one camera, in a spatial forward section (test 3),
- the stability of the relative orientation between several cameras (test 4) and
- the offset of the camera projection centers in relation to the center of a mounted target (test 5).

These efforts lead to five tests:

- 10 repetitions of images of a specific cam, here “cam 4”,
- 10 repetitions of images of a specific cam, here “cam 4” with usual movements between the shots,
- two images of each camera performing a spatial forward section,
- combined images in pairs and triples of all neighboring cameras to get the relative orientations and
- calibration of some cameras relative to a mounted target.

In the next section (cf. sec. 2) the applied mathematical methods are described. In section 3 results of the tests for the V10 are shown. The article ends with the conclusions and an outlook (cf. sec. 4).

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2. METHOD

2.1 Camera orientation

2.1.1 Exterior orientation The main equation for all analysis is the projection equation of a 3d-point onto the image plane. In homogeneous coordinates this can be written by (Faugeras and Luong, 20011)

\[ x_1 = Hx \]  

(2)

with \( H \) the projection matrix, \( x_1 = (x_1, y_1, z_1)^T \) the vector of ideal image coordinates and \( x = (X - X_o, Y - Y_o, Z - Z_o, 1)^T \) object coordinates. The projection matrix \( H \) is

\[ H = \begin{pmatrix} r_{11} & r_{12} & r_{13} & 0 \\ r_{21} & r_{22} & r_{23} & 0 \\ r_{31} & r_{32} & r_{33} & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \]  

(3)

For projected image points \( x_1 \) holds \( z_1 = -c. \) This yields to the fundamental equation of photogrammetry

\[ x_1 = -r_{11}(X - X_o) + r_{12}(Y - Y_o) + r_{13}(Z - Z_o) \]
\[ r_{13}(X - X_o) + r_{23}(Y - Y_o) + r_{33}(Z - Z_o) \]
\[ y_1 = -r_{12}(X - X_o) + r_{22}(Y - Y_o) + r_{23}(Z - Z_o) \]
\[ r_{13}(X - X_o) + r_{23}(Y - Y_o) + r_{33}(Z - Z_o) \]

(4)

in cartesian coordinates with the ideal image point \( x_1 = (x_1, y_1) \) the focal length \( c, \) the projection center \( X_o = (X_o, Y_o, Z_o), \) and the elements of the rotation matrix from the camera coordinate system to the image coordinate system (cf. tab. 2). The projection center and the rotation matrix are called the “exterior orientation”.

2.1.2 Interior orientation Modeling the interior orientation of a camera using the interior orientation parameters is essential to achieve high reconstruction accuracies. We use the following parametrization (Technot GmbH, 2010). \((x_1, y_1)\) are the ideal image coordinates, considering an error-free lens system (pinhole-camera model). These coordinates are corrected by the radial-symmetric correction

\[ x_2 = x_1 + x_1 A_1 (r^2 - r^2_o) + x_1 A_2 (r^4 - r^4_o) \]
\[ y_2 = y_1 + y_1 A_1 (r^2 - r^2_o) + y_1 A_2 (r^4 - r^4_o), \]

(5)

with \( r^2 = x^2_o + y^2_o, \) the radial-asymmetric correction and principal point

\[ x_4 = x_o + x_2 + 2 A_3 x_1 y_1 + A_6 (r^2 + 2x_1^2) \]
\[ y_4 = y_o + y_2 + 2 A_6 x_1 y_1 + A_5 (r^2 + 2x_1^2), \]

(6)

and the affine correction

\[ x' = x_4 + x_4 A_3 + y_4 A_4 \]
\[ y' = y_4 - y_4 A_3 + x_4 A_4. \]

(7)

2.1.3 Bundle adjustment All parameters of the exterior orientation and the parameters of the interior orientation are estimated in a bundle adjustment. The parameter estimation is done by Least-Squares-Adjustment (LSA) including an outlier detection. If \( y \) is the vector of observations (e.g. image coordinates) with an a priori known covariance matrix \( D(y) = P^{-1}, \) if \( \beta \) is the vector of unknown parameters and if \( X \) is the matrix of partial derivatives of the observation equations (Schwefel and Ackermann, 1976), the estimated vector of the unknowns \( \hat{\beta} \) follows by

\[ \hat{\beta} = (X^T PX)^{-1} (XPy) \]  

(8)

with its covariance matrix

\[ \hat{D}(\hat{\beta}) = \hat{\sigma}_n^2 (X^T PX)^{-1} \]  

(9)

and

\[ \hat{\sigma}_n^2 = (\hat{\beta} - y)^T (\hat{\beta} - y). \]  

(10)

We assume for all measured image coordinates the same a priori standard deviation \( \sigma_n. \) Then the estimated \( \hat{\sigma}_n^2 \) is an important value to show the overall system accuracy. All estimated \( \hat{\sigma}_n \) have to be multiplied by the a priori \( \sigma_n \) to get the image point measurement accuracy. The high accuracies are achieved because of the automatic detection and measuring of the coded targets. Therefore the accuracy of the point measuring process can be neglected.

To study the impact of a correction on a pixel coordinate, the correction itself is calculated. The impact of the standard deviation of the parameters results from the error propagation law

\[ D(x) = AD(\beta)A^T \]

with

\[ \hat{\beta} = (x_0, y_0, A_1, A_2, A_3, A_4, A_5, A_6)^T \]

(11)

and

\[ A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ x_1 (r^2 - r^2_o) & y_1 (r^2 - r^2_o) & y_1 (r^2 - r^2_o) \\ x_4 & -y_4 & x_4 \\ y_4 & x_4 & 2x_1 y_1 \\ (r^2 + 2x_1^2) & 2x_1 y_1 & (r^2 + 2x_1^2) \end{pmatrix} \]

(12)

This is scaled by the image scale into the object space. The scale depends on the object distance.

2.2 Calibration

2.2.1 Relative orientation of the cameras The bundle adjustment provides exterior orientation parameters for all camera positions. Let us just take two of them of the same moment, say \( a \) and \( b. \) Both camera positions define camera coordinate systems \( K_a \) and \( K_b. \) The exterior orientations for all cameras include the rotation matrix \( R_a^b \) and \( R_b^a, \) which are the rotations from the object coordinate system into the specific camera coordinate system. The rotation from \( K_a \) into \( K_b \) is follows from

\[ R_b^a = R_b^o R_a^o = R_b^o (R_a^o)^{-1} = R_b^o (R_a^o)^T. \]  

(13)

Assuming only vertical or horizontal rotations the rotation can be derived easily: The column \( (r_1)^b \) of the rotation matrix is the rotated \( z- \)axis which is indeed the main viewing direction of the camera. The rotation angle results from the scalar product with the unity vector \( e_2 = (0, 1, 0)^T \)

\[ \sin \kappa = \frac{((r_1)^b \cdot e_2)}{||r_1||_2 ||e_2||} = (r_1)^b \cdot e_2 = (r_1)^b \]  

(14)

for the horizontal viewing cameras “cam1” to “cam7”.

2.2.2 Offset of a mounted target to the cameras To determine the offset between the camera projection centers and a trackable position, we add a mounted target on top of the V10. This

1 Vectors and matrices are written in bold letters.


\[
\mathbf{R}_w = \begin{pmatrix}
    r_{11} & r_{12} & r_{13} \\
    r_{21} & r_{22} & r_{23} \\
    r_{31} & r_{32} & r_{33}
\end{pmatrix} = \begin{pmatrix}
    \cos \phi \cos \kappa & \sin \omega \sin \phi \cos \kappa + \cos \omega \sin \kappa & \cos \omega \sin \phi \sin \kappa + \sin \omega \cos \kappa \\
    \sin \phi \sin \kappa - \cos \phi \cos \kappa \sin \omega & - \sin \omega \sin \phi \sin \kappa + \cos \omega \cos \kappa & \cos \omega \cos \phi \sin \kappa + \sin \omega \sin \kappa \\
    \cos \phi \sin \kappa & - \cos \omega \sin \phi \cos \kappa & - \sin \omega \sin \phi \sin \kappa + \cos \omega \cos \kappa
\end{pmatrix}
\]

(15)

Table 2: Rotation matrix with \( \omega \) rotation on \( x \)-axis, \( \phi \) rotation on \( y \)-axis and \( \kappa \) rotation on \( z \)-axis, from local camera system to object system.

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To calculate the interior orientation parameters some values have to be fixed (see tab.3): the sensor size, \( r_s \), and the a a priori \( \sigma_o \).

| Sensor size \((h \times w)\), see eq.1 | \(2.752\text{mm} \times 3.660\text{mm}\) |
|--------------------------------------|-----------------------------------------|
| \(r_s\) (see eq. 5)                  | 1.529                                   |
| A priori \(\sigma_o\)                | \(5\mu m\)                              |

Table 3: Fixed interior parameters for all cameras.

target is measured by some NIKON D800 images. Therefore, in this case, the target is realised as a coded photogrammetric target. The offsets follow from the Euklidean distance

\[
s_{oi} = \sqrt{(X_{oi} - X_i)^2 + (Y_{oi} - Y_i)^2 + (Z_{oi} - Z_i)^2},
\]

when \((X_{oi}, Y_{oi}, Z_{oi})\) are the coordinates of the reection center of the camera “cam i” and \((X_i, Y_i, Z_i)\) are the coordinates of the center of the mounted target.

3. RESULTS

The V10 has twelve cameras. The numbering of the cameras is given in fig. 3. A typical test image is shown in fig. 4. In the next section we are going through all experiments.

3.1 Test 1: Stability of each camera itself

In the first test (series A with 10 images, no movement of the camera) the repeatability of the determination of the interior camera parameter is in our focus. We use camera 10 (front camera) for this test. Without any correction of the pixel coordinates \(\sigma_o = 4.23\) which means the accuracy of the pixel coordinates is approximately four times worse than our a priori assumption of \(0.08\mu m\).

Assuming a constant camera for all ten images results in \(\sigma_o = 0.0470\) which is a reduction by the factor 100. This shows the feasibility of the calibration model. Due to effects of the robust parameter estimation no improvement can be achieved by modeling a separate camera for each image. An arbitrary set of interior calibration parameters is in fig. 4.

A second series of 10 images (series B, no movement of the camera) verified this observation, \(\sigma_o = 4.126\), \(\sigma_o = 0.0384\) and \(\sigma_o = 0.0550\) resp. All estimation results can be found in table 11 at the end of this article.

3.2 Test 2: Stability of a camera itself within moving

The second test focuses on the stability under movement of a camera of the V10. Again we use camera 10 (front camera) for this test. Without any improvement of the pixel coordinates a \(\sigma_o = 4.687\) results. Assuming a constant constant camera for all images leads to \(\sigma_o = 0.0436\). In the case of a separate interior orientation the estimated variance is \(\hat{\sigma} = 0.0565\). A parameter set of the interior orientation of an arbitrary images is shown in table 5. All estimation results can be found in table 12 (upper part) at the end of this article. In comparison to the tests in section 3.1 both results prove each other.

Summa summarum, both test show, that the front camera can be assumed as a fixed camera model with a constant interior orien-
In this section we address the question how the different cameras of the V10 should be handled for photogrammetric reconstruction. A series of image pairs of all cameras has been taken. All exterior parameters have been calculated by parameter estimation with different cases in modelling the interior orientation parameters. In table 6. In this case the interior orientation of each camera has a separate parameter set of interior orientation factors. This can be calculated from V10 testfield images (s. 5).

Again, the results of this test show an improvement of the accuracy. $\sigma_{\hat{o}}$ diminishes from $\sigma_{\hat{o}} = 4.7049$ to $\sigma_{\hat{o}} = 0.0932$ introducing one single camera correction model for all cameras. When each camera has a separate parameter set of interior orientation parameters it goes down to $\sigma_{\hat{o}} = 0.0549$, an improvement by factor approx. 2. All estimation results can be found in table 12 (lower part) at the end of this article.

To give an impression of the achievable accuracy the point 66 has been reconstructed from all image pairs. The results are listed in table 6. In this case the interior orientation of each camera is modelled separately. For object distances of approx. 1m a reconstruction accuracy far below one millimeter can be achieved with simultaneous calibration. Up to this point we just focused on single cameras. The next tests take the combination of distinct cameras into account.

### Table 5: Results of the interior orientation of test 2 with one camera model for all images (see sec. 3.2)

| No. | X    | Y    | Z    | $\sigma_X$ | $\sigma_Y$ | $\sigma_Z$ |
|-----|------|------|------|------------|------------|------------|
| 66  | 3365.5 | 1459.3 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6601 | 3365.5 | 1459.4 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6602 | 3365.5 | 1459.4 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6603 | 3365.4 | 1459.4 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6604 | 3365.6 | 1459.4 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6605 | 3365.5 | 1459.4 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6606 | 3365.6 | 1459.5 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6607 | 3365.5 | 1459.4 | 839.9 | 0.2         | 0.2        | 0.1        |
| 6608 | 3365.6 | 1459.4 | 839.9 | 0.4         | 0.3        | 0.2        |
| 6609 | 3365.6 | 1459.5 | 839.9 | 0.3         | 0.2        | 0.1        |
| 6610 | 3365.6 | 1459.4 | 839.9 | 0.3         | 0.2        | 0.1        |
| 6611 | 3365.6 | 1459.4 | 839.8 | 0.3         | 0.2        | 0.1        |
| 6612 | 3365.7 | 1459.3 | 839.9 | 0.2         | 0.2        | 0.1        |

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Table 9: Resulting coordinates of the projection centers of two cameras.

| Camera | X (μm) | Y (μm) | Z (μm) |
|--------|--------|--------|--------|
| cam2   | 2612.233731 ± 0.8 | 1070.777768 ± 2.0 | 1044.877087 ± 0.9 |
| cam3   | 2645.438123 ± 2.3 | 1051.223378 ± 1.7 | 1045.625319 ± 0.7 |

Table 10: Resulting distances of the offset calibration between two projection centers and the center of the mounted target (see sec. 3.5).

| From     | To     | Distance |
|----------|--------|----------|
| Test cam1-cam2 | cam1 | 199.3mm |
| Test cam1-cam2 | cam2 | 199.3mm |
| Test cam2-cam3 | cam2 | 199.0mm |
| Test cam2-cam3 | cam3 | 199.7mm |

Nevertheless with appropriate camera modeling, including the interior orientation parameters of all cameras acceptable accuracies for a wide range of applications can be achieved (e.g. cadastral surveying).

The test showed that the effort is quite high to get inside the technique of a MCR. Specialized software is needed to handle of necessary parameters (lense distortion, relative orientation and offsets). Trimble provides this for their users in an almost “black box”. System checks are still quite superficial. The future will show the necessity of deep system checks. A freeware software to handle MCR would be interesting for independent results of V10 data and of the data of different MCR.

These tests of a MCR are just the starting point of a series of research: for a complete description of a MCR, a calibration cage for panorama sensors and a combination of panaroma photogrammetry to terrestrial laser scanning are envisaged.

The planning of a complete surveying campaign is still challenging, although Trimble gives its users a lot of hints for their work. Starting on known survey points for taking panaromas might be a solution to approach this new measuring techniques. An other example for a starting project with fixed tacheometric control points is given in fig. 6 for a small object (birdhouse) reconstruction.

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Figure 6: Sample indoor scenario for the photogrammetric reconstruction of a small object using known control points (top: scene plan; bottom: reconstruction).

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Table 11: Results of test 1, series A and B: one camera ("cam 4") used, no movement between single shots; for each image a separate camera model has been estimated, two independent series.
Table 12: Results of test 2: one camera (“cam 4”) used, usual movement between shots; for each image a separate camera model has been estimated. Results of test 3: 12 cameras (“cam 1” to “cam 12”); two images of each camera; for each camera a separate camera model is estimated.