VERTICAL ACCURACY EVALUATION OF ASTER GDEM2 OVER A MOUNTAINOUS AREA BASED ON UAV PHOTOGRAFMETRY

Yubin Liang a, b, *, Yuankun Qu a, b, Dongxu Guo a, b, Tiejun Cui a, b
a School of Geographic and Environmental Sciences, Tianjin Normal University, Tianjin 300387, China
b Tianjin Engineering Center for Geospatial Information Technology, Tianjin Normal University, Tianjin 300387, China
lyb.whu@gmail.com; lily9961@qq.com; 1015881693@qq.com; tiejun_cui@163.com

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ABSTRACT:

Global digital elevation models (GDEM) provide elementary information on heights of the Earth's surface and objects on the ground. GDems have become an important data source for a range of applications. The vertical accuracy of a GDEM is critical for its applications. Nowadays UAVs has been widely used for large-scale surveying and mapping. Compared with traditional surveying techniques, UAV photogrammetry are more convenient and more cost-effective. UAV photogrammetry produces the DEM of the survey area with high accuracy and high spatial resolution. As a result, DEMs resulted from UAV photogrammetry can be used for a more detailed and accurate evaluation of the GDEM product. This study investigates the vertical accuracy (in terms of elevation accuracy and systematic errors) of the ASTER GDEM Version 2 dataset over a complex terrain based on UAV photogrammetry. Experimental results show that the elevation errors of ASTER GDEM2 are in normal distribution and the systematic error is quite small. The accuracy of the ASTER GDEM2 coincides well with that reported by the ASTER validation team. The accuracy in the research area is negatively correlated to both the slope of the terrain and the number of stereo observations. This study also evaluates the vertical accuracy of the up-sampled ASTER GDEM2. Experimental results show that the accuracy of the up-sampled ASTER GDEM2 data in the research area is not significantly reduced by the complexity of the terrain. The fine-grained accuracy evaluation of the ASTER GDEM2 is informative for the GDEM-supported UAV photogrammetric applications.

1. INTRODUCTION

In the present era of big data, spatial data collection modes are characterized with the progressive course of diversity, efficiency and facilitation, which are producing large sets of data. This further brings the request for advanced data processing with higher level of efficiency, automation and intelligence (Zhang and Tao, 2017). In the past decade, significant advances in global elevation modelling have been made with the release of ASTER (Advanced Spaceborne Thermal Emission and Reflection Radiometer) GDEM and SRTM (Shuttle Radar Topography Mission) GDEM. Global digital elevation models (GDEM) provide elementary information on heights of the Earth's surface and ground objects, which have become an important data source for a range of applications such as surveying and mapping, gravity-field modelling, hydrological studies, flood simulation, emergency response and many more (Teo et al., 2010). The vertical accuracy of a GDEM is critical for its applications. In the field of UAV photogrammetry, flight planners usually consider the average elevation of the whole surveying area. However, the relative flight heights of UAVs are usually lower than those of manned aircrafts. The complexity of the terrain may result in serious occlusions and under-overlap due to low flight heights. Footprints and Voronoi diagrams have been used to generate image pairs for image matching. Footprint-based method generated the footprints images and then determined the image pairs by footprint intersection (Rupnik et al., 2014). The terrain is usually assumed planar for this kind of methods. If the terrain is complex, the footprints will not lie on a plane, which makes it difficult to determine the intersections. DEMs can also be used for direct georeferencing. DEM-based direct georeferencing usually starts from an average elevation and iterates to converge. However, the iterative method may stop at a local optimal with bad initialization due to complex terrain and inaccurate Positioning and Orientation data. All of these applications including flight planning, plan evaluation, image matching, direct georeferencing and ortho-mosaicing can be improved by using simulation methods with GDEM data (Liang et al., 2017). However, the accuracies of the simulations are largely dependent on the vertical accuracy of the GDEM data used.

Since 2000 the Japanese ASTER instrument, payload on NASA's Terra satellite, acquires stereo image data with its two nadir- and backward-viewing telescopes, which are sensitive in the near infrared spectral band. The joint Japanese–US ASTER GDEM version 2 was released in October 2011 by NASA and the Ministry of Economy Trade and Industry (METI) of Japan. ASTER GDEM2 covers most of the earth surface and is freely available all over the world. The ASTER GDEM data is acquired through stereo observations and requires a large number of accurate ground control points to ensure its worldwide accuracy. With high spatial resolution (~30m), the ASTER GDEM2 has been widely used. The basic features of ASTER GDEM2 are listed in (Tachikawa et al., 2011a).

Generally, high-precision ground control points are used to evaluate the accuracy of GDEM products. There have existed a
number of researches that evaluated the accuracies of the GDEMs. Hirt et al. (2010) compared ASTER GDEM (version 1), the SRTM DEM (version 4.1) and the GEODATA DEM-9S (version 3), and evaluated them using 6392 levelling and 911 GPS/levelling ground control points.

The second version of ASTER GDEM is reported to have improved significantly with respect to its predecessor in terms of vertical height bias, striping error and voids that have been filled to some extent (Krieger et al., 2010; Carabajal, 2011; Gesh et al., 2011; Tachikawa et al., 2011a). In a summarising study by the joint Japan–US ASTER Science Team comprising a total of four independent validation studies, the vertical accuracy of ASTER GDEM2 is estimated to be around 17 m at a confidence interval of 95% (Tachikawa et al., 2011b). Rexer and Hirt (2014) evaluated the accuracy of the ASTER GDEM2 and SRTM v2.1/v4.1 over Australia using ground truth data from the Australian National Gravity Database. The research found that the elevation accuracy of these datasets were at the level of meters. However, measuring ground control points is costly and the ground sample interval is relatively low. Although ASTER GDEM2 has an overall accuracy of around 17m at the 95% confidence level, the accuracy of ASTER GDEM2 in a certain area is affected by the observation conditions, the complexity and land cover of the terrain, and the number and distribution of control points.

Nowadays UAVs has been widely used for large-scale surveying and mapping (Haala et al., 2011; Remondino et al., 2011; Colomina and Molina, 2014). Compared with traditional surveying techniques, UAV photogrammetry are more convenient and more cost-effective. UAV photogrammetry produces the DEM of the survey area with high accuracy and high spatial resolution. As a result, DEMs resulted from UAV photogrammetry can be used for a more detailed and accurate evaluation of the GDEM product.

This study investigates the vertical accuracy (in terms of elevation accuracy and systematic errors) of the ASTER GDEM Version 2 dataset over a complex and vegetation-covered terrain. The accuracy of the ASTER GDEM2 data is evaluated with a high resolution DEM produced by UAV photogrammetry. Section 2 details the methodology of the study. Experimental results are discussed in Section 3 and conclusion is made in Section 4.

2. METHODOLOGY

The production of the DEM by photogrammetry mainly consists of image matching, relative orientation, absolute orientation, dense matching and raster DEM gridding. In the image matching procedure, the image pairs are generated by using a simulation method. A set of points were sampled from the ASTER GDEM2 data and then transformed to the East-North-Up (ENU) coordinate system. The positions of the images are also transformed from the geodetic coordinate system to ENU. The orientations of the images are used to construct the rotation matrix R defined in the collinearity equation (equation (1)). The collinearity equation is a mathematical model which projects a three dimensional point P onto the image plane. Figure 1 illustrates the geometric configuration of perspective projection defined by the collinearity equation.

\[
R = \begin{bmatrix}
    a_1 & a_2 & a_3 \\
    b_1 & b_2 & b_3 \\
    c_1 & c_2 & c_3 
\end{bmatrix}
\]  

(2)

where

\[
\begin{align*}
a_1 &= \cos \varphi \cos \kappa - \sin \varphi \sin \omega \sin \kappa \\
a_2 &= -\cos \varphi \sin \kappa - \sin \varphi \sin \omega \cos \kappa \\
a_3 &= -\sin \varphi \cos \omega \\
b_1 &= \cos \varphi \sin \kappa \\
b_2 &= \cos \varphi \cos \kappa \\
b_3 &= -\sin \varphi \sin \omega \\
c_1 &= \sin \varphi \cos \kappa + \cos \varphi \sin \omega \sin \kappa \\
c_2 &= \sin \varphi \sin \kappa + \cos \varphi \sin \omega \cos \kappa \\
c_3 &= \cos \varphi \cos \omega 
\end{align*}
\]  

(3)

where \( \varphi \) is the primary rotation about the Y-axis, \( \omega \) is the secondary rotation about the X-axis, and \( \kappa \) is the tertiary rotation about the Z-axis.
\[
\begin{align*}
\text{col} &= \text{col}_{pp} - \frac{y}{d} \\
\text{row} &= \text{row}_{pp} + \frac{x}{d}
\end{align*}
\] 

where \((\text{col}_{pp}, \text{row}_{pp})\) is the position of the principal point under \(o'-\text{colrow}\), \((x, y)\) is the position of \(p\) under \(o-\text{xy}\), and \(d\) is the pixel size.

The image observations of the ground sample points are simulated by using the collinearity equation. The images with projections of the same ground sample points are considered as overlapped. The image pairs generated with the POS and DEM data are accurate and false image pairs are effectively filtered.

The tie points generated by the image matching procedure are then used for relative orientation. After the relative orientation, the positions and orientations of images under a local coordinate system are determined by an incremental bundle block adjustment procedure.

The result of the relative orientation and the coordinates of the GCPs are then used for absolute orientation. After the absolute orientation, the positions and orientations of images under ENU coordinate system are determined by a bundle block adjustment.

The dense matching is then carried out to generate the dense cloud of the research area. The Triangulated Irregular Network (TIN) is generated from the resulting dense point cloud. On the basis of the TIN, a raster DEM is interpolated and output as a GeoTIFF file.

To evaluate the vertical accuracy of the ASTER GDEM2 data, the DEM produced by UAV photogrammetry is down-sampled to the resolution of ASTER GDEM2 data. Then pixel-wise errors are derived from the ASTER DEM and the UAV DEM. The RMSE is then derived from the errors. The slope map of the research area is derived from the UAV DEM. The correlation matrix between the errors of ASTER DEM and the slopes is calculated and analyzed. In this study, the correlation matrix between the errors of ASTER DEM and the stacking numbers is calculated and analyzed.

The ASTER GDEM2 can be used for the direct georeferencing and ortho-mosaicing. In these applications, the ASTER DEM data is usually up-sampled. This study also evaluates the vertical accuracy of the up-sampled ASTER DEM. The ASTER DEM is up-sampled to the resolution of UAV DEM. Then pixel-wise errors are derived by subtracting the UAV DEM from the ASTER DEM. The RMSE is then derived from the errors and compared with the aforementioned down-sampled result.

3. RESULTS AND DISCUSSION

The rectangle-like research area is 6 km from east to west and 7 km from south to north (Figure 3). The elevation of the area is in the range of [200, 1050] meters. Most of the research area is covered by shrubs and trees.
Figure 6 shows the camera locations and number of image observations of the research area. It can be seen that most of the research area can be observed by more than 9 images.

Seven ground control points (K1-K7) were measured under WGS84 coordinate system by using the static networked GPS and the accuracy of the ground control points were at the level of decimetres. Figure 7 shows the positions of the GCPs.

The accuracy of the aerial triangulation at control points after the bundle block adjustment is given by Table 1. The maximum absolute Z error is 0.703m and the RMSE along Z axis is 0.278m. It can be seen that the vertical accuracy of the aerial triangulation is at the level of decimetres, which is accurate enough to evaluate the ASTER GDEM2 data.

To evaluate the vertical accuracy of the ASTER GDEM2 data, the DEM produced by UAV photogrammetry is down-sampled to 30m by nearest neighbor interpolation.

The elevation values at the edges of the DEM are not reliable due to few image observations (Figure 6). Therefore, the UAV DEM is masked to clip the edges. After the masking process, pixel-wise errors are derived by subtracting the UAV DEM from the ASTER DEM. Figure 9 shows the spatial distribution of the errors.

The maximum and the minimum error are 100.43m and -88.30m respectively. The histogram of the errors is shown in Figure 10. Statistical test shows that the errors come from a...
normal distribution with a mean of 0.06m and a standard deviation of 15.04m, which indicates that the systematic error is quite small. The RMSE is 15.04m, which coincides well with the 15.85m accuracy reported by (Tachikawa et al., 2011a).

The stacking map of the research area is derived from the QA file with the ASTER GDEM2 data (30m resolution). The stacking map is shown in Figure 12. It can be seen that the stacking number of most areas is larger than 4.

Table 2 shows the statistics of the errors of ASTER GDEM and the stacking numbers. The maximum and the minimum stacking number are 7 and 1 respectively. The mean and the standard deviation of the stacking numbers are 5.8178 and 0.8690. The correlation coefficient between the errors of ASTER DEM and the stacking numbers is -0.009, which indicates that the accuracy of ASTER GDEM2 is negatively correlated to the number of stereo observations. The errors tend to be large in the areas that less observed and small otherwise.

Table 3. Statistics of the stacking numbers and the errors

<table>
<thead>
<tr>
<th>Layer</th>
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<tr>
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<td>Errors (m)</td>
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CORRELATION MATRIX

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To evaluate the vertical accuracy of the up-sampled ASTER DEM, the ASTER DEM data is up-sampled to 0.3m by nearest neighbor interpolation. The pixel-wise errors are derived by subtracting the UAV DEM from the ASTER DEM. Figure 13 shows the spatial distribution of the errors.

Table 2. Statistics of the slopes and the errors

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Figure 10. The histogram of the errors (30m resolution)

Figure 11 shows the slope map of the research area derived from the UAV DEM (30m resolution). It can be seen that the slopes of most areas are greater than 30 degrees.

Table 3. Statistics of the slopes and the errors

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Figure 11. The slope map of the research area (Unit: degrees)

Table 2 shows the statistics of the errors of ASTER GDEM and the slopes. The maximum and the minimum slope are 0 and 89.3837 degrees respectively. The mean and the standard deviation of the slopes are 35.7365 and 16.6769 degrees. The correlation coefficient between the errors of ASTER DEM and the slopes is -0.034, which indicates that the accuracy of ASTER GDEM2 data in the research area is negatively correlated to the slope of the terrain. The errors tend to be large in the flat areas and small in the rough areas. It can be seen from Figure 3 and Figure 11 that most of the flat areas are in the valleys. The large errors at these areas may be caused by worse observation condition.

Figure 12. The stacking map of the ASTER GDEM2 data

Table 3 shows the statistics of the errors of ASTER GDEM and the stacking numbers. The maximum and the minimum stacking number are 7 and 1 respectively. The mean and the standard deviation of the stacking numbers are 5.8178 and 0.8690. The correlation coefficient between the errors of ASTER DEM and the stacking numbers is -0.009, which indicates that the accuracy of ASTER GDEM2 is negatively correlated to the number of stereo observations. The errors tend to be large in the areas that less observed and small otherwise.

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Figure 13. The spatial distribution of the errors derived from the ASTER DEM and the UAV DEM (0.3m resolution)
The maximum and the minimum error are 110.95m and -94.18m respectively. The histogram of the errors is shown in Figure 14. Statistical test shows that the errors come from a normal distribution with a mean of 0.68m and a standard deviation of 15.22m. Although the RMSE (15.22m) is larger than that of the down-sampled result (15.04m), it seems that the accuracy of the up-sampled ASTER GDEM2 data in the research area is not significantly reduced by the complexity of the terrain.

Figure 14. The histogram of the errors (0.3m resolution)

4. CONCLUSIONS AND FUTURE WORK

Global digital elevation models have become an important data source for a range of applications. This study investigates the vertical accuracy of the ASTER GDEM2 over a mountainous area based on UAV photogrammetry. Experimental results show that the elevation errors of ASTER GDEM2 are in normal distribution and the systematic error is quite small. The accuracy of the ASTER GDEM2 coincides well with that reported by the ASTER validation team. It is worth mentioning that the accuracy in the research area is negatively correlated to both the slope of the terrain and the number of stereo observations. This study also evaluates the vertical accuracy of the up-sampled ASTER DEM. Although the accuracy is a little lower than that of the down-sampled result, it seems that the accuracy of the up-sampled ASTER GDEM2 data in the research area is not significantly reduced by the complexity of the terrain.

This study provides fine-grained accuracy evaluation of the ASTER GDEM2 and is informative for the GDEM-supported UAV photogrammetric applications. The future work will focus on accuracy analysis of POS-supported bundle block adjustment, direct georeferencing and ortho-mosaicing.

ACKNOWLEDGMENT

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