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Instruments and Methods

Measuring glacier surface roughness using plot-scale, close-range digital photogrammetry

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ABSTRACT. Glacier roughness at sub-metre scales is an important control on the ice surface energy balance and has implications for scattering energy measured by remote-sensing instruments. Ice surface roughness is dynamic as a consequence of spatial and temporal variation in ablation. To date, studies relying on singular and/or spatially discrete two-dimensional profiles to describe ice surface roughness have failed to resolve common patterns or causes of variation in glacier surface morphology. Here we demonstrate the potential of close-range digital photogrammetry as a rapid and cost-effective method to retrieve three-dimensional data detailing plot-scale supraglacial topography. The photogrammetric approach here employed a calibrated, consumer-grade 5 Mpix digital camera repeatedly imaging a plot-scale (∼25 m²) ice surface area on Midtre Lovénbreen, Svalbard. From stereo-pair images, digital surface models (DSMs) with sub-centimetre horizontal resolution and 3 mm vertical precision were achieved at plot scales ∼4 m². Extraction of roughness metrics including estimates of aerodynamic roughness length (z₀) was readily achievable, and temporal variations in the glacier surface topography were captured. Close-range photogrammetry, with appropriate camera calibration and image acquisition geometry, is shown to be a robust method to record sub-centimetre variations in ablating ice topography. While the DSM plot area may be limited through use of stereo-pair images and issues of obliquity, emerging photogrammetric packages are likely to overcome such limitations.

KEYWORDS: applied glaciology, energy balance, glaciological instruments and methods, snow/ice surface processes, surface melt

INTRODUCTION

Understanding the energy balance is crucial to quantifying the melt rate of snow and ice surfaces. Broadly, the energy balance can be divided into two components: radiative and turbulent energy fluxes. Predominantly, melt occurs in response to radiative energy, which may account for >70% of the net energy (Hock, 2005); however, in maritime settings (Moore and Owens, 1984; Ishikawa and others, 1992) turbulent fluxes may be more significant, contributing up to ∼80% of the melt energy (Willis and others, 2002). Critically, melt energy available from the two energy components can be strongly influenced by the micro-topography or roughness of the melting surface itself.

Impurities at the ice/atmosphere interface contribute to the development of surface roughness at metre to sub-metre scales. Mineral dust content and associated biological consortia can significantly increase the shortwave incident radiation contributing to melt on snow (Warren, 1984; Kohshima and others, 1994; Thomas and Duval, 1995; Conway and others, 1996; Painter and others, 2007) and glacier ice (Kohshima and others, 1993; Adhikary and others, 2000; Takeuchi, 2002) where water content is similarly influential (Cutler and Munro, 1996). Consequently, topography can become exaggerated or inverted

as the relative proportions of radiative and turbulent energies vary, particularly at synoptic timescales (Müller and Keeler, 1969; McIntyre, 1984; Rhodes and others, 1987; Fassnacht and others, 2010). Such dynamics of snow/ice surface roughness modulates the response of remote satellite-mounted sensors, particularly those utilizing microwave wavelengths whose data products relate to signal backscatter and surface dielectric properties (Jin and Simpson, 1999; König and others, 2001; Nolin and others, 2002). Moreover, varying roughness influences radiative incidence angles and surface albedo (Cutler and Munro, 1996; Warren and others, 1998), and defines turbulent energy fluxes at the ice/atmosphere boundary layer (Munro and Davies, 1977).

At Haut Glacier d’Arolla, Switzerland, Brock and others (2006) suggested that variation to within one standard deviation from the mean surface aerodynamic roughness length resulted in turbulent energy flux changes of up to 20%. The temporal evolution of snow surface roughness appears progressive, as described by independent meteorological variables (e.g. Brock and others, 2006; Fassnacht, 2010); however, results for small-scale glacier ice surface roughness are contradictory, with evidence both for and against similar systematic change in space and/or time.
(cf. Dunykerke and Van den Broeke, 1994; Smeets and others, 1999; Brock and others, 2006; Smeets and Van den Broeke, 2008). Consequently, because of the logistical challenges in monitoring melting ice surfaces, the dynamical values of glacier surface roughness remain poorly constrained and surprisingly sparse (Brock and others, 2006).

The absence of well-described surface roughness parameters, especially for changing ice surfaces over time and space, is critical to predictive, numerical runoff models (Hock, 2005). Many spatially distributed ice melt models have ignored variations in roughness, opting for a constant value (Arnold and others, 2006) or empirical derivations of turbulent fluxes (Klok and Oerlemans, 2002), while others have employed statistical roughness distributions (Brock and others, 2000). Yet, to study and forecast ice melt more accurately, it is necessary to know the detailed site-specific variation of the surface roughness over time, highlighting the research necessity to develop a reproducible method to derive such empirical relationships (Andreas, 2011). With increasing research into the use of low-cost photogrammetry for environmental applications (e.g. Chandler, 1999; Lascelles and others, 2002; Smith and others, 2009; James and Robson, 2012; Westoby and others, 2012; Whitehead and others, 2013), the aim of this paper is to evaluate the use of close-range digital photogrammetry as a rapid, cost-effective method to quantify and characterize the dynamics of glacier ice surface roughness. In particular, we build upon our earlier work (Sanz-Ablanedo and others, 2012a) and look to describe spatial and temporal variability of the ice surface, assessing the validity and reproducibility of the method at the plot scale.

**PHOTOGRAMMETRY AND GLACIER SURFACE CHARACTERIZATION**

Over the past 15 years, the slow and manual data acquisition phase of photogrammetry has become fully and digitally automated (Chandler, 1999; Baltsavias and others, 2001; Fox and Gooch, 2001; Westoby and others, 2012; Whitehead and others, 2013; Javenick and others, 2014), allowing high-resolution datasets or digital surface models (DSMs) to be generated. This has resulted in the use of cheaper, consumer-grade digital cameras, of ever-increasing resolution (Chandler and others, 2005; Rieke-Zapp and Nearing, 2005; Taconet and Ciarletti, 2007; Heng and others, 2010). The digital revolution has also provided opportunities to relax strict geometric constraints, allowing greater freedom during initial data acquisition (Heng and others, 2010). Automated camera calibration software has simplified the extraction of relevant parameters describing internal camera geometries, as necessary for accurate data acquisition (Brown, 1971; Fryer and others, 2007; Sanz-Ablanedo and others, 2012b), and continues to advance rapidly. Prior to these developments, close-range plot-scale photogrammetry had played a leading role in deriving roughness of soil surfaces (Welch and others, 1984) and water-worked gravel beds (Butler and others, 1998). However, more recently, scientists have made increasing use of high-resolution DSMs derived from digital images and pixel-matching algorithms to represent soil surfaces and derive measures of roughness without necessitating the direct measurement of specific transects (e.g. Rieke-Zapp and Nearing, 2005; Taconet and Ciarletti, 2007). A review of parameterizations of soil surface roughness from high-resolution DSMs at plot scales reveals a wide variety of roughness metrics including: elevation standard deviation (e.g. Kuipers, 1957), slope angle and tortuosity index (e.g. Boiffin, 1984) and elevation autocovariance (e.g. Taconet and Ciarletti, 2007).

In glaciological contexts, there is a long history of measuring ice surfaces using photogrammetry, particularly over large areas where aircraft mounted cameras offer many logistical advantages (Kääb, 2010). Baltsavias and others (2001) provide a useful historical review and used automated DSM extraction tools based upon photogrammetric area correlation to generate DSMs of Unteraragletscher, Switzerland, which achieved good results compared to airborne laser scanning data, except in areas where image texture (the spatial contrasts and variations in colour and/or light intensity) was low. At closer range, Kaufmann and Ladstädter (2004) used terrestrial photogrammetry to monitor glacier ablation at Goessnitzkees, Austria, with decimetre uncertainties using imagery sources extending over 15 years. For close-range applications, DSMs of ice surface areas of ~100 m² using similar methods have been produced but exhibit similar vertical uncertainties (e.g. Pitkänen and Kajutti, 2004). Difficulties in DSM creation generally arise from areas exhibiting strong shadows or, conversely, large expanses of white ice/snow (see Fox and Gooch, 2001; Hopkinson and others, 2009). Consequently, airborne laser scanning (ALS) of glacier surfaces has been used more frequently to retrieve glacier surface DSMs (e.g. Kennett and Eiken, 1997; Bamber and others, 2005; Hopkinson and others, 2009) and large-scale ice surface roughness metrics (e.g. Van der Veen and others, 1998, 2009; Rees and Arnold, 2006). However, the sampling interval and vertical accuracy for ALS are typically of the order of ~1 m and 0.05 m, respectively, and accordingly inappropriate for energy balance and some remote-sensing applications.

Mirroring successful use of portable laser scanners (Huang and Bradford, 1992; Flanagan and others, 1995) and terrestrial laser scanning (TLS) (Schmid and others, 2004; Perez-Gutierrez and others, 2007) for soil roughness applications, TLS has been trialled for snow and glacier surfaces (e.g. Hopkinson, 2004; Avian and Bauer, 2006; Kerr and others, 2009; Kassalainen and others, 2011; Nield and others, 2013). From such data, DSMs and surface profiles can be retrieved at sub-centimetre horizontal and vertical resolution over intermediate (<200 m) scan distances. Nonetheless, plot-scale TLS point clouds exhibit a number of errors related to scanner hardware, data acquisition, scan and surface geometries, and point-cloud processing (Hodge and others, 2009). The success of TLS on icy surfaces can be influenced by factors including the specific laser wavelength, scan repeatability and the associated signal loss on the varied surface materials and textures in supraglacial environments (Hopkinson, 2004; Kerr and others, 2009). Moreover, for a melting ice surface, it may also be necessary to consider through-water correction where the laser pulse penetrates a film of sediment-free surface water (e.g. Smith and others, 2012).

To date, plot-scale (<10 m) ice surface roughness has principally been quantified using manual surveying methods (Munro, 1989; Brock and others, 2006) or using close-range two-dimensional (2-D) digital photogrammetry. Arnold and Rees (2003) derived surface roughness metrics for both snow and glacier ice at ≤1 m horizontal length scales, and sub-centimetre horizontal and vertical resolution, using predefined planes marked by a black board inserted directly into
the snow or ice surface. Similarly, high-resolution work on
snow surface roughness has employed such a technique (e.g.
Fassnacht and others, 2009a; Manninen and others, 2012).
However, these examples focus on singular, site-specific,
2-D digital profiles, and fail to retrieve three-dimensional
(3-D) data at the plot scale. Therefore, in field settings, appli-
cation of close-range photogrammetry with modern con-
sumer-grade digital cameras offers a potentially rapid, useful
and complementary research tool for deriving surface rough-
ness metrics over space and time at scales critical to both
energy-balance modelling and remote-sensing research.

METHODOLOGY

Field site and image acquisition
This study was conducted during 2010 on Midtre Lovén-
breen, Svalbard (78°52' N, 12°05' E; Fig. 1a). The glacier’s
surface topography is critical to its energy balance and
seasonal ablation (Arnold and others, 2006), with previous
estimates of the supraglacial aerodynamic roughness length
\( z_0 \) between 0.29 and 2.68 mm (Rees and Arnold, 2006). At
site RC3 (Fig. 1a), a location ~200 m a.s.l. on Midtre
Lovénbreen’s centre line, a 2.5 m polyethylene pole was
drilled and secured on 16 July such that the base of the pole
froze into the ice for the duration of observations reported
here (Fig. 1b). The pole provided a fixed reference point
relative to the glacier surface for all surveys at the site.

Digital images of the glacier surface were acquired at the
RC3 plot on 21 and 28 July and 4 and 18 August. The ice
surface for both the first and last surveys was partially snow-
covered: initially as the transient snowline retreated, and
latterly due to a summer snowfall event in mid-August.
Surface roughness elements were preferentially oriented
down-glacier, aligned with the dominant katabatic wind
direction. For each imaging survey, during periods of low
wind speed, two taut 5 m nylon survey strings were tied from
the reference pole to temporary vertical poles secured with
ice screws to provide a horizontal reference plane (Fig. 1b);
to maximize the string tautness, a taut-line hitch was used.
The strings, marked at 0.5 m intervals, were oriented at 90°
defined by a set-square, with one perpendicular to the ice
surface slope and prevailing wind direction, and the second
aligned down-glacier. Imagery was acquired obliquely from
eye level (~1.6 m; Fig. 1b). One or two pairs of convergent
photographs were taken using a 5 Mpix Nikon 5400
consumer-grade digital camera, which had been calibrated
at infinity focus (see below). Photo pairs were captured at
three distances, allowing coverage of the RC3 area of
interest at differing scales: plots of 25, 4 and 1 m
were imaged, using an approximate separation-to-distance ratio
of 1 : 5 to create a suitable stereo-pair with the same image
centre point. In addition to image capture, a coincident
manual roughness survey was conducted, following Brock
and others (2006): measurements of distance between the
ice surface and horizontal marker string to an accuracy of
\( \pm 2.5 \) mm were taken at 100 mm intervals along the single
5 m profile across-glacier.

Photogrammetric processing
Contemporary photogrammetry is a technique that allows
the extraction of the 3-D coordinates of any identifiable point
appearing in two (or more) photos taken from different view-
points (Mikhail and others, 2001; Fryer and others, 2007).
Here we provide a brief synopsis for this particular study.

When an object is photographed, each visible 3-D point
of the object corresponds to a point on the 2-D image plane,
according to the principles of central projection. The ideal
straight path of the light connecting the object point, the
centre of projection and the point on the image is affected
by real elements in camera. These modified paths need to be
calculated and mathematically modelled; this is achieved by
a procedure called ‘self-calibration’. For this study, self-
calibration was realized with 15 specific and convergent
photographs taken at distances 3–6 m from a calibration
surface, which included 132 distinct target points, all visible
on all frames. This target design allowed all targets to be
measured fully (Sanz-Ablanedo and others, 2012b).

Fig. 1. (a) Oblique aerial view of Midtre Lovénbreen indicating site
RC3 (arrow). (b) Illustration of image acquisition at RC3; note the
glacier orientation, and the horizontal reference strings and vertical
reference pole (photograph courtesy of Jon Bridge). (c) Example
visualization of 3-D ice surface reconstruction using PhotoModeler
Scanner™ for an area approximately 3 m x 3 m, with stereo-pair
images acquired on 28 July. The polyethylene marker pole for RC3
is evident.
A second stage of photogrammetric processing involved calculating the position and orientation of the camera at the time of image acquisition. This is achieved by marking homologous points in the pictures. Although strictly just four points are required, a far higher number of such ‘homologous’ points are normally used to help isolate inaccurate or erroneous points. Automated detection algorithms typically facilitate the measurement of several hundred points, thereby allowing reliable determination of the relative position and orientation of the camera at the time of image acquisition, within an initial arbitrary coordinate reference system which can then be redefined. In this project, we manually defined the coordinates of the marks on the taut survey strings and the polyethylene anchor pole. This then allowed translation, rotation and scaling such that each point was related to the reference datum defined by the pole and strings. Once the absolute position and orientation of the two cameras is known in the desired coordinate system, it is possible to calculate the 3-D coordinates of any point visible on both photographs, using an intersection process.

The final part of photogrammetric processing therefore involves generating a high-resolution point cloud to capture and represent all features. This process is generally known as ‘dense surfacing’ and is a well-established and automated procedure (Chandler, 1999), which extracts 3-D information from homologous points appearing on a pair of images with a regular grid pattern. Although 5 Mpix images were acquired, only a portion of these data relate to the specific plot area of interest. Moreover, given the different viewpoint of both images, there are areas that appear in only one of the photographs. Additionally, the procedure relies upon identifying common texture appearing within small patches of pixels, so not all pixels can be recognized reliably (Chandler, 1999). Such areas will not generate reliable 3-D information and will result in holes in the point cloud, as can be identified by the gaps in Figure 1c. In this study, the procedure was particularly influenced by the viewing angle and texture of the surface of the glacier. For the photographic stereo-pairs of areas of 25 m², only a few tens of points were recovered (see Sanz-Ablanedo and others, 2012a), while in the areas of 4 m² and 1 m² the improved texture due to greater contrast and reduced topographic shadowing yielded point clouds consisting of 2500 points m⁻² and 100 000 points m⁻², respectively. Due to the low point-cloud recovery (typically <10 points m⁻²), data from the 25 m² plot areas were excluded from these analyses. Finally, to facilitate comparisons between the four individual image acquisition dates, or with manual measurements, the point clouds were converted to regular gridded meshes using a kriging interpolation routine (e.g. Stein, 1999). Sanz-Ablanedo and others (2012a) provide further details of the photogrammetric procedures adopted.

To quicken the photogrammetric processing, which involves many measurement and iterative computational operations, the use of software tools is preferable. Currently, there is a rapidly increasing and varied range of software packages for photogrammetric applications, although many have been developed for conventional mapping using vertical aerial images. Photomodeler Scanner is a commercial package marketed by Eos Systems Inc., which implements all stages of photogrammetric processing. This package was chosen for several reasons: it is specifically designed for convergent imagery and close-range ground-based imaging; includes tools for automatic measurement and recognition of points required for all stages of data processing; facilitates and implements camera calibration; and finally, provides the user with a high degree of control in processing steps and parameters when automatically generating high-resolution DSMs.

**Roughness metric extraction**

As in soil science, there is a range of definitions of surface roughness used for glaciological applications (Van der Veen and others, 2009). Owing to the large number of potential roughness metrics, the choice of roughness parameter should be informed by the specific process or feature under investigation (Smith and others, 2011). Of specific interest for energy-balance considerations is the aerodynamic roughness length parameter $(z_0)$, which can be estimated by the eddy correlation method, using detailed observations of near-surface wind speed profiles and temperature gradients (Munro and Davies, 1978; Sneets and Van den Broeke, 2008). However, such detailed and sensitive boundary layer measurements are logistically challenging in the context of the hydrometeorological conditions in glacial settings, and Munro (1989) suggested that using micro-topographic data (i.e. high-resolution surface profiles) was the most robust approximation to the surface roughness problem, especially where installation of the necessarily responsive instrumentation may be unfeasible.

Conventional, spatially discrete, micro-topographical techniques to estimate $z_0$ are based on Lettau (1969): along a profile of length $L$, perpendicular to the prevailing wind direction, distances to the ice surface $(h)$ are taken from a local horizontal reference datum, and the roughness parameter $z_0$ is given as

$$z_0 = \frac{\sigma_h s}{S},$$

where $\sigma_h$ is half the effective roughness element height, $s$ is the typical roughness element silhouette area and $S$ is the density of roughness elements (Munro, 1989; Brock and others, 2006). To estimate the effective roughness height, the data series for $h$ is rescaled with a mean equal to zero $(h_0)$, where $\sigma_h$ is the standard deviation of $h_0$ and is equivalent to the ‘random roughness’ detailed by Kuipers (1957). Both approximations for $s$ and $S$ are dependent on the typical roughness element width, defined as $L$ divided by the number of groups of positive deviations above the zero mean across the profile $(f)$. Consequently, to approximate the aerodynamic roughness length, Eqn (1) can be rewritten

$$z_0 = \frac{\sigma_h^2 f}{L}$$

The sampling interval of $h$ needed for adequate representation of $z_0$, varies, from ≤10 cm for aerodynamic roughness estimates over metre scales, to ≤1 cm over sub-metre scales for remote-sensing applications (Rees and Arnold, 2006).

From the photogrammetric DSMs, surface elevation data were extracted along profiles from the DSMs at 10 and 20 mm intervals, respectively, for the 1 and 4 m² plot areas. A total of 95 profiles were employed in both across- and down-glacier directions for each DSM (after Fassnacht and others, 2009b). This sampling approach, although reducing the source data available, significantly improves on any previously published work assessing plot-scale ice roughness, as a much greater density of topographic data is available. The surface elevation $(h)$ data were then detrended. Linear detrending over the profile length was necessary to eliminate the bias introduced by the influence
of larger-scale surface slope within the area of interest and, in light of the non-stationarity of the surface, to allow direct comparison of calculated bidirectional roughness metrics. In addition to estimating \( z_0 \) from the detrended profile data using Eqn (2), derived roughness metrics included (i) effective roughness height \( (\sigma_h) \) described above, (ii) the sum of absolute slopes \( (\Sigma S) \) between each observation point over a profile window of length \( L \) (Currence and Lovely, 1970), and (iii) a dimensionless micro-topography index \( (MI; \text{Romkens and Wang, 1987}) \):

\[
MI = \left( \frac{\sum_i |h_i|}{n} \right) f \frac{1}{L}
\]

for which \( n \) is the number of observations within the profile analysed. Comparable roughness values were also derived from the four discrete ice-surface profiles manually recorded coincident with the dates of image acquisition.

### ANALYTICAL RESULTS

#### Uncertainty and validation

Based upon the scale and geometry defined by the imagery acquired, it is possible to quantify the precision of data generated using photogrammetry. Here this was achieved internally by the PhotoModeler software using variance propagation methods and is conveyed by the ‘point quality’ output table. By using the average of achieved precisions for all points used for geo-referencing the surveys (18 points in 1 m² surveys and 17 points in 4 m² surveys), global precision can be estimated in three dimensions. In this configuration, the horizontal \( x \)-axis is approximately parallel to the camera lens whilst both the \( y \)- and vertical axes are at an oblique angle relative to the plane of the camera lens. For the 1 m² surveys, average precisions for these axes were approximately 2.5, 5.0 and 3.4 mm. For the 4 m² surveys, imagery was acquired less steeply and the associated precisions were 3.8, 6.4 and 2.6 mm respectively. As expected, the axis approximately orthogonal to the plane of the camera lens exhibited the lowest precision, while the axis most closely aligned with the camera lens plane yields the highest precision. Consequently, the best precision is obtained along the vertical axis in the 4 m² test area, where the degree of convergence of imagery is lower. Note that point precision varies spatially: optimum precision is achieved for those points located closest to the camera, but precision gradually degrades with increasing distance.

The scale of the DSMs created is defined by the control markers located upon the two string lines established at the RC3 supraglacial plot. Attempts were made to make these as taut as possible, but inevitably these string lines did exhibit a small degree of sag. It is possible to quantify the degree of sag using a catenary equation, generally used for correcting survey measurements acquired using a steel band (Uren and Price, 2005):

\[
C_{\text{catenary}} = \frac{w^2D^3(\cos 2\theta)}{24T^2}
\]

where \( D \) is the string line length, \( w \) is weight per unit length of tape, \( \theta \) is the vertical angle between end-points and \( T \) is the applied tension. Assuming a modest tension of 10 N, for the 5 m string line length with a weight per unit length of 0.15 N m⁻¹, the appropriate correction would be just 1.1 mm for the horizontal string lines used in this study. This correction is comparatively minor and it must be remembered that such a small-scale discrepancy would have negligible effect on the vertical precision of the derived DSMs or measures of roughness generated here.

To validate the reconstructed DSMs, data from the manually measured, concurrent surface profiles at 100 mm horizontal intervals were used for comparison with the surface profiles derived from the DSMs (Fig. 2). Because the DSM and manually collected datasets are independent, and both contain uncertainty, total least-squares regression was used to appropriately quantify the ‘degree of fit’ between the datasets. Where edge effects were noted, data were removed from the regression. Given the known importance of image texture and shadowing in stereo-pair photogrammetry (Chandler, 1999; Fox and Gooch, 2001; Hopkinson and others, 2009), the regressions were re-run following removal of manually measured data points noted as being snow or small-scale (< 3 cm diameter) debris holes. Table 1 presents the measures of error retrieved from the residuals. Both the root-mean-square error (RMSE) and mean absolute error (MAE) were typically \( \leq 16 \text{ mm} \), and this reduced to sub-centimetre uncertainties when accounting for surface characteristics likely to result in greater uncertainty in the final DSM (e.g. textureless snow, or isolated debris holes). Such uncertainties improve on the 5 mm reported for singular 2-D surveys of ice surface profiles by Arnold and Rees (2003). Critically, our photogrammetric uncertainties compare particularly well to real-time-kinematic differential GPS systems, whose vertical accuracy may only be of the order of >17 mm (Wheaton and others, 2010). Appreciably, this validation is limited to the cross-glacier direction, but provides reasonable confidence in the data retrieved by the close-range photogrammetry.

### Plot-scale datasets

Figure 3 shows an example of the DSMs created for the 1 and 4 m² plot areas, and the corresponding roughness metrics for the sampled profiles. Plots of the four roughness
metrics showed a clear contrast between the down- and cross-glacier directions. Table 2 and Figure 4 summarize data for the four roughness metrics retrieved from the DSMs. Despite approximately linear standard-deviation–interquartile-range (IQR) relationships for the data in the down-glacier direction (Fig. 5), the profiles’ surface elevation data deviated from the pattern expected for normally distributed data. This observation was more marked for the across-glacier dataset. With consistently low or negative kurtosis (less than 3), the surface elevation in all four surveys was non-normally distributed in both directions. The roughness metrics similarly exhibited non-normal distributions, thus requiring non-parametric methods of comparison.

Figure 4 presents box-and-whisker plots of the distribution of the cross- and down-glacier roughness metrics for the four discrete DSMs. To explore ice surface anisotropy, Wilcoxon rank sum tests were used to examine whether the roughness metric distributions in cross- and down-glacier directions exhibited statistically similar median values. For the 1 m² area, a significant difference was evident between all cross- and down-glacier roughness measures (p < 0.00021); for the 4 m² plot, all metrics showed a significant directional contrast (p < 0.00016) with the exception of S on 18 August (p = 0.42). Similarly, comparison between 1 m² and 4 m² plot area scales through application of rank tests demonstrated that with the exception of down-glacier σh and S on 18 August (p > 0.46), no similarity between the roughness measures at the two contrasting length scales was observed for each of the four surveys.

The plots of surface elevation descriptors for the four surveys (Figs 4 and 5) suggested that, particularly for the cross-glacier direction, there were differences over time. Difference models of the DSMs for 28 July and 4 August (Fig. 6a) emphasized spatial variation in surface change (equivalent to melt rates), with a maximum of ~0.19 m ice ablation or 0.028 m d⁻¹. Over this period, mean ice surface lowering over the plot areas was 0.044 ± 0.035 m and 0.044 ± 0.033 m for 4 and 1 m², respectively. Critically, there was spatial variability in ice ablation: heightened melt was observed in elongated areas oriented parallel with ice flow direction (down-glacier) at ~0.8 and 1.9 m, and in areas that were initially at lower elevations. Kruskal–Wallis

<table>
<thead>
<tr>
<th>Dataset (date, area)</th>
<th>Number of points, n</th>
<th>Sum of absolute errors</th>
<th>RMSE</th>
<th>Standard error of the mean</th>
<th>MAE</th>
<th>Mean point elevation difference for profile</th>
</tr>
</thead>
<tbody>
<tr>
<td>28 July, 1 m²</td>
<td>10 (8)</td>
<td>0.119 (0.040)</td>
<td>0.016 (0.008)</td>
<td>0.005 (0.003)</td>
<td>0.012 (0.005)</td>
<td>0.004 (~0.002)</td>
</tr>
<tr>
<td>28 July, 4 m²</td>
<td>18 (17)</td>
<td>0.209 (0.129)</td>
<td>0.016 (0.011)</td>
<td>0.004 (0.003)</td>
<td>0.012 (0.008)</td>
<td>0.001 (~0.005)</td>
</tr>
<tr>
<td>4 Aug, 1 m²</td>
<td>10 (7)</td>
<td>0.073 (0.043)</td>
<td>0.009 (0.007)</td>
<td>0.003 (0.003)</td>
<td>0.007 (0.006)</td>
<td>0.028 (0.016)</td>
</tr>
<tr>
<td>4 Aug, 4 m²</td>
<td>19 (19)</td>
<td>0.308 (0.308)</td>
<td>0.024 (0.024)</td>
<td>0.006 (0.006)</td>
<td>0.016 (0.016)</td>
<td>0.009 (0.009)</td>
</tr>
<tr>
<td>18 Aug, 1 m²</td>
<td>10 (7)</td>
<td>0.054 (0.031)</td>
<td>0.006 (0.006)</td>
<td>0.002 (0.002)</td>
<td>0.005 (0.004)</td>
<td>0.003 (~0.001)</td>
</tr>
<tr>
<td>18 Aug, 4 m²</td>
<td>19 (14)</td>
<td>0.176 (0.041)</td>
<td>0.013 (0.004)</td>
<td>0.003 (0.001)</td>
<td>0.009 (0.003)</td>
<td>0.011 (0.002)</td>
</tr>
</tbody>
</table>

Table 2. Summary statistics for roughness metrics z₀, σh, S, and MI for the 1 and 4 m² areas over the four observation dates: data are reported as mean, median and standard deviation (σ) for the n = 95 observation sets in cross- and down-glacier directions. Results from the manual survey are shown for comparison

<table>
<thead>
<tr>
<th>Day</th>
<th>Value</th>
<th>1 m²</th>
<th>Down-glacier</th>
<th>4 m²</th>
<th>Cross-glacier</th>
<th>1 m²</th>
<th>4 m²</th>
<th>1 m²</th>
<th>Cross-glacier</th>
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<td>21 July*</td>
<td>Mean</td>
<td>0.0003</td>
<td>0.200</td>
<td>0.008</td>
<td>0.093</td>
<td>0.0004</td>
<td>0.012</td>
<td>0.456</td>
<td>0.189</td>
</tr>
<tr>
<td></td>
<td>Median</td>
<td>0.0001</td>
<td>0.194</td>
<td>0.008</td>
<td>0.087</td>
<td>0.0004</td>
<td>0.011</td>
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<td></td>
<td>σ</td>
<td>0.0001</td>
<td>0.032</td>
<td>0.002</td>
<td>0.030</td>
<td>0.0002</td>
<td>0.005</td>
<td>0.051</td>
<td>0.076</td>
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<td>28 July</td>
<td>Mean</td>
<td>0.0001</td>
<td>0.494</td>
<td>0.013</td>
<td>0.242</td>
<td>0.0005</td>
<td>0.013</td>
<td>0.463</td>
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<td>Median</td>
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<td>0.459</td>
<td>0.011</td>
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<td>0.0005</td>
<td>0.012</td>
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<tr>
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<td>σ</td>
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<td>0.172</td>
<td>0.006</td>
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<td>0.0003</td>
<td>0.005</td>
<td>0.093</td>
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<tr>
<td>4 Aug</td>
<td>Mean</td>
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<td>0.286</td>
<td>0.013</td>
<td>0.151</td>
<td>0.0008</td>
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<td>0.010</td>
<td>0.119</td>
<td>0.0006</td>
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<tr>
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<td>Mean</td>
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<td>0.007</td>
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<tr>
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<td>–</td>
<td>–</td>
<td>–</td>
<td>–</td>
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</tr>
</tbody>
</table>

*Due to an edge artefact in the DSM generation, ten profiles in the cross- and down-glacier directions were removed from analyses for this survey.
tests were used to examine differences between the four discrete surveys, revealing statistically significant ($p < 0.05$) contrast for all roughness metrics at both length scales and both orientations (Fig. 4). Of interest in Figure 4 is the asynchronous variation in temporal trends of roughness metrics; for example, cross-glacier $\sigma_h$ appears to increase between 21 July and 4 August, while cross-glacier MI peaks on 28 July and decreases thereafter. The down-glacier metrics exhibited the least variation over time, as reflected in the ice ablation patterns on the surface.

Simple comparison of the spatially discrete manual roughness survey against the data derived from the 95 cross-glacier profiles for the 4 m$^2$ plot (Table 2) highlighted that, with the exception of 4 August, the estimates for $z_0$ lie within one standard deviation of the mean value. However, the 0.003 mm difference between the two values retrieved for 4 August suggested that while discrete measurements represent an invaluable estimate of roughness descriptors, at the length scales of 1–5 m these may not necessarily reveal appropriate or representative values.

**Repeatability**

To examine the repeatability of the digital surface reconstruction process, DSMs derived for both the 1 and 4 m$^2$ plots on 28 July were compared to data derived from a second image stereo-pair acquired at the same time by examining the surface difference (Fig. 6b). The mean elevation difference across the plot area was relatively small in both instances: for 1 m$^2$, $-0.005 \pm 0.013$ m and for 4 m$^2$, $-0.003 \pm 0.015$ m. Such differences lie remarkably close to the sub-centimetre uncertainties estimated for each of the DSMs, and provide confidence in the method, particularly for the 4 m$^2$ plot. Areas indicating greatest uncertainty in ice surface topography were observed at the domain edge, in discrete topographic lows where topographic shadowing occurred (e.g. debris or cryoconite holes) and where point

![Shaded contour maps for (a) 1 m$^2$ and (b) 4 m$^2$ plot areas for 28 July plotting distance from the reference datum; contours at 0.0015 m intervals. Corresponding graphs below each DSM illustrate values for roughness metrics ($z_0$, $\sigma_h$, $\delta S$ and MI) for the respective plot areas for each of the 95 cross-glacier (dashed) and down-glacier (solid) profiles. The co-location of the two plot areas is indicated by the dashed outline on the 4 m$^2$ plot (b).](image-url)
Fig. 4. Box-and-whisker plots describing the distribution of four roughness metrics, (a) $z_0$, (b) $\sigma_h$, (c) $\Sigma S$ and (d) MI, for the four image acquisition dates for both 1 and 4 m$^2$ plots. Clear plots indicate the across-glacier direction, and greyed plots indicate down-glacier. For each survey date, the first box-plot pair represent the 1 m$^2$ area.

Fig. 5. Plots of standard deviation vs IQR for the raw surface elevation profile data for (a) the 4 m$^2$ plot areas and (b) the 1 m$^2$ plot areas. The dashed lines represent the distribution expected for a normal (Gaussian) dataset.
cloud returns were absent (cf. Figs 1c, 3a and 6a). Comparison of all the roughness metrics with a Wilcoxon rank sum test revealed that at the 4 m² scale, with the exception of cross-glacier $z_0$, all metrics exhibited statistically similar medians at the 99% confidence level for the two paired surveys. The results for the paired 1 m² plots were different: all metrics exhibited significantly different medians ($p \leq 0.01$), with the exception of down-glacier $\sigma_h$ ($p = 0.034$).

**DISCUSSION AND CONCLUSION**

This paper provides the first assessment of close-range digital photogrammetry as a tool to retrieve data relating to glacier surface topography at the plot scale. Here, using a commercially available software package and a consumer-grade 5 Mpix camera, we demonstrate successful DSM generation to describe a glacier surface from stereo-pair imagery. Horizontal resolution of the 1 m² ice surface DSM was sub-centimetre, with 3.4 mm vertical precision, while for the 4 m² plot area, comparable horizontal resolution and vertical precision of 10 and 2.6 mm, respectively, was achieved. Validation of the DSMs against manual, across-glacier surveys of the ice surface suggested mean vertical uncertainties were <16 mm, being the same order or less than those associated with traditional 2-D profiling methods (e.g. Arnold and Rees, 2003) and real-time-kinematic differential GPS (Wheaton and others, 2010). Comparison of DSMs created from two concurrent but independent stereo-pairs indicated surface elevation differences were broadly comparable to the uncertainties associated with the DSMs themselves. The effectiveness of this photogrammetric method is dependent on the ice surface texture; improved results are achieved where there is increased rugosity and distributed, fine debris or cryoconite deposits, two inextricably linked properties of an ablating ice surface (McIntyre, 1984; Irvine-Fynn and others, 2011). Critically, there has been increasing recognition of dust and cryoconite as a characteristic feature on glacier ablation zones worldwide (Hodson and others, 2008), suggesting the photogrammetric method to characterize glacier surfaces could be widely adopted.

The maximum and mean plot-scale ice surface lowering, 28 and 6.3 mm d⁻¹, respectively, exceeds the sub-centimetre uncertainties in our DSM data, and compares well to the

![Fig. 6. Spatial difference plots showing surface elevation change between 28 July and 4 August for (a) the 1 m² and (b) 4 m² plot areas, and between DSMs generated using two independent stereo-pairs acquired on 28 July for (c) the 1 m² and (d) 4 m² plot areas. Contours in (a–d) are at 0.025 m intervals. Note that in (c) and (d) sites of significant difference are associated with topographic lows and as discrete point locations are suggestive of the impact of isolated debris holes which experience topographic or self-shadowing, thereby influencing the vertical precision of the photogrammetrically derived DSM data in these locations.](image)
seasonally averaged ablation rate of $15.5 \pm 2.8 \text{mm d}^{-1}$ observed at Midtre Lovénbreen during 2010. The greatest variability in surface lowering was seen in the cross-slope direction, corresponding to the existing topographic lows, and indicating the role that meltwater drainage at the rill (centimetre to decimetre) scale has in changing ice topography. Variability in the glacier surface, and its ablation rates over the four measurement dates, influenced by the partial cover from summer snow, suggested temporal change in roughness metrics could be expected. The estimates of mean supraglacial $\tau_y$ between 0.007 and 0.043 m correspond well with equivalent metrics for rough snow of $\sim 0.01$ m (Fassnacht and others, 2009a,b), while the order-of-magnitude range of mean $z_\tau$ values 0.0002–0.002 m agrees with existing estimates for supraglacial aerodynamic roughness values (Brock and others, 2006), and even matches those for the same glacier (Rees and Arnold, 2006). The contrast between the metrics and their median values for the two plot scales, using the photogrammetry-derived data and manual measurements, may be related to the fact this study bridged the scale-independent ($<1$ m) and scale-dependent ($>1$ m) length scales (Rees and Arnold, 2006).

One key pattern found with all four roughness metrics reported here was the contrast between down- and cross-glacier directions. In all cases, the down-glacier metrics were consistently lower than across-glacier. The use of detrending prior to calculating roughness metrics lends weight to this result. Interestingly, this bidirectional roughness variation was found at both the scale-dependent and scale-independent length scales. Such surface anisotropy has been recorded for glaciers, both at the large scale (Smith and others, 2006) and the small to micro-scale (Rees and Arnold, 2006), and is well known for sea ice as an important variation to account for in remote-sensing applications (Jin and Simpson, 1999).

There was variability in the surface roughness measures; however, the apparent temporal trends over the four survey dates between 21 July and 18 August were neither progressive nor synchronous across the four metrics considered. This subtle difference highlights a question that is currently an active area of research (see Smith, 2014), particularly in consideration of remotely sensed data: which roughness metric is the most appropriate to use? Nonetheless, as could be expected, the roughness for times of partial snow cover (21 July and 18 August) was significantly lower than for the glacier ice. This was demonstrated by the lowered standard deviation and IQR (Fig. 5), implying that the presence of residual snow smoothed the surface and removed extremes in the micro-topography at the 1 and 4 m$^2$ plot scales. While there was no evidence of long-term systematic change in roughness (e.g. Smeets and others, 1999; Smeets and Van den Broeke, 2008), with only four snapshots considered here, we suggest the method presented, and the ability to test similarity of roughness metrics, provides a route by which temporal (and spatial) variation in roughness parameters and aerodynamic roughness approximations can be explored quantitatively. More detailed surface analyses could include utilization of semivariance (Arnold and Rees, 2003) or spectral (Hubbard and others, 2000; Rees and Arnold, 2006) analyses of plot areas. However, the implication that, for the ice surface, significant difference in roughness can be observed at a single plot suggests that Brock and others' (2006) assertion that a single, distributed micro-topographic survey of the glacier surface would be sufficient to describe spatial variation in roughness ($z_\tau$) may demand further assessment.

A limitation of the photogrammetric method described here was associated with the varying obliquity associated with capturing plot areas of $>4$ m$^2$ with an image acquisition height of 1.6 m. The increasingly shallow viewing angle for larger plot-scales increases areas of shaded, or self-shadowed, ice surface behind micro- and meso-topographic ridges or features. For the specific camera and stereo-pair input used here, more detailed analyses demonstrated that for an improved representation of the 4 m$^2$ area, the image capture height should be increased by 28 cm, to $\sim 1.9$ m, while at a height of $\sim 3$ m photogrammetric solutions of 9–25 m$^2$ ice surface plots would be possible (Sanz-Ablanedo and others, 2012a). This could be readily achieved using inspection poles designed to elevate cameras up to 10 m above the surface and allow the method presented here to be scaled up to larger plot areas. However, since the initiation of this work, photogrammetric software packages (e.g. Eos Systems’ PhotoModeler Scanner, Agisoft’s PhotoScan) and their algorithms have advanced markedly, and it is now possible to use multiple images over larger areas and with varied viewpoints to capture and reconstruct topography (e.g. Javernick and others, 2014). While incorporating additional uncertainty, this technical advance reduces the limitations afforded by stereo-pair imagery; additional images can capture otherwise self-shaded locations. Similarly, incorporation of additional images increases the size of the plot area that can be reconstructed using a photogrammetric approach. Nonetheless, the resolution of DSMs produced using this method is dependent on the camera resolution and camera-to-surface distance. For example, imagery from 300 m above a glacier surface with a 10 Mpix camera yields a horizontal surface resolution of only 0.12 m per pixel (Whitehead and others, 2013). However, sub-centimetre resolutions across the range of length scales reported here may be readily achieved using ground-based image capture with contemporary consumer-grade cameras which now typically capture $>12$ Mpix images. However, the successful DSM creation for the 1 and 4 m$^2$ plots demonstrates that a photogrammetric approach, with appropriate image settings and acquisition geometry, can provide a powerful technique to examine glacier surfaces, and variability thereof, at the plot scale in high spatial resolution.

The roughness of ice surfaces demands continued investigation, addressing problems with backscatter to satellite-mounted sensors as well as resolving appropriate aerodynamic roughness representations. Specifically, on glaciers, researchers have yet to resolve questions over assumed spatial and scale similarity (Rees and Arnold, 2006), persistence of spatial patterns (Brock and others, 2006) and progressive temporal variations (Smeets and Van den Broeke, 2008) in roughness. Recent research has even suggested that ice surface microbiology can be strongly influenced by local hydrology which is explicitly related to plot-scale topography (e.g. Edwards and others, 2011). Consequently, the ability to retrieve data to construct detailed DSMs of a glacier surface and derive roughness metrics at the resolutions and length scales reported here is a step-change from previously published 2-D datasets, and is an important technique to develop further. Close-range digital photogrammetry has advantages in the form of low-cost and rapid data acquisition, and is well suited to surveys across a wide range of spatial scales or utilizing autonomous
time-lapse image acquisition modes, with retrospective data processing to retrieve the desired variable(s). The method opens new possibilities for interrogating and constraining supraglacial processes, in particular offering the potential to feed high-density topographic datasets into studies of ice surface characteristics at the glacier scale, which are important for both remote-sensing applications and glacier ablation. Given the rapidly advancing area of less constrained photogrammetry methods for environmental monitoring (e.g. James and Robson, 2012; Westoby and others, 2012; Jaivinck and others, 2014), this assessment of a more conventional use of the technique provides a benchmark for the anticipated increase in these types of studies in glacial environments.

ACKNOWLEDGEMENTS

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Adhikary S, Nakawo M, Seko K and Shakya B (2000) Dust influence to PhotoModeler software at reduced academic rates. Help- ful comments from Tim James, Arwyn Edwards and Tom Holt are appreciated. Informative input from Duncan Quincey, and constructive and detailed insights from two anonymous reviewers significantly improved this paper.


extraction and digital close-range photogrammetry. *ISPRS Arch.*, **39**(B5), 435–440


