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Heterogeneous water storage and thermal regime of supraglacial ponds on debris-covered glaciers

C. Scott Watson¹, Duncan J. Quincey¹, Jonathan L. Carrivick¹, Mark W. Smith¹, Ann V. Rowan², Robert Richardson³

1. School of Geography and water@leeds, University of Leeds, Leeds, UK
2. Department of Geography, University of Sheffield, Sheffield, UK
3. School of Mechanical Engineering, University of Leeds, UK

Correspondence to: C. Scott Watson (scott@rockyglaciers.co.uk)

Abstract

The water storage and energy transfer roles of supraglacial ponds are poorly constrained, yet they are thought to be important components of debris-covered glacier ablation budgets. We used an unmanned surface vessel (USV) to collect sonar depth measurements for 24 ponds to derive the first empirical relationship between their area and volume applicable to the size distribution of ponds commonly encountered on debris-covered glaciers. Additionally, we instrumented nine ponds with thermistors and three with pressure transducers, characterising their thermal regime and capturing three pond drainage events. The deepest and most irregularly-shaped ponds were those associated with ice cliffs, which were connected to the surface or englacial hydrology network (maximum depth = 45.6 m), whereas hydrologically-isolated ponds without ice cliffs were both more circular and shallower (maximum depth = 9.9 m). The englacial drainage of three ponds had the potential to melt ~100 ± 20 × 10³ kg to ~470 ± 90 × 10³ kg of glacier ice owing to the large volumes of stored water. Our observations of seasonal pond growth and drainage with their associated calculations of stored thermal energy have implications for glacier ice flow, the progressive enlargement and sudden collapse of englacial conduits, and the location of glacier ablation hot-spots where ponds and ice cliffs interact. Additionally,
the evolutionary trajectory of these ponds controls large proglacial lake formation in deglaciating environments.

1. Introduction

Debris-covered glaciers are an increasingly common part of the mountain cryosphere, as glacier mass loss promotes the exhumation of englacial rock debris and the development of supraglacial debris layers [Benn et al., 2012; Thakuri et al., 2014]. A combination of low glacier surface gradients, stagnating glacier termini and negative mass balance regimes is initiating increased supraglacial water storage on Himalayan debris-covered glaciers through the positive feedback of solar radiation absorption and transmission to glacier ice [Reynolds, 2000; Sakai et al., 2000; Quincey et al., 2007; Benn et al., 2012; Salerno et al., 2012]. Coalescing ponds approaching the hydrological base level can ultimately form large proglacial lakes, which are impounded at the edge of a glacier and can expand rapidly through calving [Kirkbride, 1993; Watanabe et al., 1994; Sakai et al., 2009; Rohl, 2008; Carrivick and Tweed, 2013]. In this paper we define supraglacial ponds as water bodies ≤20,000 m² [e.g. Biggs et al., 2005]. The expansion of supraglacial ponds and proglacial lakes is ongoing across the central and eastern Himalaya [Komori, 2008; Gardelle et al., 2011; Nie et al., 2013; Wang et al., 2015; Zhang et al., 2015; Watson et al., 2016; Nie et al., 2017], and is of great concern not least due to the potential glacial lake outburst flood (GLOF) hazards for downstream communities and infrastructure [Kattelmann, 2003; Benn et al., 2012; Carrivick and Tweed, 2016; Rounce et al., 2016; Rounce et al., 2017a], but also for potential effects on glacier flow dynamics and on glacier mass balance [Carrivick and Tweed, 2013].

Glacial lake bathymetry data are predominantly collected for large proglacial lakes in order to parameterise GLOF hazards or to investigate the ice-marginal lake
interactions at glacier calving fronts (Fujita et al., 2009; Shrestha and Nakagawa, 2014; Somos-Valenzuela et al., 2014a; Lamsal et al., 2016; Purdie et al., 2016; Sugiyama et al., 2016). Cook and Quincey (2015) compared three commonly-used empirical relationships between glacial lake area and volume, the latter as derived from some knowledge of lake bathymetry, and replotted a compiled dataset of 75 lake measurements, which included several lakes surveyed multiple times. Their analysis revealed that predicted lake volume can vary by an order of magnitude for a given area due to variable lake bathymetry morphologies. A wide spread of predicted volumes is problematic when attempting to characterise an increasing number of lakes developing in deglaciating basins (Carrivick and Tweed, 2013). Additionally, the bathymetry of proglacial lakes or ponds associated with debris-covered glaciers cannot be derived using remotely sensed imagery due to high turbidity, in contrast to examples of supraglacial lakes from Greenland (e.g. Box and Ski, 2007; Moussavi et al., 2016; Pope et al., 2016). Therefore, there is a clear requirement to better constrain empirical relationships between lake area and volume and to include data from the size distribution of supraglacial ponds typically found on debris-covered glaciers (Cook and Quincey, 2015).

Supraglacial ponds and adjacent ice cliffs appear as hot-spots of melt in multi-temporal digital elevation model (DEM) differencing, in contrast to subdued melt under a thick insulating debris-layer (e.g. Nicholson and Benn, 2013; Immerzeel et al., 2014; Pellicciotti et al., 2015; Ragettli et al., 2016; Thompson et al., 2016). Therefore, quantifying the spatio-temporal dynamics of supraglacial ponds and their interaction with ice cliffs is essential to assess their contribution towards glacier-wide ablation and seasonal water storage (Miles et al., 2016b; Watson et al., 2016); however, little is known about their seasonal expansion and contraction, thermal regime, or bathymetry.
Optical satellite images are often obscured by clouds during the Indian Summer Monsoon, which restricts assessments of seasonal pond dynamics. Cloud-penetrating synthetic aperture radar (SAR) data has been used to map and monitor glacial lake dynamics on the Greenland Ice Sheet (e.g. Miles et al., 2017); however, Himalayan applications are limited to large proglacial lakes thus far (Strozzi et al., 2012). Additionally, studies instrumenting ponds with temperature or water level sensors are rare (e.g. Xin et al., 2012; Horodyskyj, 2015; Miles et al., 2016a; Narama et al., 2017). Notably, the modelling of supraglacial pond energy balance by Sakai et al. (2000) and Miles et al. (2016a) revealed that ponds effectively absorb and transmit thermal energy englacially, and therefore are likely to be key contributors to glacier downwasting.

In this study, we aimed to assess the annual variation in supraglacial pond dynamics and characteristics in order to assess their role for water storage, englacial ablation, and interaction with ice cliffs on debris-covered glaciers. Specifically we: (1) derive the bathymetry, thermal regime, and seasonal water level change, of supraglacial ponds on Khumbu and Lobuche Glaciers in the Everest region of Nepal; (2) use the bathymetry of 24 supraglacial ponds to derive an empirical area-volume relationship; (3) quantify the potential englacial ablation of draining ponds; (4) quantify the variation in pond morphology in relation to the presence of ice cliffs.

2. Data collection and methods

2.1 Study sites

Field data were collected on the debris-covered zones of Khumbu and Lobuche Glaciers in the Everest region of Nepal during three field campaigns (October/ November 2015, May 2016, and October 2016) (Figure 1). Khumbu Glacier is ~17 km long with a ~10 km debris-covered ablation area, of which the lower ~4 km is essentially stagnant (Quincey et al., 2009). Within this stagnating part of the glacier,
supraglacial ponds are increasingly coalescing to form a connected chain of surface water that is approaching the glacier’s hydrological base level [Watson et al., 2016], which is expected to transition into a proglacial lake [Naito et al., 2000; Bolch et al., 2011]. By comparison, Lobuche Glacier is smaller, with a relic debris-covered ablation zone ~1 km in length that is now disconnected from clean ice at higher elevations in the accumulation area.

Figure 4.1. (a) The location of the study ponds on Khumbu and Lobuche Glaciers, Nepal with a hillshaded Pleiades DEM background. (b) The lower terminus of Khumbu Glacier (c) The disconnected terminus of Lobuche Glacier. (d) The upper ablation zone of Khumbu Glacier at the transition from debris-covered to clean ice. Pond catchments are shown as light blue polygons (b and c). Inset backdrops are a panchromatic Pleiades satellite image (16th May 2016).
2.2 Pond depth surveys

An unmanned surface vessel (USV) was custom-built for the acquisition of global positioning system (GPS)-referenced depth measurements from unfrozen supraglacial ponds (Table 1, Figure 2). The USV was deployed on 24 supraglacial ponds during a field campaign in May 2016 and completed a total survey distance of 18 km. Ponds were surveyed along transects (Figure 2c), although the track spacing was variable between ponds and was greatest for larger ponds.

Table 4.1. USV specifications

<table>
<thead>
<tr>
<th>Feature</th>
<th>Specification</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dimensions L × W × H</td>
<td>56 × 45 × 16 cm</td>
</tr>
<tr>
<td>Weight</td>
<td>5.5 kg</td>
</tr>
<tr>
<td>Power</td>
<td>2 × 5800 mAh 11.1V LiPo batteries</td>
</tr>
<tr>
<td>Speed</td>
<td>~3 km/h</td>
</tr>
<tr>
<td>Sonar</td>
<td>Furuno 235dt-pse 235 kHz</td>
</tr>
<tr>
<td></td>
<td>Maximum depth reading: 100 m</td>
</tr>
<tr>
<td></td>
<td>1 measurement per second</td>
</tr>
<tr>
<td></td>
<td>Accuracy: ±3%</td>
</tr>
<tr>
<td>Assembly time</td>
<td>15 minutes</td>
</tr>
<tr>
<td>Remote control range</td>
<td>150 metres</td>
</tr>
</tbody>
</table>
Figure 4.2. (a) The unmanned surface vessel deployed on a small isolated ‘green’ pond with no ice cliffs present, and (b) a large connected highly-turbid ‘grey’ pond with ice cliffs. (c) An example of the point depth measurements collected during a survey, and (d) bathymetry derived by interpolation of the data in (c). (e) The unmanned surface vehicle and modular components.

2.3 Supraglacial pond characteristics

Ponds were classified into three categories to characterise their hydrological connectivity and turbidity (e.g. Wessels et al., 2002; Takeuchi et al., 2012): ‘isolated ponds’ (n = 9) did not have inflows/outflows of water or ice cliffs; ‘connected ponds
without ice cliffs’ \((n = 7)\) were of higher turbidity and had inflows or outflows of water; and ‘connected ponds with ice cliffs’ \((n = 11)\) were of highest turbidity, had inflows or outflows of water, and exhibited regular debris input due to ice cliff retreat. In this study, all ponds with ice cliffs had hydrological connectivity beyond their individual basins, although we note that this may not always be the case. Similarly, ponds may transition from one classification to another. The pond classification can generally be recognised by characteristic colours: isolated ponds were either green or clear, ponds connected without ice cliffs were turquoise, and ponds connected with ice cliffs were grey.

Hereon, ponds are referred to by their ID number (Figure 1), prefixed by ‘K’ for ponds on Khumbu Glacier or ‘L’ for ponds on Lobuche Glacier. The area and boundary of ponds was derived following Watson et al. \(2016\), by applying object-based image analysis (OBIA) classification applied to a panchromatic Pleiades satellite image (16\(^{th}\) May 2016) followed by manual inspection and editing using corresponding multi-spectral imagery. All bathymetric surveys were conducted within 14 days of this image, hence we assumed that pond area did not change within this interval. However, pond boundaries buffered by one pixel were used in an uncertainty assessment described in the next section.

2.4 Supraglacial pond bathymetry

Sonar depth measurements collected by the USV were interpolated using the Natural Neighbour algorithm in ArcGIS, which preserves the values of measurement points, to derive the bathymetry for each supraglacial pond. Interpolated depths were gridded at 0.5 m resolution and forced to zero at the pond boundary derived from the Pleiades imagery, other than where an ice cliff was present, since cliffs are associated with pond deepening (Miles et al. 2016a; Watson et al. in press). The surface area and volume of each pond were used to derive a power-law area-volume relationship. We
performed a leave-one-out analysis to assess the uncertainty when deriving pond volumes using this area-volume relationship. Here, the area and volume of individual ponds were omitted from the dataset and an updated area-volume relationship was derived. We then used this updated relationship to predict the omitted pond’s volume and assess the difference compared to the volume predicted using the original relationship (i.e. with the complete pond dataset). Uncertainties in pond area and volume were derived from 1000 Monte Carlo simulations of pond area and volume for each pond. Here, a Gaussian error model with a standard deviation equal to 3 % of the maximum pond depth (based on the sonar sensor uncertainty) was added to interpolated pond bathymetries using pond extents derived from the Pleaides satellite image (500 simulations), and these same pond extents with a one pixel buffer (500 simulations). These buffered pond extents represented potential pond expansion in the 14 day interval between the acquisition of the satellite image, and the final bathymetric surveys. Each simulation produced an estimate of pond area and volume, and the uncertainties reported in Table 2 represent one standard deviation of the 1000 simulations.

To investigate the spatial relationship between water depth and ice cliff presence, we generated sub samples of pond depth for all ponds with and without ice cliffs present. For ponds without an ice cliff present, polygonal buffers were generated from the pond shorelines at 0–5 m and 5–10 m. For ponds with an ice cliff present, polygonal buffers were also generated from pond shorelines at 0–5 m and 5–10 m, and additional buffers of the same distance were generated from the shoreline bounded by ice cliffs. These buffers were used to extracted depth measurements using both the raw data points and interpolated pond bathymetry. Pair-wise Mann-Whitney U tests were then used to test for differences between each subsample of depth measurements.
2.5 Supraglacial pond instrumentation and monitoring

Ibutton thermistors (DS1922L-F5, number \(n = 18\), with a manufacturer’s stated accuracy ±0.5°) were deployed in eight supraglacial ponds in October 2015 to monitor the thermal regime for one year, with a recording frequency of 60 minutes. Thermistors were shielded in a metal casing and deployed on a string, with one sensor floating at the pond surface attached below a floating buoy and one anchored at the bed by a weight. The strings were reset after downloading data in May 2016 where despite a clear water level rise for several ponds, the buoys remained at the surface due to slack left in the system. Therefore, we expect the top sensors were within 20 cm of the surface at all times. Pond K12 had two thermistor strings, one near an inlet to the pond and one near an ice cliff. Water level change was recorded at three ponds which captured different pond characteristics and drainage regimes (K9, K19, K18/21) using Solinst Levelogger Junior Edge pressure transducers.

Thermistors and pressure transducers were calibrated in a controlled temperature environment, revealing a maximum deviation of ±0.22°C, which was within the manufacturer’s stated accuracy of ±0.5°C. Individual temperature calibrations were applied to the field data based on the mean deviation from one thermistor used as a reference logger located in pond K9. Pressure transducers were barometrically compensated using data from the Barologger to report the water level change in ponds K9, K19 and K18/21. Here, water level refers to depth of water above the pressure transducer, which was not the point of maximum pond depth for K9 and K18/21, but was for K19.

Ponds K18/21 and K9 were turbid, with ice cliffs present, and had englacial (K18/21) and supraglacial (K9) drainage outlets. Pond K19 appeared less turbid and did not have an ice cliff present. The pressure transducer in the channel connecting K18/21...
was withdrawn in May 2016 following entrapment on the bed. A Barologger Edge was deployed on Khumbu Glacier, which was shielded from solar radiation and recorded 1 m above-ground temperature and atmospheric pressure, for correction of the pond data.

We used corresponding pond bathymetry and water level data to derive a volumetric time-series for ponds K9 and K19. This was derived by simulating pond drainage at 20 cm increments and calculating the remaining pond area and volume for each iteration. Polynomial trend lines were fitted to area-volume scatter plots and the relationship was used to estimate the pond volume for each water level measurement. This was not possible for K18/21, since the pressure transducer became exposed and was withdrawn before bathymetry was collected in May 2016. Supraglacial pond drainage was simulated for all ponds with bathymetry data to derive individual area-volume relationships. Finally, study pond catchments were derived using ArcHydro Tools in ArcGIS and the circularity of pond boundaries was calculated using:

$$\text{Circularity} = \frac{P^2}{4\pi A}$$

where P and A are pond perimeter (m) and area (m$^2$) respectively. A circle would have a score of one.

3. Results

3.1 Supraglacial pond bathymetry and volume

The area of supraglacial ponds ranged from 39 m$^2$ (K14, Maximum depth 1.3 m) to 18,744 m$^2$ (K21, Maximum depth 45.6 m) with interpolated volumes of 24 m$^3$ and 290,928 m$^3$ respectively (Table 2). K21 was also the deepest pond at 45.6 m. Surveyed ponds had a total volume of $407,214 \pm 2908$ m$^3$ and $92,020 \pm 680$ m$^3$ on Khumbu and Lobuche Glaciers respectively. The water volume on Lobuche Glacier...
represents the total visible surface water storage since all supraglacial ponds were surveyed, whereas an estimated 26% of total pond area was surveyed on Khumbu Glacier.

Table 4.2. Supraglacial pond characteristics

<table>
<thead>
<tr>
<th>Pond ID</th>
<th>Area (m²)</th>
<th>Volume (m³)</th>
<th>Maximum depth (m)</th>
<th>Connectivity</th>
<th>Circularity</th>
<th>Notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>K1</td>
<td>943 ± 36</td>
<td>1443 ± 41</td>
<td>5.5</td>
<td>I</td>
<td>1.90</td>
<td></td>
</tr>
<tr>
<td>K2</td>
<td>2729 ± 63</td>
<td>9049 ± 111</td>
<td>9.9</td>
<td>I</td>
<td>2.37</td>
<td></td>
</tr>
<tr>
<td>K3</td>
<td></td>
<td></td>
<td></td>
<td>C</td>
<td></td>
<td></td>
</tr>
<tr>
<td>K4</td>
<td>958 ± 34</td>
<td>1460 ± 29</td>
<td>3.5</td>
<td>Cl</td>
<td>1.83</td>
<td>Data gaps*</td>
</tr>
<tr>
<td>K5</td>
<td></td>
<td></td>
<td></td>
<td>CI</td>
<td></td>
<td>Data gaps*</td>
</tr>
<tr>
<td>K6</td>
<td>4010 ± 56</td>
<td>19042 ± 119</td>
<td>9.1</td>
<td>I</td>
<td>1.28</td>
<td></td>
</tr>
<tr>
<td>K7</td>
<td>255 ± 10</td>
<td>379 ± 18</td>
<td>4.3</td>
<td>I</td>
<td>1.38</td>
<td></td>
</tr>
<tr>
<td>K8</td>
<td>250 ± 13</td>
<td>445 ± 18</td>
<td>4.4</td>
<td>C</td>
<td>1.14</td>
<td></td>
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<tr>
<td>K9</td>
<td>5422 ± 54</td>
<td>26959 ± 213</td>
<td>14.6</td>
<td>CI</td>
<td>3.47</td>
<td></td>
</tr>
<tr>
<td>K10</td>
<td>392 ± 11</td>
<td>752 ± 24</td>
<td>4.5</td>
<td>I</td>
<td>1.27</td>
<td></td>
</tr>
<tr>
<td>K11</td>
<td>600 ± 28</td>
<td>882 ± 22</td>
<td>3.4</td>
<td>CI</td>
<td>1.98</td>
<td></td>
</tr>
<tr>
<td>K12</td>
<td>3654 ± 107</td>
<td>5746 ± 158</td>
<td>6.0</td>
<td>CI</td>
<td>6.63</td>
<td></td>
</tr>
<tr>
<td>K13</td>
<td>231 ± 10</td>
<td>342 ± 13</td>
<td>3.3</td>
<td>I</td>
<td>1.14</td>
<td></td>
</tr>
<tr>
<td>K14</td>
<td>39 ± 6</td>
<td>24 ± 4</td>
<td>1.3</td>
<td>I</td>
<td>1.36</td>
<td></td>
</tr>
<tr>
<td>K15</td>
<td>596 ± 18</td>
<td>1087 ± 28</td>
<td>4.0</td>
<td>CI</td>
<td>1.73</td>
<td></td>
</tr>
<tr>
<td>K16</td>
<td>625 ± 16</td>
<td>1497 ± 32</td>
<td>5.5</td>
<td>I</td>
<td>1.55</td>
<td></td>
</tr>
<tr>
<td>K18</td>
<td>6482 ± 119</td>
<td>30058 ± 293</td>
<td>13.5</td>
<td>C</td>
<td>3.40</td>
<td></td>
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<tr>
<td>K19</td>
<td>1193 ± 59</td>
<td>2116 ± 66</td>
<td>6.1</td>
<td>C</td>
<td>4.00</td>
<td></td>
</tr>
<tr>
<td>K20</td>
<td>3502 ± 56</td>
<td>13369 ± 163</td>
<td>8.8</td>
<td>CI</td>
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</tr>
<tr>
<td>K21</td>
<td>18744 ± 97</td>
<td>290928 ± 1528</td>
<td>45.6</td>
<td>CI</td>
<td>1.56</td>
<td>Main drainage event 19-25/07/16</td>
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<tr>
<td>K22</td>
<td>754 ± 24</td>
<td>1634 ± 28</td>
<td>4.5</td>
<td>I</td>
<td>1.27</td>
<td>Drained 27-29/08/16</td>
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<tr>
<td>K23</td>
<td></td>
<td></td>
<td></td>
<td>C</td>
<td></td>
<td>Drainage initiation on 02/07/16</td>
</tr>
<tr>
<td>L1</td>
<td>3789 ± 38</td>
<td>16986 ± 101</td>
<td>9.8</td>
<td>CI</td>
<td>1.63</td>
<td></td>
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<tr>
<td>L2</td>
<td>207 ± 10</td>
<td>235 ± 16</td>
<td>3.1</td>
<td>C</td>
<td>2.05</td>
<td></td>
</tr>
<tr>
<td>L3</td>
<td>318 ± 11</td>
<td>496 ± 19</td>
<td>4.3</td>
<td>C</td>
<td>1.53</td>
<td></td>
</tr>
<tr>
<td>L4</td>
<td>13101 ± 192</td>
<td>71514 ± 502</td>
<td>17.1</td>
<td>CI</td>
<td>2.71</td>
<td></td>
</tr>
</tbody>
</table>
The bathymetric data revealed that ponds rapidly deepen from the shoreline into one or more distinct basins (Figure 3). The maximum pond depths for connected ponds with an ice cliff, connected ponds without an ice cliff, and isolated ponds, were 45.6 m (K21), 13.5 m (K18), and 9.9 m (K2) respectively. Hydrologically-isolated ponds tended to be more circular in shape (mean circularity 1.50 ± 0.40) when compared to connected ponds with or without ice cliffs (mean circularity 2.54 ± 0.50) (Table 2), which had elongated or multiple basins (e.g. Figure 3f, g). Despite being initially classified as an isolated pond, an ice cliff appeared at the southern margin of K6 over the summer monsoon and the water level had dropped ~1 m between May–October 2016. Therefore, the pond classifications used in this study are not exclusive and ponds may transition between classes. Connected ponds with ice cliffs were generally deeper in areas adjacent to the cliff faces and there was no evidence of an ice ramp protruding from any of the faces (e.g. Figure 3b, c, d). However, this was less apparent for several smaller ice cliffs (e.g. Figure 3a) or those bounded by areas of slumping debris, where pond depth increased more gradually from the shoreline.
Figure 4.3. Examples of pond bathymetry for connected ponds with ice cliffs (a–d), connected ponds without ice cliffs (e–h), and isolated ponds (i–l). The top edge of adjacent ice cliffs are shown as black lines. Note different scales between panels. A smoothed (low-pass filter) hillshade overlay is shown with transparency.

We analysed the spatial variation in pond bathymetry using two shoreline buffers (0–5 m and 5–10 m) for groups of ponds with and without ice cliffs (Figure 4). The buffers for ponds with ice cliffs were split into cliff and non-cliff shoreline zones. Considering pond depth characteristics using the raw data points (e.g. Figure 2c), for ponds with ice cliffs, the area of pond 5–10 m from the ice cliff face had the highest median and mean depth (4.70 and 7.62 m respectively) (Figure 4). The median depth 0–5 m from ice cliffs (4.00 m) was significantly deeper than the median depth 0–5 m from non-cliff shorelines (1.60 m) (pair-wise Mann-Whitney U test, p < 0.05). Considering the non-cliff shoreline of ponds with and without ice cliffs, the 5–10 m zones were statistically different although this difference was small (median depths of 3.40 m and 3.70 m respectively). Pond depth comparisons using the interpolated pond bathymetry were
similar to those using the raw data points, but generally of shallower depth (Supplementary Figure 1).

Figure 4. Depth characteristics of all ponds with and without cliffs using the raw data points sampled with buffers (illustrated in the inset). All buffers were tested against each other using pair-wise Mann-Whitney U tests. Medians (bold) were statistically significant at $p < 0.05$ unless indicated with a red circle. Mean values are shown above the median.

3.2 Area-volume relationships

The area and volume of surveyed supraglacial ponds in this study followed a power-law trend with a strong positive correlation ($R^2 = 0.98$) (Figure 5a); however, connected ponds with ice cliffs displayed greatest variability. The observed power-law relationship was comparable to that for the compiled dataset of Cook and Quincey (2015) and effectively extended the coverage for supraglacial ponds by three orders of magnitude (Figure 5b). When comparing the area-volume data points during simulated pond drainage ($n = 761$), the gradient of the power-law increased and had a slightly lower $R^2$ of 0.97 (Figure 5c). The calculated area-volume relationship for K6
and K20 highlighted the effects of variable pond geometry (Figure 5d). K6, which had a regular shoreline and simple geometry draining into a central basin (Figure 3i), had a good fit to the power-law (Figure 5d). In contrast, K20 (Figure 3c) had an irregular shoreline and featured high-depth areas adjacent to an ice cliff, which had greater deviation from the power-law (Figure 5d). The ability to predict an individual pond volume $V$ (m$^3$) using area $A$ (m$^2$) using the power law (Equation 2) was not related to pond size (Supplementary Figure 2), and instead likely reflects variable pond morphologies. The mean difference between pond volumes calculated using the bathymetry and predicted pond volumes was 22.8%, or 18.5% with an outlier (K21) removed. Based on the addition of the new data from our study, the modified area-volume relationship of Cook and Quincey (2015) becomes:

$$V = 0.1535A^{1.39}$$  \hspace{1cm} (2)

from the original:

$$V = 0.2A^{1.37}$$  \hspace{1cm} (3)

The percentage differences between volumes predicted using the new versus old equation ranged from 7% (K21) to 19% (K14). We also note that the pond with an ice cliff studied by Miles et al. [2016a] on Lirung Glacier, Nepal, had an area of 650 m$^2$ and estimated volume of 1250 m$^3$, which is accurately predicted by Equation (2) as 1245 m$^3$. 
Figure 4.5. (a) Area-volume relationships derived from data in this study. (b) The data of this study combined with the compiled dataset of Cook and Quincey (2015). (c) The data of (b) with additional datapoints from simulating the drainage of surveyed ponds. (d) The area-volume relationship for ponds K6 and K20 during simulated drainage.

3.3 Supraglacial pond thermal regime and drainage events

After deployment of the pressure transducer in October 2015, the water level at K19, which was connected without an ice cliff, gradually increased before stabilising December 2015 to March 2016 (Figure 6a). The water level started to rise in the first two weeks of April 2016 and the diurnal temperature range increased. The water level reached a peak on the 20th June 2016 at 7.1 m with an estimated volume of 3,145 m³. The water level then decreased gradually until the 18th July 2016. Subsequent drainage occurred 19th–25th July 2016 from a water level of ~6.4 m to ~3.4 m, which
corresponded to an estimated 2,106 m$^3$ decrease in volume. Diurnal temperature range increased following this drainage (Figure 6a). The final drainage event began 10$^{th}$–11$^{th}$ September 2016 and the water level was zero by 22:00 on the 12$^{th}$ September 2016. Notably, the drainage event initiated on the 19$^{th}$ July 2016 was coincident with a rapid temperature rise in the neighbouring pond K20 (Supplementary Figure 3), which we interpret as a coincident drainage event exposing the temperature logger to the atmosphere. K20 was not instrumented with a pressure transducer but field observations confirmed complete drainage by the start of October 2016.

Figure 4.6. Water level and temperature for (a) K19, (b) K9, and (c) K18/21. Slumped ice plates are visible along the shoreline of K21 in November 2015 (c). Pond bathymetry was used to derive a volumetric time series for (a) and (b). Pond temperatures represent the pond bed. Yellow rectangles indicate the author for scale. Dates are dd/mm/yy.
The water level of K9, which was connected and with several ice cliffs, decreased during winter until ~20\textsuperscript{th} March 2016 (Figure 6b). The water level then increased during summer reaching peaks on the 12\textsuperscript{th} July 2016 (5.5 m) and 21\textsuperscript{st} August 2016 (5.7 m) before decreasing throughout September 2016. The rising and falling limbs both include a diurnal temperature cycle, but not a diurnal water level fluctuation. The initial and final water level of K9 were comparable (4.6 m on 22\textsuperscript{nd} October 2015 and 4.1 m on 6\textsuperscript{th} October 2016). Notably, the water level of K18/21 also decreased during the ~1 month of observations in winter, and was also apparent from slumped ice plates around the margin of the pond (Figure 6c).

The absolute temperature of K9 was lower than K19; however, the diurnal temperature variation was greatest for K9 with a summer variation of ~0.5°C (Figure 7). In contrast, the summer diurnal temperature variation for K19 was ~0.2°C. There was evidence of a subdued winter diurnal temperature cycle in K9 but not in K19; however, the water temperature remained above 4°C in K19 during winter.

Three complete pond drainage events were observed over the study period at K19, K20, and K22. The potential internal glacier ablation resulting from pond drainage through downstream ice Mi, assuming a temperature drop to 0°C, can be estimated using Equation (4) [e.g. Thompson et al., 2016]:

\[
M_i = M_w \Delta T \left( \frac{C}{L} \right) \tag{4}
\]

where \(M_w\) is the water mass, \(\Delta T\) indicates the difference in temperature between observed and zero, \(C\) the specific heat capacity of water (4.2 kJ kg\(^{-1}\) k\(^{-1}\)) and \(L\) the latent heat of melting (344 kJ kg\(^{-1}\) k\(^{-1}\)). For \(\Delta T\) we used the median of the mean surface and bottom temperatures of the day prior to drainage. Measurement uncertainties in pond temperature (± 0.5°C) and volume (Table 2) were used to derive a confidence
interval when calculating the englacial ablation potential of drained ponds. The
drainage of K22 with a water mass of $1634 \pm 28 \times 10^3$ kg and temperature change
$10.3 \pm 0.5$ °C had an englacial ablation potential of $203 \pm 14 \times 10^3$ kg. The drainage
of K20 with a water mass of $13369 \pm 163 \times 10^3$ kg and a temperature change of $2.9 \pm
0.5$ °C had an englacial ablation potential of $473 \pm 88 \times 10^3$ kg. The drainage of K19
with a water mass of $2116 \pm 66 \times 10^3$ kg and temperature change of $4.0 \pm 0.5$ °C had
an englacial ablation potential of $103 \pm 17 \times 10^3$ kg.

Figure 4.7. Mean and standard deviation of 20 minute daily pond bottom water
temperatures for ponds (a) K9 and (b) K19. Note the broken y-axis (b). Winter
and summer intervals are 28/10/15−19/03/16 and 20/03/16−01/09/16
respectively.

3.4 Pond temperature
The mean water temperature of isolated ponds and connected ponds without ice cliffs,
was greater than connected ponds with ice cliffs (Figure 8). Mean bottom water
temperatures for connected ponds with ice cliffs ranged from 0.1 to 0.5°C for winter
and 0.2 to 2.0°C for summer. In contrast, the mean bottom water temperatures for
isolated and connected ponds without ice cliffs ranged from 4.1 to 5.0°C for winter and
4.2 to 9.8°C for summer. Summer surface water temperatures were comparable to
summer bottom water temperatures, suggesting that ponds were well mixed. The
greatest difference in mean summer surface and bottom water temperature was 1.2°C
for K13. Winter surface temperatures were distinctly different from the winter bottom water temperatures for five ponds (K12, K23, K19, K13, K22), suggesting that temperature loggers became frozen in a layer of ice over winter, which was supported by field observations. K9 had an inflow and outflow of water and was largely ice-free on 22\textsuperscript{nd} October 2015, but had developed a surface layer of ice \~10 cm thick by 11\textsuperscript{th} November 2015. Other ponds without hydrological connectivity were observed to be freezing over upon arrival into the field on 20\textsuperscript{th} October 2015.

Figure 4.8. Seasonal supraglacial pond bottom and surface water mean temperatures. Error bars indicate one standard deviation. Coloured x-axis represents indicative pond colour for connected ponds with an ice cliff (grey), connected ponds without an ice cliff (turquoise), and isolated ponds (green).

4. Discussion

4.1 Supraglacial pond bathymetry
Our data contribute to a sparse bathymetry dataset of supraglacial ponds on Himalayan debris-covered glaciers, which is due to the significant effort that is required to make distributed depth measurements. Point measurements of pond or lake depths
are usually made using a sonar after cutting through the frozen surface (Benn et al., 2001); directly through the frozen surface using sonar (Thompson et al., 2012); from a boat (Rohl, 2008; Somos-Valenzuela et al., 2014b); or more recently using ground-penetrating radar (GPR) (Mertes et al., 2016), which provides additional information on basal sediment. The spatial resolution of surveys can be increased when using an USV (Horodyskyj, 2015), which also allows sampling close to ice cliffs where falling debris would otherwise restrict safe access.

Hydrologically-isolated ponds without ice cliffs were located along the western margin of Khumbu Glacier, which is stagnant and supporting some scrub vegetation in parts (Inoue and Yoshida, 1980); however, the sub-debris ice content is unknown. These isolated ponds often exhibited a green appearance due to algal growth and low turbidity (Takeuchi et al., 2012), and were generally smaller and shallower than the other ponds surveyed. However, isolated ponds K2 and K6 had depths of 9.9 m and 9.1 m respectively (Table 2). Notably, an ice cliff developed along the margin of K6 and the water level dropped by over 1 m during the 2016 summer monsoon. Pond K6 was identified by Iwata et al. (1980) in 1978 and has persisted in a similar shape with little change in areal extent (Watson et al., 2016). It is not clear whether a degrading ice core in this area will lead to further drainage. Nonetheless, it indicates that apparently stable ponds on stagnant parts of the glacier may still be actively deepening towards the hydrological base level.

Connected ponds with ice cliffs had the greatest depths and often elongated or multiple basins (Figure 3). The greatest measured depth was 45.6 m at K21, which is located in an area of maximum ice thickness of ~100 m (Gades et al., 2000). The maximum depth was observed adjacent to a large ice cliff (Figure 3d), similar to observations by Thompson et al. (2016) and Horodyskyj (2015) on Ngozumpa Glacier,
who observed maximum depths of 27 m and 54 m close to ice cliffs. Miles et al. [2016a] observed linearly increasing pond depth approaching an ice cliff on Lirung Glacier, although depth measurements were limited due to rockfall hazard. Whilst pond depth generally increased approaching large ice cliff faces in our observations (Figure 3), this was not the case for many smaller ice cliffs or those in areas of slumping debris. This variability led to highest depth observations in areas 5−10 m from cliff faces (Figure 4). Shallower pond depth around small ice cliffs likely reflects newly exposed cliff faces or areas where thick subaqueous debris cover due to slumping restricts basal melt [Mertes et al., 2016]. In contrast, larger cliff faces develop and persist where subaqueous thermal erosion matches or exceeds subaerial melt or calving [Benn et al., 2001; Sakai et al., 2009]. Here, rapid cliff retreat limits debris accumulation at the cliff base and hence it is likely that subaqueous melt of the cliff and pond bed promotes deepening approaching the cliff face. Therefore, the bathymetry of ponds with ice cliffs indicates whether pond expansion is likely, and hence the evolutionary trajectory of the cliff-pond coupling.

4.2 Area-volume relationships

We have derived a new empirical area-volume relationship by extending the dataset of Cook and Quincey (2015), through measuring the bathymetry of 24 supraglacial ponds on two debris-covered glaciers not included in their original analysis. Watson et al. [2016] found that individual ponds <3,600 m$^2$ made up 48−88% of pond area on nine debris-covered glaciers in the Everest region, which is well represented by the range of ponds in this study (39−18,744 m$^2$), but was not represented in existing datasets [Cook and Quincey, 2015]. All surveyed ponds followed a power-law relationship between area and volume (Figure 5a); however, ponds with ice cliffs displayed greatest deviation from this trend, owing to the prevalence of deep zones
adjacent to ice cliffs (e.g. K21 Figure 3d), or to elongated irregular shaped basins (e.g. K12). This variable morphology was apparent when using a power law to predict individual pond volumes (Supplementary Figure 2), where the mean difference between the pond volume and predicted pond volume was 22.8%, or 18.5% with an outlier (K21) removed. An even larger bathymetric dataset would further refine empirical area-volume relationships and facilitate further analysis of ponds of variable ice cliff presence, different hydrological connectivity, and at different stages of development. However, it is clear that area-volume relationships derived from predominantly proglacial lakes can be extended to smaller ponds.

A mean of ~50% of ice cliffs featured a supraglacial pond in the Everest region [Watson et al., 2017], therefore around half of all pond morphologies are expected to be highly dynamic due to ice cliff retreat. In contrast, subaqueous sub-debris melt rates and therefore geometry changes are expected to be low for those without ice cliffs [Miles et al., 2016a]. Multi-temporal supraglacial pond bathymetry data are therefore required to assess pond evolution and the trajectory of proglacial lake development. Simulating pond drainage increased the number of area-volume data points (Figure 5C) and was used to reconstruct volume change, which is useful to assess the seasonal water storage of ponds. However, these represent a snapshot of pond bed morphology and do not reveal anything about its genesis.

4.3 Supraglacial pond drainage and thermal regime

Supraglacial ponds display considerable inter- and intra-annual variation in their surface area [Miles et al., 2016b; Watson et al., 2016], including evidence of a seasonal peak related to the onset of the ablation season with the Indian Summer Monsoon in June [Liu et al., 2015; Miles et al., 2016b; Narama et al., 2017]. The observed hydrological regime of ponds K19 and K9 displayed a seasonal trend and
water levels peaked in late June and early July 2016 respectively (Figure 6a, b). The
water level of K9 dropped slightly before a secondary peak on the 21st August. K9 had
both an inflow and outflow of water and formed part of a connected series of ponds on
the easterly margin of Khumbu Glacier (Figure 1b). Therefore, the seasonal water level
peak of this pond reflects maximum meltwater generation during the monsoon, before
the water level began to drop in late August 2016.

The sporadic drainage of K19 was likely due to the interception and enlargement of
an englacial conduit, which drained most of the pond and was more rapid than the
gradual seasonal expansion and drainage observed at K9. The main drainage event
at K19 occurred 19th–25th July 2016; however, the water level of the pond began to
drop gradually several days prior to this. Notably, the adjacent pond K20 also
appeared to drain on the 19th July 2016, inferred by the aerial exposure of the
temperature logger (Supplementary Figure 3). The temperature of K19 rose from 4°C
to 5.5°C within 24 hours during this drainage event (Figure 6a), which may be due to
greater local radiation heating as the pond volume decreased. K20 had a lower
temperature (~3°C), although it appears likely that both ponds exploited the same
englacial conduit to drain and hence developed a sub-surface connection. It is clear
that ponds in close proximity are likely to exhibit or develop sub-surface connections
that are not apparent from surface observations. Notably, the timing and connectivity
of these ponds is also consistent with the observations of Narama et al. [2017], who
also observed englacial drainage June-July for supraglacial ponds in the Tien Shan
Mountains.

Water temperatures of ponds with ice cliffs (n = 6) were generally close to zero
degrees Celsius and featured a frozen surface during winter, and summer surface
temperatures did not exceed 4.2°C (Figure 8). The pond monitored by Xin et al. [2012]
also had an ice cliff present but was located at ~1685 m lower elevation and had an average summer surface temperature of 9.0°C. The pond with an ice cliff monitored by Miles et al. (2016a) at an elevation of 4070 m had an average temperature of 1–1.5°C and was more comparable to our study ponds. From surveys of ponds on debris-covered glaciers in Nepal, Sakai et al. (2009) revealed that subaqueous melt exceeds subaerial ice cliff melt at ponds with a water temperature of 2–4°C and a fetch of >20 m, and that a fetch >~80 m is required for calving. During our study, full-slab calving was observed at K15 and block calving was apparent prior to K12 being instrumented in October 2015 (Supplementary Figure 4). These ice blocks at K12 had melted by May 2016. The ponds had fetches of ~35 m and ~140 m respectively, estimated following Sakai et al. (2009). The temperature at K15 was not recorded but the basin was connected to K12, which had a mean temperature of <1°C, and hence did not meet the expected calving criteria. We note that Sakai et al. (2009) also observed a calving cliff with a pond temperature of less than <1°C on Khumbu Glacier, but with a larger fetch of 94 m. It is likely that the higher turnover of water into K15 via an inlet, promoted thermal undercutting and calving despite the small fetch. The role of water inflows should therefore be considered alongside wind-driven currents in the energy-balance modelling of supraglacial ponds (e.g. Miles et al., 2016a).

The temperature of isolated ponds was notably higher than those with ice cliffs, and all had mean temperatures >4°C at the pond bed over winter, suggesting that their thermal energy was stored over winter, insulated by a layer of snow-covered ice (Figure 8). Ponds K19, K20, and K22 drained over the study period with englacial ablation potentials of $103 \pm 17 \times 10^3$ kg, $473 \pm 88 \times 10^3$ kg, and $203 \pm 14 \times 10^3$ kg respectively. Notably K20 had the greatest englacial ablation potential despite having the lowest temperature (1.8°C), due to its volume. These values only represent
englacial ablation due to drainage without considering the turnover of water through each pond during their lifespan. Nonetheless, our observations support the role of seasonal pond dynamics acting as a notable source of glacier ablation [Miles et al., 2016a].

4.4 Implications of seasonal pond dynamics

Through revealing short-term supraglacial pond dynamics on two Himalayan debris-covered glaciers, our study adds further evidence to support observations of cyclical pond growth and drainage due to ablation and precipitation input during the summer monsoon, followed by winter freezing of the pond surface, which subdues or inhibits diurnal temperature cycles. However, we observed ponds continuing to drain throughout winter (e.g. K9 and K18/K21), suggesting that the hydrological system and hence englacial ablation remained active at this time. Sporadic drainage events are imposed on the seasonal cycle of pond expansion as ponds intercept englacial conduits and stored thermal energy is transmitted englacially, contributing notably to glacier-wide ablation [Sakai et al., 2000; Miles et al., 2016a]. Additionally, meltwater was conveyed through ponds by a network of supraglacial and englacial channels (e.g. K15), which leads to a high overturning rate conducive to the undercutting and calving of adjacent ice cliffs [Sakai et al., 2009; Miles et al., 2016a].

Our observations of glacier surface hydrology in the lower ablation zone suggest that seasonal pond expansion and drainage was linked to meltwater generation from glacier ablation; however, this seasonal trend was occasionally interrupted by sporadic short-term drainage events, where ponds intercepted an englacial conduit. These conduits become enlarged when supporting a drainage event and grade towards the hydrological base level, and may support multiple drainage events in between periods of dormancy [Gulley and Benn, 2007; Gulley et al., 2009; Benn et al., 2009]. In this
study we investigated the potential englacial ablation from draining ponds; however, the quantity and ablative role of water stored in blocked englacial conduits remains unknown. Englacial water storage likely also responds to seasonal ablation processes driving meltwater generation, which is stored and released during outburst flood events [e.g. Rounce et al., 2017b] or buffered by the internal drainage system for larger low-gradient glaciers. In the upper ablation zone of debris-covered glaciers, observations of increased summer velocities on Lirung and Ngozumpa Glaciers, suggest that basal-sliding is promoted where surface meltwater is routed the glacier bed (e.g. via crevasses) [Kraaijenbrink et al., 2016, Benn et al., 2017]. Seasonal velocity observations are lacking for Khumbu Glacier; however, it is likely that surface-bed hydrological connections play a similar role.

Knowledge of short-term pond dynamics and evolution as presented in this study allows a greater understanding of their potential role for glacier-wide ablation, in addition to developing future links between observed mass loss and the meltwater budget. Several key lines of enquiry should be pursued to better constrain the role of the pond-cliff interaction for glacier-wide ablation, and to quantify seasonal meltwater fluxes, which will become increasingly important as the reservoirs of freshwater locked in glaciers decline over the coming century:

1. Observations of pond dynamics throughout the summer monsoon with high temporal resolution optical or SAR imagery could be used in association with empirical area-volume relationships to predict water fluxes.

2. Distributed instrumentation of ponds on other debris-covered glaciers combined with pond energy balance modelling [e.g. Miles et al., 2016a] would allow a spatio-temporal assessment of pond thermal regimes and water level
fluctuations, which could be used to model their total contribution to glacier-wide ablation.

(3) Multi-temporal pond bathymetry is required to quantify expansion processes related to ice cliff retreat or debris infilling (e.g. Thompson et al., 2016; Mertes et al., 2016), and to better understand the interface between the thickness of basal sediment and glacier ice, which has a large influence on subaqueous melt rates.

5. Conclusions

A spatially distributed bathymetric survey of supraglacial ponds has extended the compiled dataset of Cook and Quincey (2015), to determine an empirical relationship between pond area and volume for the size-distribution of ponds commonly encountered on Himalayan debris-covered glaciers. We revealed evidence of pond deepening in association with the presence of ice cliffs up to a depth of 45.6 m, which supports observations of thermal undercutting and their role as hot-spots of melt on debris-covered glaciers. Downward grading of the pond bed in association with ice cliff retreat is likely to promote ice cliff persistence (Watson et al., in press), and a contrasting evolutionary trajectory compared to ponds without ice cliffs. The water temperatures of ponds with ice cliffs (n = 5) were generally close to zero degrees Celsius, and summer surface temperatures did not exceed 4.2°C. The temperature of hydrologically isolated ponds was notably higher and all had mean temperatures >4°C at the pond bed over winter, suggesting that their thermal energy is stored over winter, trapped in by an insulating layer of snow-covered ice.

Seasonal expansion and drainage of ponds was observed, which supports satellite remote sensing observations (Liu et al., 2015; Miles et al., 2016b; Watson et al., 2016). Continued pond drainage throughout winter suggests the hydrological system and
therefore englacial ablation remains at least partially active throughout the year.

Sporadic drainage events were imposed on this seasonal cycle as ponds intercepted englacial conduits and transferred their stored thermal energy englacially. Increased meltwater generation during the summer monsoon is also likely to be expressed through englacial water storage and release \cite{Rounce2017}, in addition to conduit enlargement and collapse, leading to the formation of new surface depressions and ice cliffs.

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Supplementary information

Heterogeneous water storage and thermal regime of supraglacial ponds on debris-covered glaciers

C. Scott Watson¹, Duncan J. Quincey¹, Jonathan L. Carrivick¹, Mark W. Smith¹, Ann V. Rowan², Robert Richardson³

1. School of Geography and water@leeds, University of Leeds, Leeds, UK
2. Department of Geography, University of Sheffield, Sheffield, UK
3. School of Mechanical Engineering, University of Leeds, UK
Supplementary Figure 1. Depth characteristics of ponds with and without cliffs using interpolated pond bathymetry. Medians were statistically significant (pair-wise Mann-Whitney U tests) at $p < 0.05$ unless indicated with a red circle.

Supplementary Figure 2. Relationship between pond area and the percentage difference between pond volume calculated using bathymetry, and predicted pond volume using the power-law trend (Fig. 5b) and a leave-one-out analysis.
Supplementary Figure 3. Pond temperature at K20 alongside air temperature measured 1 m above the surface of Khumbu Glacier. It is thought the surface temperature logger became exposed on the 19th July due to pond drainage leading to a rapid temperature rise, which was coincident with the drainage of K19.

Supplementary Figure 4. Block and full-slab calving at K12 and K15 respectively. Red circles indicate corresponding features in each pair of images. A well-established undercut notch was present at K15 in October 2015, which facilitated the calving event.