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Quantifying ice cliff evolution with multi-temporal point clouds on the debris-covered Khumbu Glacier, Nepal

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ABSTRACT

Measurements of glacier ice cliff evolution are sparse, but where they do exist, they indicate that such areas of exposed ice contribute a disproportionate amount of melt to the glacier ablation budget. We used Structure from Motion (SfM) photogrammetry with Multi-View Stereo (MVS) to derive 3D point clouds for nine ice cliffs on Khumbu Glacier, Nepal (in November 2015, May 2016 and October 2016). By differencing these clouds, we could quantify the magnitude, seasonality, and spatial variability of ice cliff retreat. Mean retreat rates of 0.30 to 1.49 cm d⁻¹ were observed during the winter interval (November 2015 to May 2016) and 0.74 to 5.18 cm d⁻¹ were observed during the summer (May 2016 to October 2016). Four ice cliffs, which all featured supraglacial ponds, persisted over the full study period. In contrast, ice cliffs without a pond or with a steep back-slope degraded over the same period. The rate of thermo-erosional undercutting was over double that of subaerial retreat. Overall, 3D topographic differencing allowed an improved process-based understanding of cliff evolution and cliff-pond coupling, which will become increasingly important for monitoring and modelling the evolution of thinning debris-covered glaciers.

1. INTRODUCTION

In coming decades, ongoing mass loss from Himalayan glaciers and changing runoff trends will affect the water resources of over a billion people in, including those who require it for agricultural, energy production, and domestic usage (Immerzeel and others, 2009; Immerzeel and others, 2010; Lutz and others, 2014; Mukherji and others, 2015; Shea and Immerzeel, 2016). A negative mass balance regime prevails across glaciers in the central and eastern Himalaya (Bolch and others, 2011; Fujita and Nuimura, 2011; Benn and others, 2012; Kääb and others, 2012; Kääb and others, 2015; King and others, 2017), which are widely recognised to be out of equilibrium with current climate (Yang and others, 2006; Shrestha and Aryal, 2011; Salerno and others, 2015). Deglaciation is leading to the development of large proglacial lakes, which may expand rapidly through ice cliff calving (Bolch and others, 2008; Benn and others, 2012; Thompson and others, 2012; Thakuri and others, 2016), and pose potential glacial lake outburst flood hazards (e.g. Carrivick and Tweed, 2013; Carrivick and Tweed, 2016; Rounce and others, 2016; Rounce and others, 2017).

Debris-covered glaciers have a hummocky, pitted surface, caused by variable melt rates under different debris thicknesses, and include extensive coverage of ice cliffs and supraglacial ponds (Hambrey and others, 2008; Thompson and others, 2016; Watson and others, 2016; Watson and others, 2017). Studies using Digital Elevation Model (DEM) differencing to quantify elevation change over debris-covered tongues have revealed an association between glacier surface lowering and the presence of ice cliffs and supraglacial ponds (Immerzeel and others, 2014; Pellicciotti and others, 2015; Ragettli and others, 2016; Thompson and others, 2016), confirming historical ice cliff observations (e.g. Inoue and Yoshida, 1980; Sakai and others, 1998; Benn and others, 2001; Sakai and others, 2002).

1 However, raster-based DEMs generally give a poor representation of steep slopes or steeply-
2 sloping topography (Kolecka, 2012) and their differencing incorporates a mixed signal
3 containing surface elevation change related to debris cover, ice cliff dynamics, supraglacial
4 ponds, and glacier emergence velocity (Vincent and others, 2016).

5 Models of glacier evolution do not consider pond dynamics or ice cliff dynamics
6 explicitly, because this requires an understanding of their spatio-temporal distribution (e.g.
7 Sakai and others, 2002; Watson and others, 2017), energy balance modelling of the ice cliff
8 surface (e.g. Reid and Brock, 2014; Steiner and others, 2015; Buri and others, 2016b; Buri
9 and others, 2016a), and cliff-scale observations of retreat rates (e.g. Brun and others, 2016).
10 Several studies have exploited topographic models derived from unmanned aerial vehicle
11 surveys of Lirung Glacier in the Langtang region of Nepal, to make substantial progress
12 towards understanding ice cliff dynamics (Immerzeel and others, 2014; Steiner and others,
13 2015; Brun and others, 2016; Buri and others, 2016b; Buri and others, 2016a; Miles and
14 others, 2016a). However, techniques to perform direct comparisons of multi-temporal point
15 clouds without simplification have yet to be exploited.

16 In this study, we explore ice cliff evolution using multi-temporal point clouds
17 obtained on Khumbu Glacier, Nepal. Specifically we: (1) quantify the retreat of ice cliffs for
18 pre-monsoon and monsoon time periods; (2) compare the spatial variation in retreat across
19 ice cliff faces; and (3) assess the change in ice cliff morphology through time in relation to
20 local topography and the presence of supraglacial ponds.

21 2. STUDY SITE

22 Field data were obtained on Khumbu Glacier in the Everest region of Nepal during three field
23 campaigns (post-monsoon November 2015, pre-monsoon May 2016, and late-monsoon
24 October 2016). The November 2015 to May 2016 and May 2016 to October 2016 survey
25 intervals are referred to as ‘winter’ and ‘summer’ respectively. The Indian summer monsoon
26 spans the months of June to mid-October (Bollasina and others, 2002; Bonasoni and others,
27 2008) and is when ~80% of annual precipitation falls (Wagnon and others, 2013).

28 Khumbu Glacier is ~17 km long, of which the lower 10 km is debris covered (Fig. 1)
29 and the lower ~4 km is essentially stagnant (Quincey and others, 2009). Supraglacial debris
30 thickness is >2 m in this stagnating region and decreases up-glacier (Nakawo and others,
31 1986; Rowan and others, 2015). However, the thickness of the debris layer is locally
32 heterogeneous owing to the pitted surface and the presence of ice cliffs and supraglacial
33 ponds. We studied nine ice cliffs on the lower debris-covered glacier (Fig. 1), which is a
34 region of particular interest since supraglacial ponds have begun to coalesce here over the
35 past five years (Watson and others, 2016), and a large glacial lake is expected to form (Naito
36 and others, 2000; Bolch and others, 2011; Haritashya and others, 2015).

37 “Fig. 1 near here”

38 3. DATA AND METHODS

39 3.1 Data collection

40 Terrestrial photographic surveys of nine ice cliffs were carried out during the three
41 field campaigns. Our study cliffs represented approximately 2% of the total ice cliff extent on
42 Khumbu Glacier, based on the top-edge cliff delineation of Watson and others (2017). We
43 sought to survey cliffs that were broadly representative of the range of cliffs found on
44 Khumbu Glacier, with and without supraglacial ponds, of variable aspect, and of variable
45 size, noting the terrestrial survey constraints that precluded surveys of very large cliffs. Four
46 of our nine study cliffs had a supraglacial pond present during the initial survey and the mean
47 length of ice cliffs was 57 m. This compares to the observation that on Khumbu Glacier 47%
48 of ice cliffs were associated with a pond in 2015, and cliffs had a mean length of 54 m

1 (Watson and others, 2017). We note from Watson and others (2017) that cliffs 20–40 m in
2 length were most common, but that some cliffs exceeded 200 m in length.

3 Two out of the seven individual study sites (Fig. 1) included both northerly- and southerly-
4 facing cliff faces. These southerly-facing cliffs are labelled ‘-SF’ hereafter. Within the first
5 two field campaigns, surveys were conducted at intervals of 7–11 days at cliffs C, D, E and F,
6 which are referred to as ‘weekly’ surveys. ‘Seasonal’ surveys refer to those between field
7 campaigns. Each survey typically took <1 hour and 122–564 photos were taken of each ice
8 cliff with a highly convergent geometry (Fig. 2a) using a Panasonic DMC-TZ60 18.1
9 megapixel digital camera. In order to capture the surrounding topography, each photo was
10 taken from a different position but was not necessarily orientated towards the ice cliff. High-
11 contrast temporary ground control points (GCPs) (number of GCPs (n) = 6–15) were
12 distributed around each ice cliff to encompass the survey area extents and surveyed using a
13 Leica GS10 global navigation satellite system (GNSS). Each GCP was occupied in static
14 mode for ~5 minutes. A base station was located on the lateral moraine of the glacier <2 km
15 from our survey sites for the duration of each field campaign and was set to record each day.

16 “Fig. 2 near here”

17 3.2 Post-processing

18 Our GNSS base station data were post-processed against the Syangboche permanent station
19 (27.8142 N, 86.7125 E) located ~20 km from our field site using Global Positioning System
20 (GPS) and GLOBal NAVigation Satellite System (GLONASS) satellites. Our field GCPs were
21 then adjusted with reference to the field base station data following a relative carrier phase
22 positioning strategy. The mean 3D positional uncertainty was 3.9 mm across all our GCPs (n
23 = 281).

24 Photographs were input into Agisoft PhotoScan 1.2.3 to derive 3D point clouds of the
25 ice cliff topography following a Structure-from-Motion with Multiview Stereo (SfM-MVS)
26 workflow (e.g. James and Robson, 2012; Westoby and others, 2012; Smith and others, 2015).
27 First, photographs were aligned to produce a sparse point cloud by matching coincident
28 features. This stage also estimated internal camera lens distortion parameters and scene
29 geometry using a bundle adjustment with high redundancy, owing to large overlapping
30 photographic datasets (Westoby and others, 2012). Only points with a reprojection error of
31 <0.6 were retained and clear outliers (e.g. areas of shadow under overhanging cliffs) were
32 removed manually. Second, GCPs were identified in each photograph to georeference the
33 sparse cloud. GCP placement accuracy was <10 mm (e.g. Fig. 2b). Uncertainties from GCP
34 placement and the post-processed coordinates were used as weights to optimise the point
35 cloud georeferencing to minimise root-mean-square error (RMSE) (Javernick and others,
36 2014; Stumpf and others, 2015; Smith and others, 2016; Westoby and others, 2016). Third, a
37 dense point cloud was produced using PhotoScan’s multiview stereo (MVS) algorithm (Fig.
38 3). The dense cloud was subsequently edited to remove points that were not on solid surfaces
39 (e.g. on supraglacial ponds) and clear outliers. All PhotoScan’s processes were run on high
40 quality settings. Georeferencing uncertainty in the final point clouds was <0.035 m (RMSE;
41 Table 1). The final point clouds were sub-sampled using an octree filter in CloudCompare to
42 unify point density across the surveys for each individual ice cliff. Subsequent point cloud
43 densities ranged between 2,185–14,581 points per m².

44 “Table 1. near here”

45 “Fig. 3 near here”

46 3.3 Ice cliff displacement

47 3.3.1 Ice cliff surveys between field campaigns

48 The study cliffs were located down-glacier of the expected active–inactive ice transition on
49 Khumbu Glacier (Quincey and others, 2009). However, small magnitude displacements (<3

1 m a⁻¹) were observed in this region from dGPS surveys of tagged boulders (Supplementary Fig. 1). Therefore, to correct for ice cliff displacement between field campaigns (November 2015 to May 2016 and May 2016 to Oct 2016) we co-registered point clouds using image features that could be identified in multiple surveys (e.g. identifiable marks on boulders). For each survey, the coordinates were derived for these features (as ‘Markers’ in PhotoScan), enabling a transform (2D translation-rotation) to be calculated to co-register the later model with the earlier one. The RMSE between these co-registered features are reported in Table 1, and were used as the total error (E_T) in subsequent point cloud differencing. Co-registration errors were subject to sub-debris melt over each differencing period; however, thick debris cover (1–2 m) (Nakawo and others, 1986) and low sub-debris melt rates of 0.0015 m d⁻¹ (Inoue and Yoshida, 1980) over our study area minimised these errors. Glacier emergence velocity could not be calculated for this study; however, this is expected to be low for the slow-moving and gently-sloping debris-covered study area (Nuimura and others, 2012).

14 3.3.2 Ice cliff surveys within field campaigns

15 To account for cliff displacement between repeat surveys within each field campaign (e.g. Cliff C 03/11/2015 and 10/11/2015), we take the mean daily displacement of the respective cliff between field campaigns (e.g. 0.0023 m d⁻¹) and multiply this by the time-separation of the repeat models (e.g. seven days). The resulting shift (e.g. 0.0161 m) was treated as an additional uncertainty in addition to the respective georeferencing errors. We did not shift these models as described in section 3.3.1, since the expected displacement was <0.04 m for all ice cliffs, which was similar to the uncertainty in identifying coincident features in each model.

23 To calculate the total error (E_T) for each cliff model comparison within a field campaign, the individual errors were propagated using (1):

$$25 \quad (1) E_T = \sqrt{GE_{C1}^2 + GE_{C2}^2 + DE_{C1-C2}^2}$$

26 Where GE_{C1} and GE_{C2} are the georeferencing RMS errors associated with clouds $C1$ and $C2$, and DE_{C1-C2} is the displacement error between clouds $C1$ and $C2$.

28 3.4 Point cloud characteristics and differencing

29 The mean slope and aspect of ice cliffs were calculated in CloudCompare using the dip direction and angle tool. The aspect of overhanging cliff sections required correction through 180°. The area of cliffs was calculated by fitting a mesh to each point cloud using the Poisson Surface Reconstruction tool in CloudCompare (Kazhdan and Hoppe, 2013).

33 Cloud-to-cloud differencing was carried out in the open source CloudCompare software using the Multiscale Model to Model Cloud Comparison (M3C2) method (e.g. Barnhart and Crosby 2013; Lague and others, 2013; Gomez-Gutierrez and others, 2015; Stumpf and others, 2015; Westoby and others, 2016). M3C2 was created by Lague and others (2013) to quantify the 3D distance between two point clouds along the normal surface direction and provide a 38 95% confidence interval based on the point cloud roughness and co-registration uncertainty. The method is therefore ideally suited to quantifying statistically significant ice cliff evolution where the geometry changes in 3D, and is robust to changes in point density and point cloud noise (Barnhart and Crosby, 2013; Lague and others, 2013).

42 For point clouds derived from photogrammetric techniques, uncertainty is spatially variable but highly correlated locally, since points in close proximity to one another are derived from the same images (James and others, 2017). Thus, although point-cloud roughness could represent a component of the measurement precision required to derive confidence intervals, it would not include broader photogrammetric contributions. In order to visualise spatially variable photogrammetric and georeferencing precision, we derived 3D precision maps for May 2016 study cliffs using the method of James and others (2017)

1 (Supplementary Fig. 2). Repeated bundle adjustments implemented in PhotoScan (4000
 2 Monte Carlo iterations) were applied to the sparse point cloud of each ice cliff using GCP and
 3 tie point uncertainties. Point precision estimates were then interpolated onto a 1 m raster grid
 4 in sfm_georef v3.0 (James and Robson, 2012; James and others, 2017). Large uncertainties
 5 were apparent at the survey edges (e.g. Cliff A, C, D, E), and around supraglacial ponds (e.g.
 6 Cliff E) due to poor photograph coverage. In contrast, the mean precision estimates ranged
 7 from 7–38 mm and were uniform across individual cliff faces. Hence, given the large
 8 magnitudes of the measured ice cliff changes, the M3C2-PM (Precision Maps) variant based
 9 on photogrammetry-derived precision maps was not required and the native M3C2 algorithm,
 10 implemented in CloudCompare, was used throughout.

11 M3C2 requires two user-defined parameters: (1) the normal scale D , which is used to
 12 calculate surface normals for each point and is dependent upon surface roughness and point
 13 cloud geometry, and (2) the projection scale d over which the cloud-to-cloud distance
 14 calculation is averaged, which should be large enough to average a minimum of 30 points
 15 (Lague and others, 2013). We estimated the normal scale D for each point cloud following a
 16 trial-and-error approach similar that of Westoby and others (2016), to reduce the estimated
 17 normal error, E_{norm} (%), through refinement of a rescaled measure of the normal scale $n(i)$:

$$(2) n(i) = \frac{D}{\sigma_i(D)}$$

18 $n(i)$ is the normal scale D divided by the roughness σ measured at the same scale around i .
 19 Where $n(i)$ falls in the range 20–25, $E_{norm} < 2\%$ (Lague and others, 2013). In this study,
 20 normal scales D ranged from 1.5–8 m and the projection scale d was fixed at 0.3 m. The
 21 Level of Detection (LoD) threshold for a 95% confidence level is given by:

$$(3) LOD_{95\%}(d) = \pm 1.96 \left(\sqrt{\frac{\sigma_1(d)^2}{n_1} + \frac{\sigma_2(d)^2}{n_2}} + reg \right)$$

23 where σ_1 and σ_2 represent the roughness of each point in sub-clouds of diameter d and size n_1
 24 and n_2 , and reg is the cloud-to-cloud co-registration error, for which E_T is substituted. The
 25 error is assumed to be isotropic and spatially uniform across the dataset (Lague and others,
 26 2013).

27 Distance calculations were masked to exclude points where the change was lower than the
 28 Level of Detection (LoD) threshold and were clipped to individual ice cliff faces. Ice cliff
 29 retreat rates were divided by respective survey intervals to derive daily retreat rates. Since
 30 cliffs often exhibited large changes in geometry between surveys, some cliff normals at time
 31 one intersected debris cover at time two. The total retreat of the cliff face was therefore
 32 reported, in addition to cliff-to-cliff retreat (i.e. where cliff normals at time one intersected a
 33 cliff at time two).

34 3.5 Other data

35 Volumetric loss due to ice cliff retreat was estimated from DEM differencing using point
 36 clouds gridded at 0.5 m. Cliff retreat rates were also calculated using these volumetric
 37 changes for comparison with M3C2 retreat rates, by dividing the volume loss over the
 38 respective time period by the cliff area. The zone of ice cliff retreat was defined as the area
 39 connected by cliff outlines at respective time periods. Where a cliff partially or completely
 40 degraded and hence was not represented by an outline at time two, the zone of cliff retreat
 41 was estimated using M3C2 distance measurements (representing spatially variable ice cliff
 42 retreat) from the cliff at time one. These distance measurements were used to define a
 43 variable-width buffer in ArcGIS to delineate the maximum extent of ice cliff retreat.

44 The drainage of the supraglacial pond adjacent to Cliff E provided an opportunity to
 45 reconstruct the bathymetry and maximum pond depth using the historic water level from May

1 2016, with the assumption that subaqueous basal melt and debris inputs to the basin were
2 minimal. Additionally, air temperature at 1 m above the surface was recorded at 20 minute
3 intervals on Khumbu Glacier using a Solinst Barologger Edge, which was sited behind Cliff
4 G. Measurements were recorded from October 2015 to October 2016; however, the station
5 collapsed in August 2016 due to the retreat of Cliff G. The logger was found in an air pocket
6 buried by debris, hence data shown after this collapse revealed a subdued diurnal temperature
7 cycle.

8 4. RESULTS

9 4.1 Summary ice cliff characteristics

10 Ice cliffs: ranged from 4 to 23 m in height; were all within a 43 m elevation range; had mean
11 slopes of 50 to 73°; and were of variable aspect, with both northerly- and southerly-facing
12 cliffs represented (Fig. 4, Table 2). Of the four cliffs with a supraglacial pond present in
13 November 2015, only Cliff B had a pond remaining in October 2016. Overall, four study
14 cliffs persisted throughout the study period and the other five were buried under debris
15 between May 2016 and October 2016 (Fig. 4c).

16 “Table 2. near here”

17 The mean slope, maximum cliff height, cliff area, and mean cliff aspect were
18 evaluated across our study period (Fig. 4). Southerly-facing cliffs generally featured higher
19 mean slopes, and the greatest changes in cliff slope were observed on southerly-facing cliffs
20 B-SF and E (Fig. 4a). Maximum cliff height reduced for all cliffs over the study period,
21 although this change was generally small for those cliffs that persisted through the study.
22 Cliff E, which lost ~5 m in height over summer (Fig. 4b), was an exception. Notably,
23 persistent cliffs had a starting height greater than 10 m; however, we note that Cliff F
24 decayed, despite a starting height greater than 10 m. With the exception of Cliff B-SF which
25 increased in area, all other cliffs declined in area over the study; however, the rate of this area
26 loss varied from cliff to cliff. Of the four cliffs persisting over the study, two were southerly-
27 facing and two were northerly-facing and all had a supraglacial pond present for part of the
28 study period (Table 2). However, pond dynamics between the observation dates were
29 unknown. The largest changes in cliff aspect were for cliffs C and D-SF, which became
30 increasingly northerly and westerly orientated by 25° and 23° respectively (Fig. 4d).

31 “Fig. 4 near here”

32 4.2 Ice cliff retreat

33 With the exception of Cliff D-SF, cliff retreat rates were higher over summer than the
34 preceding winter, which corresponds with consistently higher air temperatures during
35 summer (Fig. 5, Table 3). Mean winter temperatures were generally below 0°C, whereas
36 summer temperatures were generally above 0°C, although several discrete periods of positive
37 air temperature also occurred in winter. Similarly, volumetric losses due to cliff retreat were
38 generally higher during summer, although they were small where cliffs degraded (e.g. D-SF,
39 Table 3). Notably, the M3C2- and DEM-based retreat rates were comparable for most cliffs.

40 The highest mean retreat rate occurred at Cliff B and B-SF during summer, although this was
41 a combination of subaerial retreat and a large-scale cliff collapse involving a section of the
42 cliff ~30 m in length. Excluding this cliff face, the highest mean retreat rates observed and
43 the largest seasonal differences in retreat rates were from ice cliffs with an adjacent
44 supraglacial pond including cliffs A (1.75 cm d⁻¹), E (4.26 cm d⁻¹) and G (2.62 cm d⁻¹) (Fig.
45 5b). The retreat rates for weekly surveys were generally higher than seasonal retreat rates,
46 with the exception of Cliff E (Fig. 5). Retreat rates for cliffs that degraded over the study
47 period (A, C, D, D-SF, and F) included a transition to sub-debris melt during the summer,
48 and hence were expected to have lower retreat rates and volume losses compared to persistent

1 cliffs. Similarly, where cliffs partially degraded between survey intervals, the cliff-to-cliff
2 retreat was generally greater than the total retreat rates, which included areas of cliff-to-debris
3 transition (Table 3). Here, retreat rates attributed only to persistent areas of cliff ranged from
4 $0.36\text{--}1.68\text{ cm d}^{-1}$ (winter), and $3.84\text{--}5.85\text{ cm d}^{-1}$ (summer).

5 “Fig. 5 near here”

6 “Table 3 near here”

7 Across individual cliff faces, the observed retreat was related to the presence of supraglacial
8 ponds, englacial conduits (expressed as an opening within or below ice cliff faces), cliff
9 aspect, cliff slope, and the formation of runnels (Fig. 6). Mean winter ice cliff retreat showed
10 a clear relationship with aspect and peaked at a south easterly aspect of 155° , although this
11 was not the case during summer (Fig. 6h). Maximum retreat rates (10.65 cm d^{-1}) were
12 observed at cliffs B and B-SF (Fig. 6b) where a notable calving event occurred in the
13 summer. This was followed by northerly-facing Cliff G in association with a supraglacial
14 pond, with maximum retreat rates of 6.18 cm d^{-1} in the zone of thermal undercutting (Fig.
15 6g). The surface of this pond was frozen between the November 2015 and May 2016 surveys
16 and the pond had drained by October 2016.

17 Runnels were locations of locally differential retreat on the north-facing cliffs B and
18 G during winter (Fig. 6b, g) and these cliffs also had the lowest mean initial slopes of 54° and
19 50° respectively (Table 2). During winter, the relative increase in retreat at Cliff G at
20 locations of runnels was $\sim 0.12\text{--}0.24\text{ cm d}^{-1}$ (Fig. 6g). However, the presence of runnels was
21 localised and, when considering the whole cliff, the rate of runnel retreat ($\sim 0.47\text{--}0.58\text{ cm d}^{-1}$)
22 was otherwise comparable to the mean retreat rate during winter (0.47 cm d^{-1}). Evidence of a
23 vertical retreat gradient was apparent on several cliffs during winter (e.g. Fig. 6c, d, f), and
24 was similar during summer, other than where thermal undercutting was apparent (e.g. Cliff
25 G). Cliff B-SF featured the most apparent aspect-related control on retreat during winter with
26 westerly facing ice melting at the slowest rate ($\sim 0.84\text{ cm d}^{-1}$) compared to southerly faces
27 ($\sim 2.62\text{ cm d}^{-1}$, Fig. 6b). Cliff A featured an englacial conduit large enough to enable crouched
28 access into a void behind the cliff face, which was the area of greatest retreat for this cliff
29 ($\sim 4\text{ cm d}^{-1}$ during summer) as the void likely became exposed and degraded (Fig. 6a).

30 “Fig. 6 near here”

31 **4.3 Ice cliff evolutionary traits**

32 2D profiles through selected ice cliffs revealed different characteristics of retreat, including
33 ice cliff burial under debris, ice cliff collapse, and undercutting by adjacent supraglacial
34 ponds (Fig. 7). Cliff A maintained a similar slope during its retreat over winter, although
35 during summer the slope angle decreased to $\sim 35^\circ$, which led to burial under debris (Fig. 7a).
36 There was a small supraglacial pond shallower than 1 m adjacent to Cliff A in the November
37 2015 and May 2016 surveys, and an associated undercut notch. The pond drained prior to the
38 October 2016 survey, and the steep cliff back-slope led to a relaxation of the cliff slope and
39 hence degradation of Cliff A by October 2016.

40 The profile through Cliff B revealed greater retreat on the southerly face through
41 winter, compared to the northerly face (Fig. 7b). A section of Cliff B collapsed prior to the
42 final survey in October 16, suggesting that the undercut notch caused the cliff to collapse in a
43 northerly direction. The supraglacial pond in contact with Cliff B-SF expanded throughout
44 the study and was in contact with the southerly face (although not at the 2D profile shown),
45 which exposed a new ice cliff face and caused an increase in cliff area (Fig. 4c). A
46 supraglacial pond was present at the northerly-facing Cliff B in May 2016; however, the
47 water level was well below historic water cut notches.

1 The two opposing faces of Cliff D and D-SF both became buried over the study (Fig.
2 7c). The southerly-facing cliff retreated faster than the northerly face during winter; however,
3 the steep back-slope and small area of this southerly face limited the retreat during summer.
4 Debris infill was apparent in the May 2016 profile, caused by the retreat of the southerly face,
5 which had an inwardly sloping cliff top.

6 Cliff E featured a supraglacial pond over 9.95 m deep, which drained over the
7 summer. There was evidence of deepening towards the cliff faces at profiles 1 and 2 and
8 thermal undercutting of both cliff faces (Fig. 8c, d). The pond at Cliff G also drained over
9 the summer and was also associated with an undercut notch. Cliff G had a gentle back-slope
10 and maintained a similar slope (50–54°) during retreat (Fig. 7d). The gentle back-slope of
11 Cliff G allowed continued retreat, in contrast to cliffs A and D where a steep back-slope lead
12 to cliff degradation.

13 “Fig. 7 near here”

14 “Fig. 8 near here”

15 5. DISCUSSION

16 5.1 Multi-temporal ice cliff surveys

17 In this study we have presented the first application of 3D point cloud differencing to multi-
18 temporal topographic surveys of ice cliffs, revealing evolutionary traits for a variety of ice
19 cliffs present on the lower ablation area of Khumbu Glacier. This method has specific
20 advantages for quantifying the importance of ice cliff retreat and the cliff–pond interaction
21 when compared to previous approaches. First, the retreat attributed to ice cliffs is calculated
22 along the normal direction of the cliff face, thereby minimising the conflation of topographic
23 change from debris cover, ice cliffs and supraglacial ponds, that exists in vertical DEM
24 differencing (e.g. Thompson and others, 2016). However, where a cliff decays between two
25 survey dates, retreat calculations include topographic change related to processes in addition
26 to cliff retreat such as sub-debris melt, hence short survey intervals are preferable. M3C2
27 allows quantification of the variability of retreat across a cliff face, for example in relation to
28 slope and aspect (Buri and others, 2016a), the presence of runnels (Watson and others,
29 2017), and supraglacial ponds (Miles and others, 2016a). Second, the mechanism controlling
30 topographic change can be evaluated in three dimensions, revealing the role of ice cliff back-
31 slope in ice cliff persistence, and thermo-erosional undercutting by supraglacial ponds.

32 5.2 Ice cliff retreat

33 Observations of ice cliff retreat have previously been obtained from point ablation stake
34 measurements or using static markers on the back-slopes of ice cliffs (e.g. Inoue and
35 Yoshida, 1980; Sakai and others, 1998; Purdie and Fitzharris, 1999; Sakai and others, 2002;
36 Benn and others, 2001; Han and others, 2010; Reid and Brock, 2014; Steiner and others,
37 2015). Use of ablation stakes restricts assessment of the spatial variation in retreat across an
38 ice cliff face, since stake placements are likely to be aligned vertically down the cliff face and
39 biased towards areas of comparatively safer access. In comparison, Brun and others (2016)
40 used multi-temporal fine-resolution topographic surveys to estimate the volumetric mass loss
41 and mean retreat rates from five ice cliffs on the debris-covered Lirung Glacier. Their study
42 cliffs were generally larger (maximum face size of 6441 m² compared to 1313 m² in this
43 study), and at ~800 m lower elevation.

44 5.2.1 Ice cliff retreat through time

45 Four of nine ice cliffs, which all had a maximum height >10 m (Fig. 4b) and featured an
46 adjacent supraglacial pond (Table 2), persisted over one year in this study. In contrast, all
47 study cliffs of Brun and others (2016) persisted over one year and were all ≥9 m high. Mean
48 retreat rates obtained in this study ranged from 0.30 to 1.49 cm d⁻¹ (winter), and 0.74 to 5.18

1 cm d⁻¹ (summer) and were comparable to those of Brun and others (2016), 0.70 to 1.20 cm d⁻¹
2 (winter) and 2.2 to 4.5 cm d⁻¹ (summer), despite the higher elevation and smaller size of our
3 study cliffs. In our study, ice cliff retreat rates were generally higher during summer and for
4 southerly-facing ice cliffs, and summer featured the largest variability in retreat rates amongst
5 cliffs (Fig. 5b, c). Lower retreat rates during winter correspond with cooler air temperatures
6 (Fig. 5a). Higher retreat rates during the November 2015 weekly surveys compared to
7 respective seasonal surveys, reflected warmer temperatures before winter. Similarly, the May
8 2016 surveys generally had higher retreat rates than the May 2016 to October 2016 surveys
9 (Fig. 5b, c). For cliffs C, D and F, this likely reflects cliff degradation before October 2016
10 and hence a transition from cliff retreat to sub-debris melt for part of the survey interval. The
11 unknown date of cliff degradation is a limitation to assessing mass loss relating only to ice
12 cliff retreat. However, the retreat rates for cliffs that partially degraded between surveys can
13 be quantified by considering the total cliff retreat, which includes areas where the cliff
14 degraded, and the cliff-to-cliff retreat, where cliff normals at time one intersect a cliff at time
15 two (Table 3, Supplementary Table 1). The former is representative of the total cliff
16 evolution, which includes areas of ice cliff degrading and becoming buried by debris,
17 whereas the latter is representative of the retreat rates for persisting areas of cliff.

18 Cliff B suffered a partial collapse of a ~30 m segment during summer, causing high mass
19 loss due to the effect of this calving event (Fig. 5b, Table 3). Similarly high retreat rates May
20 2016 to October 2016 were observed at cliffs E and G, which both featured supraglacial
21 ponds becoming active (i.e. thawed) during summer. The high variation in retreat rates over
22 summer suggests that more frequent monitoring would be beneficial to assess cliff dynamics
23 such as burial under debris and the interaction with seasonally expanding supraglacial ponds.

24 5.2.2 *Cliff face variation in retreat*

25 Visualising the spatial distribution of retreat across individual cliff faces revealed vertical and
26 lateral gradients, and increased retreat attributed to supraglacial ponds during summer (Fig.
27 6). The influence of cliff aspect is apparent on Cliff B-SF, where a southerly face retreated
28 1.78 cm d⁻¹ faster than an adjacent westerly face (Fig. 6b). Both the north-facing cliffs B and
29 G displayed evidence of locally enhanced retreat attributed to the presence of vertical runnels
30 observed during the winter surveys (Fig. 6b, g). Runnels were also observed by Watson and
31 others (2017) on other ice cliffs on Khumbu Glacier. The low solar radiation receipt on
32 northerly-facing cliffs during winter may mean that meltwater generated at the less shaded
33 cliff top from sub-debris melt and from melt on the cliff face is able to incise runnels faster
34 than the background rate of cliff retreat. In contrast, during summer the higher magnitude of
35 retreat likely masks this influence of micro-scale cliff topography (Fig. 6), although the
36 runnels may persist. Runnels also act as preferential pathways for debris slumping from the
37 cliff top and differential retreat may also occur in response to albedo variations across the
38 cliff face. The morphology of the cliff face, including runnel development and self-shading,
39 is therefore likely to locally influence retreat rates as evidenced in this study, but should be
40 explored further with additional 3D surveys in order to assess their importance seasonally,
41 and at a glacier scale.

42 Cliff tops exhibited the highest retreat rates in several cases (e.g. Fig. 6c, d, f), which was
43 also observed in the modelled retreat rates at two ice cliffs by Buri et al. (2016b). Cliff G also
44 displayed a vertical gradient during the summer; however, this was locally reversed in the
45 area undercut by a supraglacial pond (Fig. 6g).

46 5.2.3 *The influence of aspect*

47 Several studies have observed a prevalence of northerly-facing ice cliffs on debris-covered
48 glaciers in the Northern Hemisphere, suggesting that solar radiation receipt plays a key role
49 in controlling ice cliff development (Sakai and others, 2002; Kraaijenbrink and others, 2016b;
50 Thompson and others, 2016; Watson and others, 2017). Southerly-facing ice cliffs are

1 expected to decay quickly after formation due to high solar radiation receipt, whereas
2 northerly-facing cliffs are more persistent (Sakai and others, 2002). Slope relaxation was
3 apparent on southerly-facing cliffs B-SF and E, which decreased by 14 and 13° respectively;
4 however, both of these cliffs persisted throughout the study period.

5 We observed highest ice cliff retreat rates on ice cliffs with a southeasterly aspect (155°)
6 (Fig. 6h), although this trend was only apparent during winter. Also on Khumbu Glacier,
7 Inoue and Yoshida (1980) revealed maximum ice cliff retreat at an aspect of ~190°. Cliff
8 aspect likely has a stronger influence over cliff retreat in the winter due to the low solar angle
9 and cliff self-shading (e.g. Steiner and others, 2015). Additionally, direct solar radiation
10 receipt is reduced during the summer monsoon due to the prevalence of cloud cover
11 (Supplementary Fig. 4). Therefore at this time, diffuse radiation, air temperature, and local
12 ice cliff characteristics such as the presence of a supraglacial pond were likely stronger
13 controls on ice cliff retreat than the cliff aspect.

14 5.3 Local controls on ice cliff evolution

15 The back-slope of individual ice cliffs influences their longevity, since there is a finite
16 volume of ice for the cliff to retreat into unless accompanied by simultaneous downwasting
17 of a supraglacial pond. In our study, slope relaxation and cliff degradation (Fig. 4a, 7a and c)
18 were observed on both northerly- and southerly-facing ice cliffs. This contrasts with the
19 observations of Sakai and others (2002) where slope relaxation was a trait of southerly-facing
20 ice cliffs, highlighting the importance of local topography and cliff characteristics, which
21 determine the longevity of individual ice cliffs over an ablation season.

22
23 Several studies have observed strong spatial coincidence of ice cliffs and supraglacial
24 ponds (Thompson and others, 2016; Watson and others, 2017) and notable subaqueous melt
25 rates (Sakai and others, 2009; Miles and others, 2016a). The potential importance of ponds
26 for enhancing ice cliff retreat on Himalayan debris-covered glaciers is analogous to the
27 'wandering lakes' on Antarctic ice-cored moraines (e.g. Pickard, 1983). Thompson and
28 others (2016) observed that 75% of ice cliffs were associated with a supraglacial pond on the
29 Ngozumpa Glacier, and an average of 49% was observed by Watson and others (2017) across
30 14 glaciers in the Everest region of Nepal; however, these associations are likely to be
31 seasonally variable. Our study revealed greater retreat for ice cliffs associated with a
32 supraglacial pond, and mean retreat rates of pond-contact ice were estimated to be double that
33 of subaerial retreat at Cliff G. However, the pond at Cliff G drained prior to the final survey
34 such that the role of the pond could not be fully isolated from subaerial retreat. All persisting
35 ice cliffs featured a supraglacial pond during their lifespan. We suggest that undercut notches
36 allowed the cliff angle to be maintained during retreat, which promoted cliff persistence (e.g.
37 Fig 7d, 8d). Therefore our observations, in addition to strong association of cliffs and ponds
38 (e.g. Watson and others, 2017), suggest that supraglacial ponds are likely to play a key role in
39 ice cliff retreat and persistence at a glacier scale. However, quantifying subaqueous retreat
40 using point cloud differencing is hindered by submerged topography, and manual field
41 measurements (e.g. Rohl, 2006) are restricted by falling debris. Additionally, we cannot
42 comment on the spatial variation in the importance of ice cliff retreat, which likely decreases
43 with distance up-glacier using due to thinning debris cover (Thompson and others, 2016;
44 Watson and others, 2017).

45 5.4 Implications for mass loss at a glacier scale

46 The ice cliff retreat rates observed in this study support previously observed associations
47 between glacier surface lowering and the presence of ice cliffs and supraglacial ponds
48 (Immerzeel and others, 2014; Pellicciotti and others, 2015; Thompson and others, 2016;
49 Watson and others, 2017). Observed mean ice cliff retreat rates ranged from 0.30 (winter) to
50 5.18 cm d⁻¹ (summer), which is much greater than sub-debris melt of 0.15 cm d⁻¹ (Aug-1978)

1 observed in a similar region of Khumbu Glacier by Inoue and Yoshida (1980). However, we
2 note that these rates are not directly comparable since our observations represent surface-
3 normal retreat, whereas sub-debris melt represents vertical lowering. The rate of surface
4 lowering related to debris cover ranged from 0.03–0.31 cm d⁻¹ on the nearby Ngozumpa
5 Glacier based on the DEM differencing of Thompson and others (2016). However, surface
6 lowering observed from DEM differencing is a function of sub-debris melt and emergence
7 velocity. The latter was not quantified by Thompson et al. (2016), or in this study, but was
8 shown to be +0.37 m w.e a⁻¹ (water equivalent) on the debris-covered Changri Nup Glacier
9 (Vincent and others, 2016).

10
11 The volumetric loss at ice cliffs is variable and highlights the requirement to up-scale our
12 methodology to the glacier scale in order to capture the full size distribution of ice cliffs
13 present (Table 3). Additionally, knowledge of fine spatio-temporal dynamics of supraglacial
14 ponds is still limited, but reveals potentially large seasonal expansion and contraction of
15 ponds (e.g. Miles and others, 2016b; Watson and others, 2016). This restricts efforts to model
16 the ice cliff–pond interaction (Buri and others, 2016a), or to quantify subaqueous retreat with
17 multi-temporal point clouds. Nonetheless, our results suggest that undercut notches can
18 promote ice cliff persistence by maintaining the slope angle during retreat. However, this
19 requires further investigation at a glacier scale and over longer time periods, with particular
20 attention to the role of undercutting for promoting calving events. A SfM-MVS methodology
21 using time-lapse imagery is one such approach that could provide high temporal resolution.

22 5.4.1 Future work

23 M3C2 offers new opportunities to quantify 3D topographic change on debris-covered glaciers
24 and this could be used to explore debris redistribution and the formation of ice cliffs, which
25 are currently limiting factors in modelling efforts (Buri and others, 2016a). Similarly, using
26 point cloud data and M3C2 can address several problems related to fine spatio-temporal
27 resolution DEM differencing, including the conflation of several processes contributing to the
28 topographic change signal, including ice cliff retreat, sub-debris melt, and supraglacial pond
29 basal melt. Debris thickness estimated along the top edge of the cliff could be accounted for
30 when slumping into supraglacial ponds, which may otherwise be counted as mass loss in
31 DEM differencing. Comparisons of coincident measurements of 3D and 2D topographic
32 change would therefore be highly beneficial to fully quantify their limitations. Additionally,
33 the topographic change could be explored further with a greater dataset of ice cliff
34 observations, to quantify specific relationships between cliff retreat and variables such as
35 local slope, aspect, and pond presence. However, our dataset demonstrates that ice cliff
36 evolution is highly heterogeneous and that, when considering the dataset as a whole, the
37 relationship between cliff retreat and slope, aspect, and pond presence would be highly
38 complex. Moving forward, conceptualising ice cliff evolution requires both local
39 observations as presented in this study, and glacier scale multi-temporal ice cliff datasets (e.g.
40 Watson and others, 2017).

41 The M3C2 method is not without its own limitations since it is difficult to calculate
42 volumetric mass loss due to the variable alignment of surface normals; however, such
43 methods are likely to become available or can be developed independently for similar
44 applications (e.g. Brun and others, 2016). Non-uniform glacier surface displacement also
45 presents issues when co-registering multi-temporal point clouds; however, this is arguably
46 easier to achieve than using a DEM due to the availability of true-colour point data. However,
47 DEMs and corresponding orthophotos can also be used for this correction (e.g. Kraaijenbrink
48 and others, 2016a). Non-uniform glacier surface displacement is an important consideration if
49 investigating lower magnitude processes such as sub-debris melt, which also requires
50 quantification of glacier emergence velocity (Vincent and others, 2016). Emergence velocity
51 is expected to be low on slow-moving low gradient debris-covered glacier tongues (Nuimura

1 and others, 2011); however, positive surface elevation change was observed in this study (e.g.
2 Fig. 6f), which was confirmed by independent dGPS boulder surveys (used in Supplementary
3 Fig. 1). Additionally, point cloud precision estimates based on rigorous photogrammetric
4 processing rather than surface roughness allow improved topographic change detection
5 (James and others, 2017).

6 6. CONCLUSIONS

7 We have presented the first multi-temporal 3D analysis of ice cliff evolution using 3D point
8 cloud differencing, which was necessary to quantify the spatial heterogeneity of retreat across
9 individual cliff faces and their interaction with supraglacial ponds. Our results revealed the
10 importance of a gentle cliff back-slope to allow continued retreat, and the role of supraglacial
11 ponds in thermo-erosional undercutting, which maintains the cliff angle and delays burial
12 under debris. Mean ice cliff retreat rates observed in this study ranged from 0.30 to 1.49 cm
13 d^{-1} (winter), and 0.74 to 5.18 cm d^{-1} (summer). Additionally, the four ice cliffs persisting over
14 our one year study period were all influenced by supraglacial ponds, and pond-contact ice
15 was associated with a two-fold increase in retreat at Cliff G. Our findings add further
16 evidence to the role of ice cliffs as ‘hot-spots’ of mass loss on heavily debris-covered glaciers
17 and contribute to a previously sparse dataset of ice cliff observations, revealing local controls
18 on cliff retreat which can be used to validate emerging models of ice cliff evolution (Brun
19 and others, 2016; Buri and others, 2016a).

20 We observed an aspect-related control on ice cliff retreat during winter; however,
21 local ice cliff characteristics masked any cliff-scale aspect related control on retreat during
22 summer. We observed examples of northerly- and southerly-facing cliffs persisting, but also
23 examples of cliff burial under debris. The controlling factors for ice cliff persistence appeared
24 to be cliffs with a maximum height >10 m and with supraglacial pond influence. Nonetheless,
25 the prevalence of northerly-facing cliffs on debris-covered glaciers in the northern
26 hemisphere (Sakai and others, 2002; Brun and others, 2016; Watson and others, 2017)
27 suggests that over longer timescales (e.g. decadal) the persistence of northerly-facing cliffs is
28 greater in response to self-shading and supraglacial pond association.

29 M3C2 point cloud differencing was shown to be an effective tool to quantify the spatio-
30 temporal magnitude of retreat across ice cliff faces, and to offer a new opportunity to validate
31 models of ice cliff evolution. It is also more practical than point-based ablation stake
32 measurements. M3C2 could be applied to glacier scale point clouds to enable surface
33 elevation change to be partitioned into surface-normal (ice cliff retreat) and vertical (sub-
34 debris and subaqueous melt) components, and should be compared to the prevailing practice
35 of DEM differencing. These 3D point cloud data provide a much more realistic representation
36 of surface area compared to a planimetric DEM, and minimise the conflation of different
37 topographic change signals that are common to DEM differencing.

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12

13

14 **Fig. 1.** Ice cliffs and supraglacial ponds on Khumbu Glacier (a), located in eastern Nepal (b).
 15 Inset boxes show the location and ID of the ice cliffs surveyed (c). Cliff sites B and D
 16 included both northerly- and southerly-facing ice cliffs. The panchromatic background image
 17 is from the Pleiades satellite (07/10/2015), and corresponding ice cliffs and ponds are shown.
 18 Khumbu Glacier flows in a southerly direction.

19 **Fig. 2.** The generation and georeferencing of ice cliff point clouds. Photographs of each ice
 20 cliff (a) were aligned to produce a sparse point cloud, which was georeferenced using high-
 21 contrast pink and yellow markers (b). Dense point clouds were produced and manually edited
 22 to remove points not on solid surfaces (e.g. supraglacial ponds) (c).

23 **Fig. 3.** Oblique views of the 3D ice cliff point clouds in November 2015. Cliff IDs
 24 correspond to Table 1 and Figure 1. The profiles (red lines) correspond to Figure 7.

25 **Fig. 4.** The evolution of ice cliff mean slope (a), maximum height (b), area (c), and aspect (d)
 26 over the study period. Absolute cliff area change is shown in Supplementary Fig. 3.

27 **Fig. 5.** (a) Air temperature at 1 m above the surface recorded at 20 minute intervals with a
 28 seven day moving average. Survey intervals are indicated by vertical black lines. The logger
 29 mounting collapsed due to ice cliff retreat in August 2016 (shaded area represents data when
 30 the logger was partially buried by debris). Mean ice cliff retreat rates for the seasonal (b), and
 31 weekly surveys (c). Error bars show one standard deviation.

32 **Fig. 6.** Ice cliff retreat rates shown for winter (November 2015 to May 2016) and summer
 33 (May 2016 to October 2016). Note the different scale ranges. Distance measurements are
 34 clipped to the study cliffs and indicative values are shown for key features. The mean and
 35 standard deviation of non-cliff surface elevation changes are reported for winter (w) and
 36 summer (s). Ice cliff retreat rate and initial aspect for winter and summer differencing periods
 37 are shown in (h), with a sinusoidal regression line (winter). Circled points indicate ice cliffs
 38 that disappeared during summer.

39 **Fig. 7.** 2D ice cliff profiles for selected cliffs revealing topographic change over the study
 40 period. Ice cliff faces are shown as lines without a transparency, whereas debris-covered
 41 areas and water levels are shown with transparency.

42 **Fig. 8.** The drainage of a supraglacial pond interaction at Cliff E (a). The drained supraglacial
 43 pond provided an opportunity to reconstruct the historic bathymetry (b and c). The data gap at
 44 the deepest part of the pond (intersecting with Profile 1) was caused by the remnant presence
 45 of water, which had not drained, estimated to be <1 m in depth. Point cloud profiles revealed
 46 subaerial ice cliff retreat and thermos-erosional undercutting (d). The yellow star denotes an
 47 area of the cliff that was present in November 2015 (a), but had degraded by May 2016 (d).

1

2 **Table 1.** Summary statistics for each ice cliff model

Cliff ID	Survey date	Georeferencing RMSE (m) / (number of GCPs)	Tie point RMSE (pixels)	Co-registration RMSE (m) / (number of GCPs)	
				May 2016 to Nov 2015	Oct 2016 to May 2016
A	08/11/2015	0.017 (10)	0.96	*	
	13/05/2016	0.015 (7)	0.75	*0.050 (7)	*
	05/10/2016	0.014 (8)	0.73		*0.262 (8)
B	03/11/2015	0.025 (14)	0.97	*	
	16/05/2016	0.013 (14)	0.98	*0.070 (12)	*
	05/10/2016	0.012 (14)	0.84		*0.334 (11)
C	03/11/2015	0.012 (6)	1.05	*	
	10/11/2015	0.016 (7)	1.03		
	16/05/2016	0.011 (7)	0.92	*0.025 (7)	*
	25/05/2016	0.013 (7)	0.87		
	03/10/2016	0.016 (6)	0.86		*0.183 (6)
D	01/11/2015	0.016 (8)	1.05	*	
	10/11/2015	0.012 (8)	0.96		
	15/05/2016	0.011 (8)	0.80	*0.100 (6)	*
	25/05/2016	0.011 (8)	0.71		
	03/10/2016	0.011 (7)	0.68		*0.176 (6)
E	01/11/2015	0.034 (15)	0.87		
	12/11/2015	0.011 (15)	0.85	*	
	15/05/2016	0.018 (15)	0.84	*0.122 (9)	*
	26/05/2016	0.014 (15)	0.81		
	04/10/2016	0.013 (14)	0.89		*0.140 (9)
F	04/11/2015	0.012 (8)	1.14	*	
	14/11/2015	0.017 (9)	1.02		
	14/05/2016	0.008 (9)	1.05	*0.041 (8)	*
	25/05/2016	0.010 (9)	0.96		
	04/10/2016	0.007 (8)	1.02		*0.186 (8)
G	07/11/2015	0.017 (7)	0.98	*	
	15/05/2016	0.004 (6)	0.91	*0.031 (7)	*
	03/10/2016	0.011 (8)	1.01		*0.224 (7)

*For point cloud differencing, May 2016 data were co-registered with November 2015, and October 2016 were co-registered with May 2016

3

4 **Table 2.** Ice cliff characteristics in November 2015

Cliff ID	Elevation (m)	Surface area (m ²)	Maximum height (m)	Mean slope (°)	Mean aspect (°)
A ^{1,2}	4959	85	5	56	5
B ²	4939	1,278	17	54	37
B-SF ^{1,2,3}	4933	1,313	23	73	235
C	4941	34	4	58	266
D	4933	56	4	57	15
D-SF	4935	37	5	63	210

E ^{1,2}	4926	782	15	65	161
F*	4916	757	13	58	132
G ^{1,2}	4923	357	10	50	1

^{1,2,3} indicate the presence of a supraglacial pond during (1) November 2015, (2) May 2016, and (3) October 2016 surveys. *A supraglacial pond was present at Cliff F prior to field surveys based on Pleiades imagery (Fig. 1)

1

2 **Table 3.** Mean ice cliff retreat rates and volume loss for winter and summer

Cliff ID	Nov 2015 to May 2016				May 2016 to Oct 2016			
	M3C2 retreat (cm d ⁻¹)	M3C2 retreat (cliff only) ¹ (cm d ⁻¹)	DEM-based volume loss (m ³)	DEM-based retreat (cm d ⁻¹)	M3C2 retreat (cm d ⁻¹)	M3C2 retreat (cliff only) ¹ (cm d ⁻¹)	DEM-based volume loss (m ³)	DEM-based retreat (cm d ⁻¹)
A	0.46	0.46	68	0.30	1.75	-	420	1.20
B	0.65	0.70	797	0.66	5.18	5.85	12426*	5.53
B-SF	1.49	1.55	3286	3.05	4.05	3.84		
C	0.39	0.58	22	0.36	0.95	-	38	0.97
D	0.30	0.36	28	0.34	1.27	-	63	1.18
D-SF	0.74	0.92	69	0.80	0.74	-	5	0.33
E	1.44	1.68	1779	2.58	4.26	4.70	2950	4.55
F	1.26	1.57	1238	1.32	1.39	-	643	1.93
G	0.47	0.49	315	0.63	2.62	4.10	1786	2.44

1. M3C2 retreat rates (cliff only) represent cliff-to-cliff retreat i.e. excluding areas where ice cliff normals at time one intersected with debris cover at time two (see Supplementary Table 1).

* Volume loss could not be separated at B and B-SF due to a calving event.

3

4

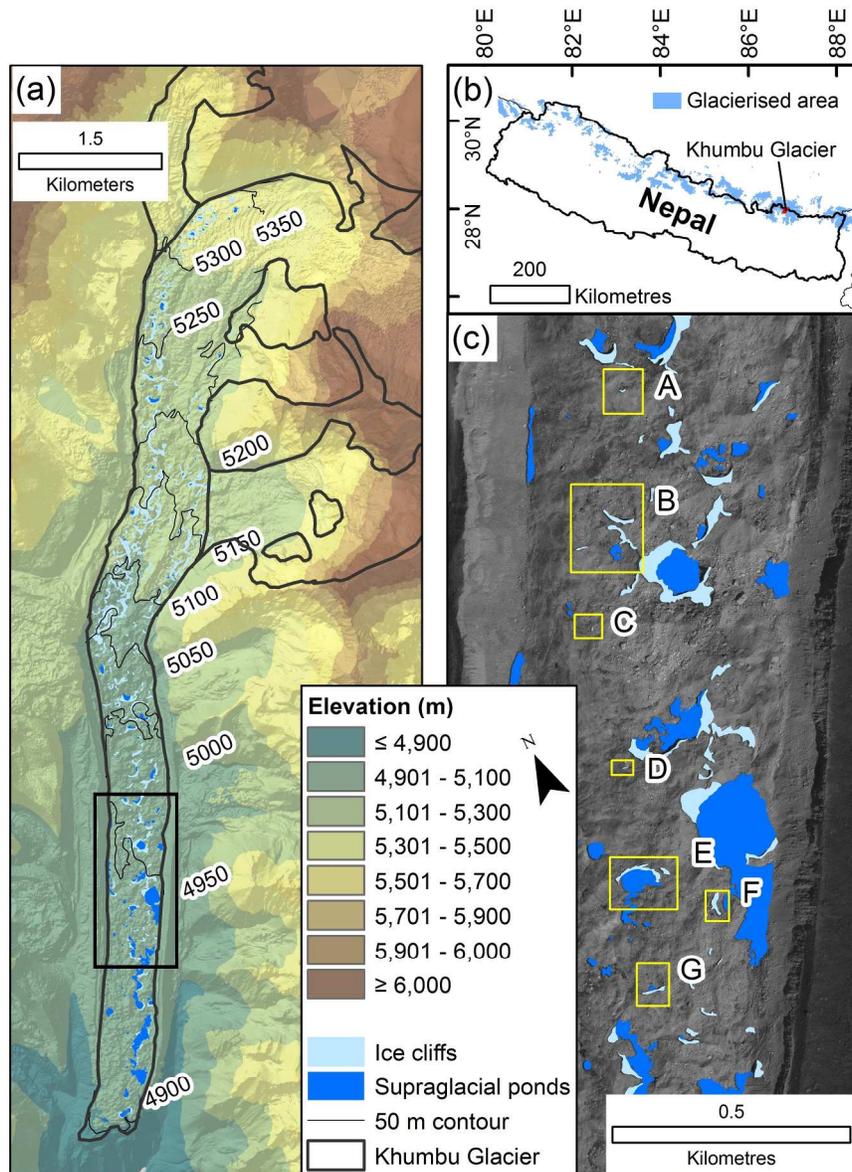


Fig. 1. Ice cliffs and supraglacial ponds on Khumbu Glacier (a), located in eastern Nepal (b). Inset boxes show the location and ID of the ice cliffs surveyed (c). Cliff sites B and D included both northerly- and southerly-facing ice cliffs. The panchromatic background image is from the Pleiades satellite (07/10/2015), and corresponding ice cliffs and ponds are shown. Khumbu Glacier flows in a southerly direction.

200x276mm (300 x 300 DPI)

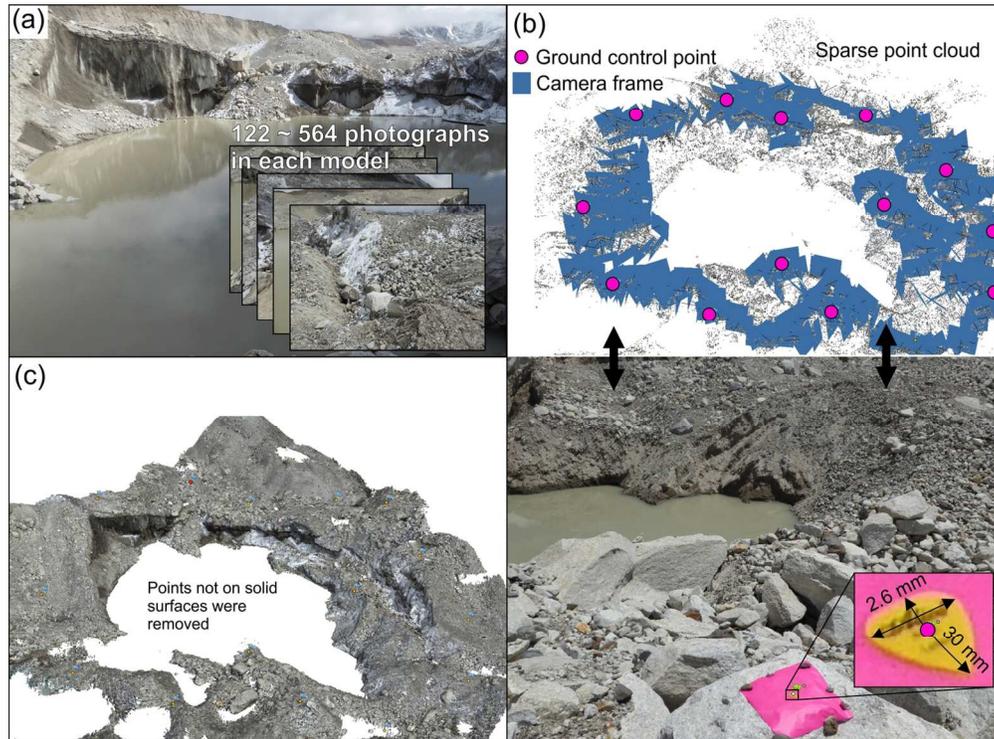


Fig. 2. The generation and georeferencing of ice cliff point clouds. Photographs of each ice cliff (a) were aligned to produce a sparse point cloud, which was georeferenced using high-contrast pink and yellow markers (b). Dense point clouds were produced and manually edited to remove points not on solid surfaces (e.g. supraglacial ponds) (c).

130x96mm (300 x 300 DPI)

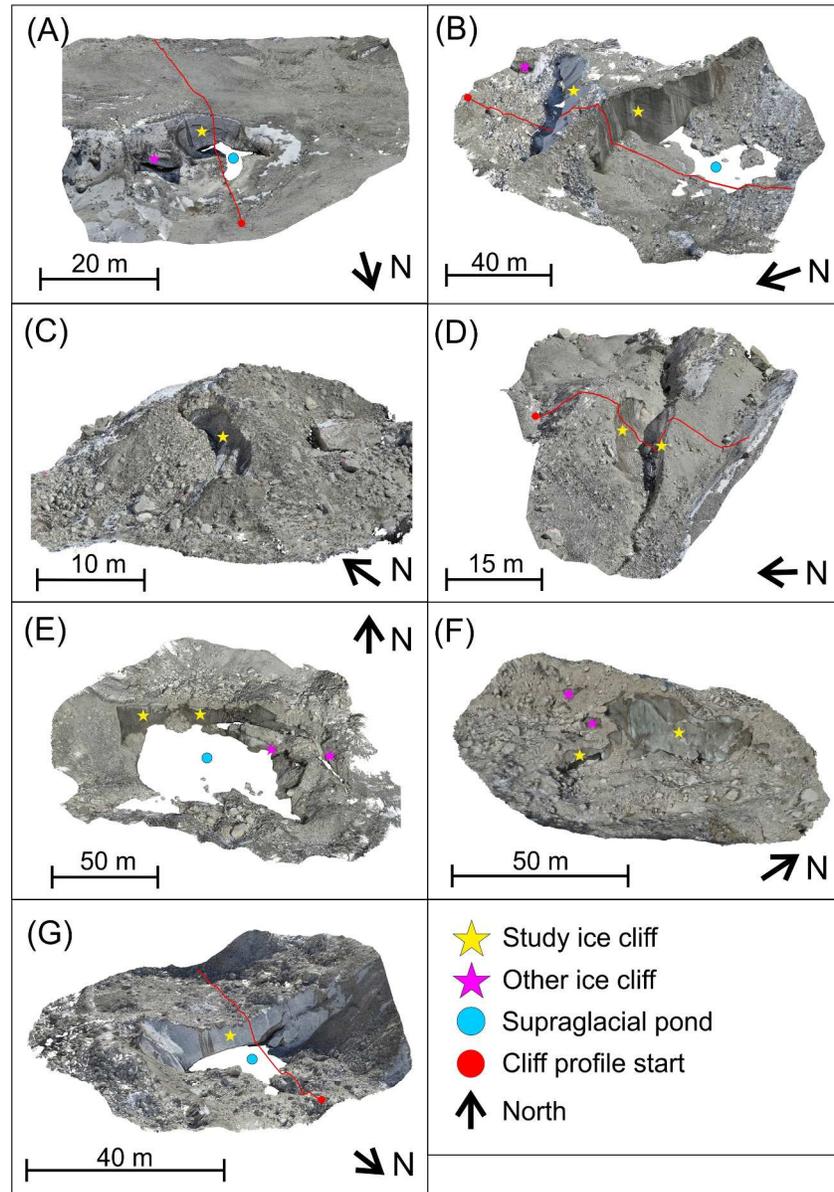


Fig. 3. Oblique views of the 3D ice cliff point clouds in November 2015. Cliff IDs correspond to Table 1 and Figure 1. The profiles (red lines) correspond to Figure 7.

223x320mm (300 x 300 DPI)

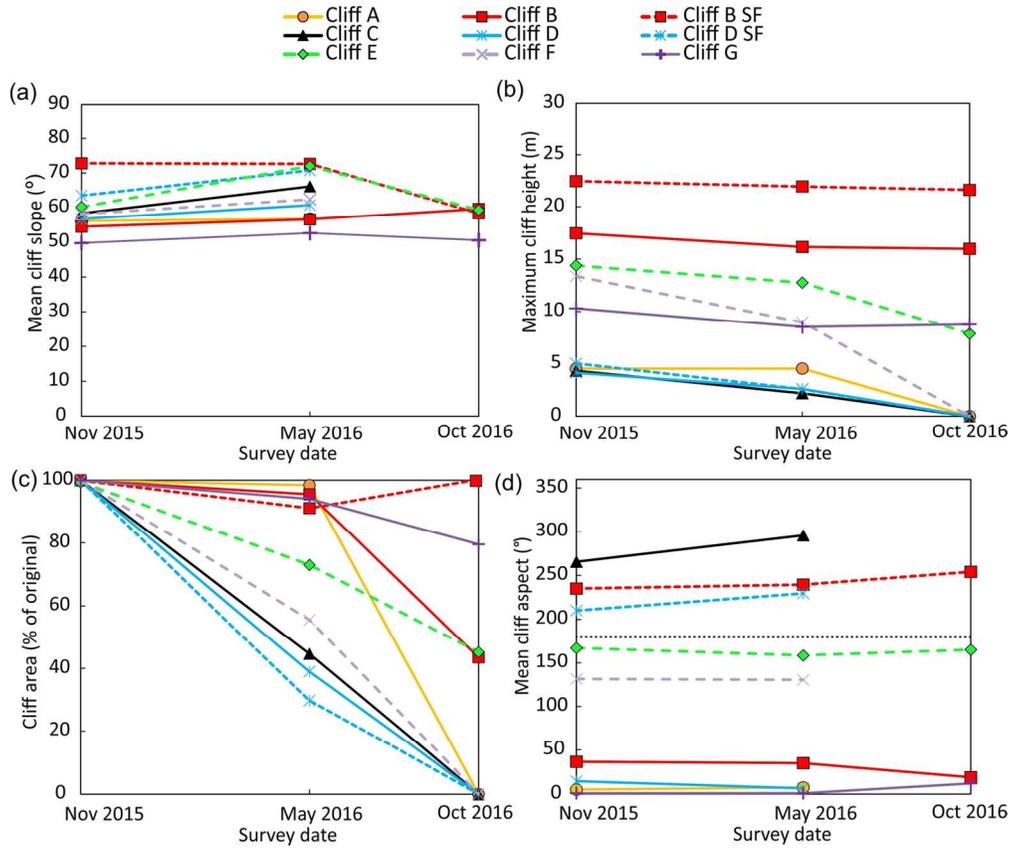


Fig. 4. The evolution of ice cliff mean slope (a), maximum height (b), area (c), and aspect (d) over the study period. Absolute cliff area change is shown in Supplementary Fig. 3.

143x120mm (300 x 300 DPI)

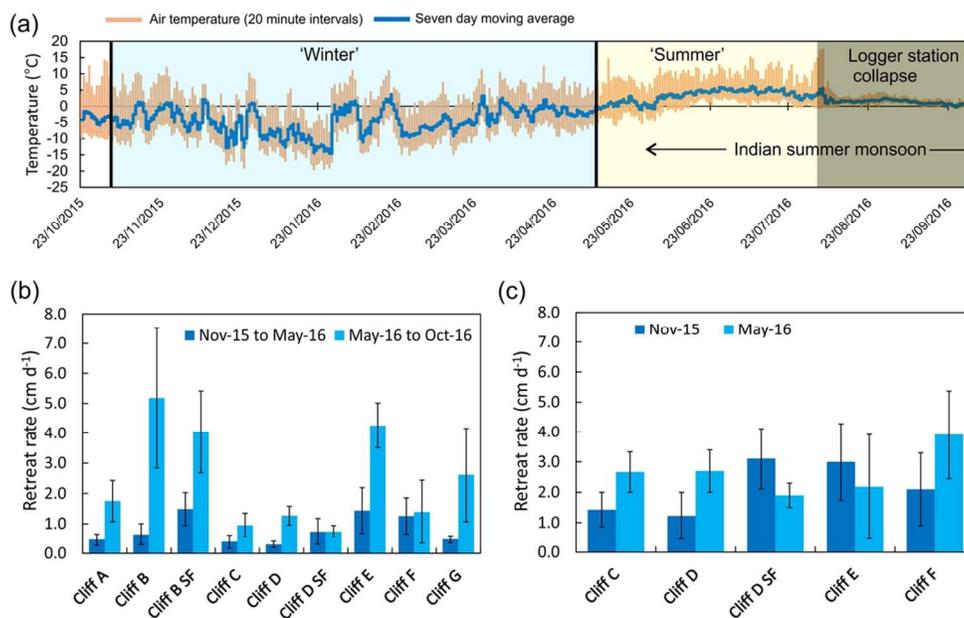


Fig. 5. (a) Air temperature at 1 m above the surface recorded at 20 minute intervals with a seven day moving average. Survey intervals are indicated by vertical black lines. The logger mounting collapsed due to ice cliff retreat in August 2016 (shaded area represents data when the logger was partially buried by debris). Mean ice cliff retreat rates for the seasonal (b), and weekly surveys (c). Error bars show one standard deviation.

101x64mm (300 x 300 DPI)

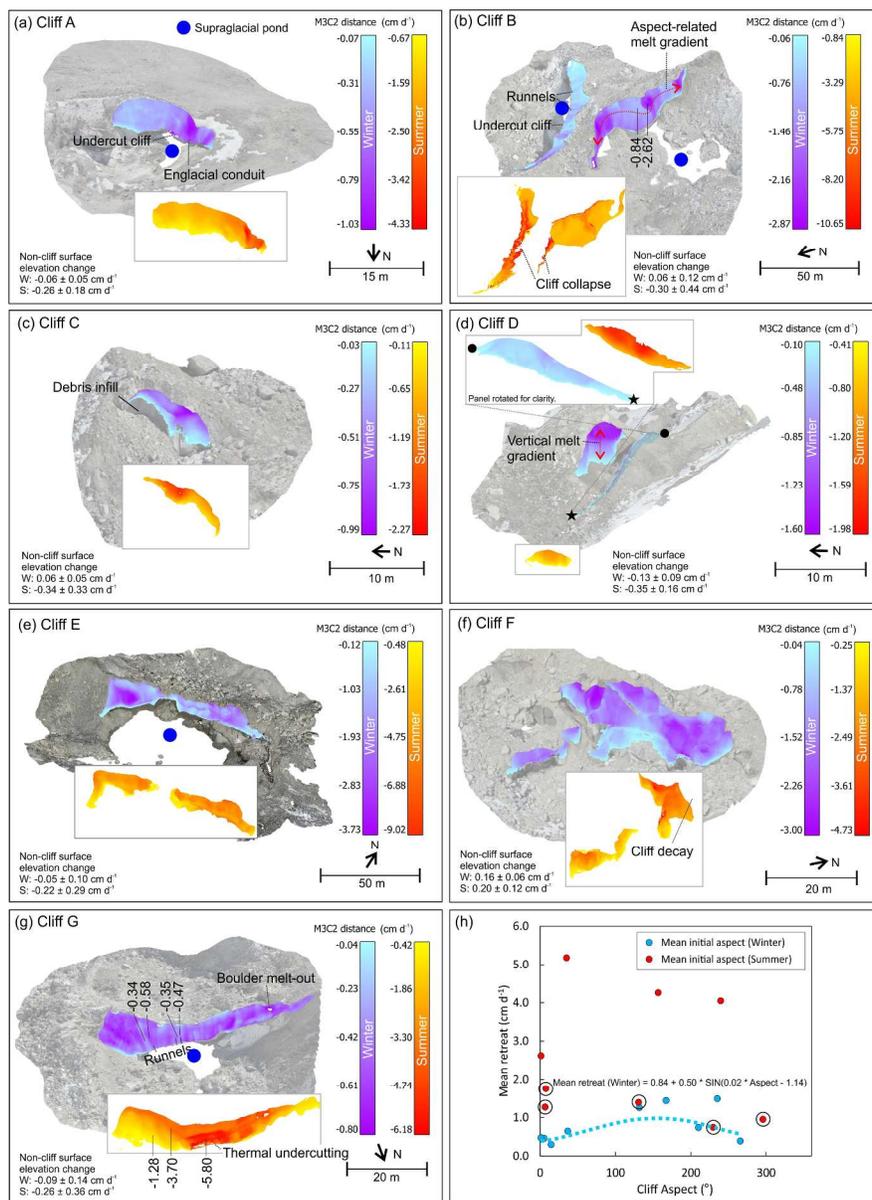


Fig. 6. Ice cliff retreat rates shown for winter (November 2015 to May 2016) and summer (May 2016 to October 2016). Note the different scale ranges. Distance measurements are clipped to the study cliffs and indicative values are shown for key features. The mean and standard deviation of non-cliff surface elevation changes are reported for winter (w) and summer (s). Ice cliff retreat rate and initial aspect for winter and summer differencing periods are shown in (h), with a sinusoidal regression line (winter). Circled points indicate ice cliffs that disappeared during summer.

252x344mm (300 x 300 DPI)

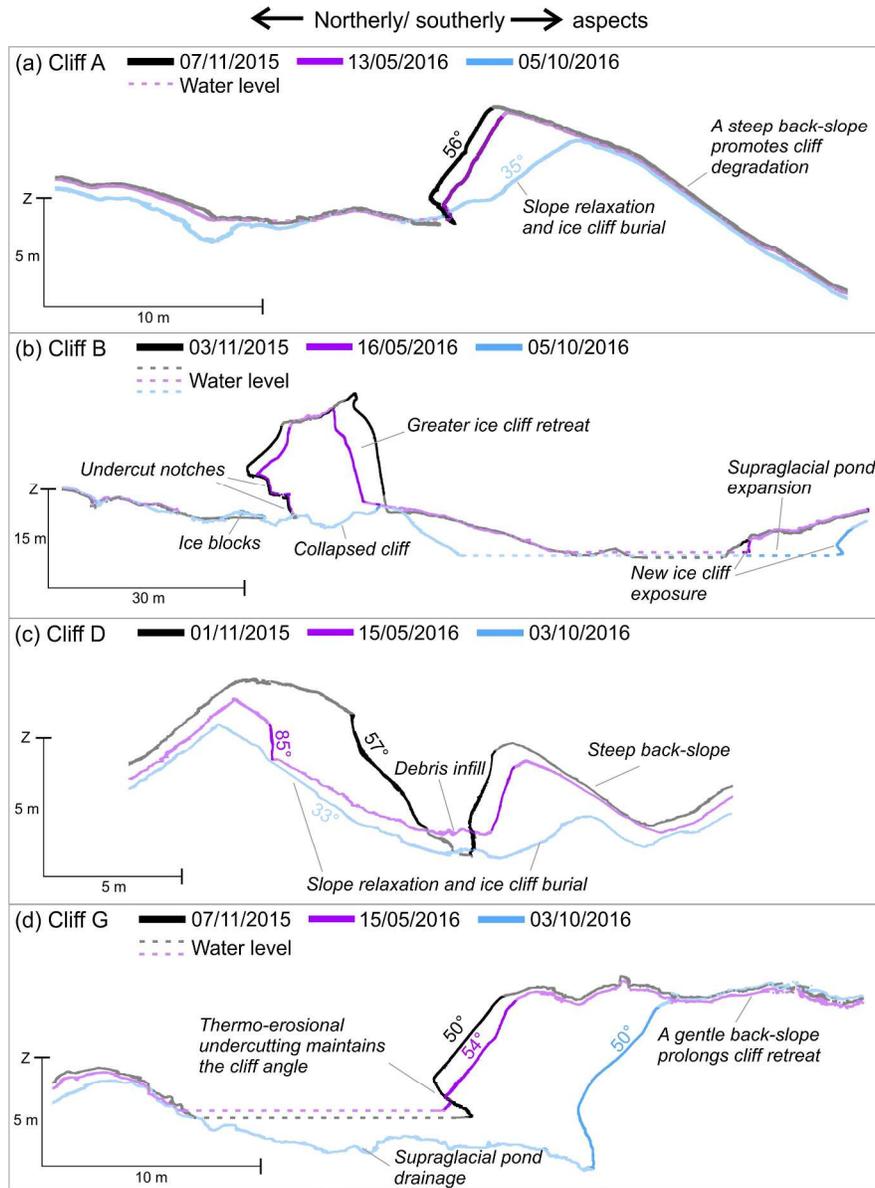


Fig. 7. 2D ice cliff profiles for selected cliffs revealing topographic change over the study period. Ice cliff faces are shown as lines without a transparency, whereas debris-covered areas and water levels are shown with transparency.

229x312mm (300 x 300 DPI)

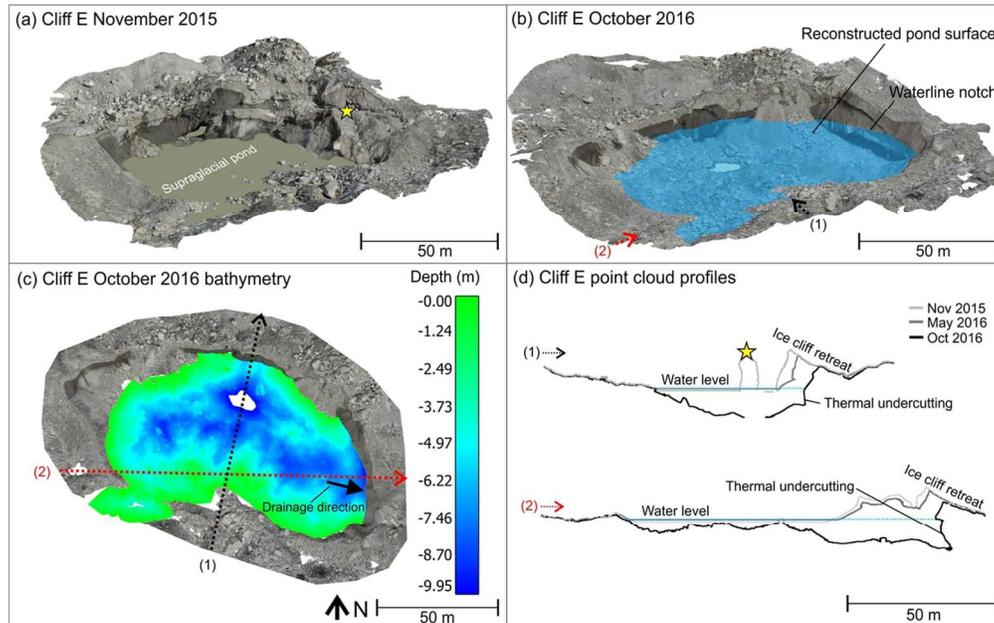


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113x71mm (300 x 300 DPI)