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## Distributed summer air temperatures across mountain glaciers: climatic sensitivity and glacier size

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**Keywords:** Air Temperature, Glaciers, Tibetan Plateau, Climatic Sensitivity

### Abstract

Near-surface air temperature ( $T_a$ ) is highly important for modelling glacier ablation, though its spatio-temporal variability over melting glaciers still remains largely unknown. We present a new dataset of distributed  $T_a$  for three glaciers of different size in the south-east Tibetan Plateau during two monsoon-dominated summer seasons. We compare on-glacier  $T_a$  to ambient  $T_a$  extrapolated from several, local off-glacier stations. We parameterise the along-flowline climatic sensitivity of  $T_a$  on these glaciers to changes in off-glacier temperatures and present the results in the context of several available distributed on-glacier datasets around the world. Climatic sensitivity decreases rapidly up to 2000-3000 m along the down-glacier flowline distance. Beyond this distance, both the  $T_a$  of the Tibetan glaciers and global glacier datasets show a slower decrease of climatic sensitivity. In general, observations on small glaciers (with < 1000 m flowline distance) are highly sensitive to temperature changes outside the glacier boundary layer. The climatology of a given region can influence the general magnitude of this climatic sensitivity, though no strong relationships are found between along-flowline climatic sensitivity and mean summer temperatures or precipitation. The terminus of some glaciers remain associated with other warm air processes that increase climatic sensitivity (such as divergent boundary layer flow, warm up-valley winds or debris heating effects) which are evident only beyond ~70% of the total glacier flowline distance. Our results therefore suggest a strong role of local effects in modulating climatic sensitivity close to the glacier terminus, although further work is still required to explain the variable presence of these effects for different glaciers.

### 1. Introduction

Near-surface air temperature ( $T_a$ ) is one of the dominant controls on glacier energy and mass balance during the ablation season (Petersen et al., 2013; Gabbi et al., 2014; Sauter and Galos, 2016; Maurer et al., 2019; Wang et al., 2019), though modelling its spatio-temporal behaviour above melting ice surfaces remains a challenge. The absence of distributed information regarding  $T_a$  has favoured the use of simple, space-time invariant relationships of  $T_a$  with elevation, typically



48 that of the free-air environmental lapse rate (*ELR*). The physical processes of the free-air (that  
49 which is independent of the surface boundary layer), however, are not appropriate to describe the  
50 variability of  $T_a$  for local glacier boundary layers (Figure 1a), especially when the above-ice  
51 temperature gradient (within ~10 m of the ice surface) heightens under warm ‘ambient’ (off-  
52 glacier) conditions (van den Broeke, 1997; Greuell and Böhm, 1998; Oerlemans, 2001;  
53 Oerlemans and Grisogono, 2002; Ayala et al., 2015). As a result, any extrapolation of  $T_a$   
54 observations from an off-glacier location, particularly those at lower elevations, are likely to lead  
55 to an overestimation of snow and ice ablation in melt simulations (Petersen and Pellicciotti, 2011;  
56 Pellicciotti et al., 2014; Shaw et al., 2017). Whilst this problem has been long understood (Greuell  
57 et al., 1997; Greuell and Böhm, 1998), only within the last decade have studies approached it in  
58 more detail (Petersen et al., 2013; Ayala et al., 2015; Carturan et al., 2015; Shaw et al., 2017;  
59 Bravo et al., 2019; Troxler et al., 2020). Until recently, modelling studies have relied upon simple  
60 lapse rates (including the *ELR*) and/or single bias offset values to account for the ‘cooling effect’  
61 of the near-surface air on-glacier (Arnold et al., 2006; Nolin et al., 2010; Ragettli et al., 2016).  
62 The variations of  $T_a$  along the glacier flowline (defined following Shea and Moore (2010) as the  
63 horizontal distance from an upslope summit or ridge), however, are much more complex (Ayala  
64 et al., 2015; Shaw et al., 2017), though a lack of available data usually restricts one’s ability to  
65 appropriately model this variable. While models applying the degree day approach can make use  
66 of off-glacier temperatures as forcing because they are heavily reliant on calibration, for  
67 physically based models and models of intermediate complexity (Pellicciotti et al., 2005; Ragettli  
68 et al., 2016) it is key to resolve the air temperature distribution over glaciers, especially for  
69 turbulent flux calculations.

70  
71 To date, two main, simplified model approaches have been developed and tested to represent air  
72 temperature over glaciers (Figure 1a). The first is the statistical model by Shea and Moore (2010)  
73 developed to reconstruct  $T_a$  across glaciers of varying size in the Canadian Rockies from ambient  
74 temperature records. This approach considered the ratio of observed on-glacier temperature and  
75 estimated ambient temperature for the elevation of a given point (hereafter ‘ $T_{aAmb}$ ’) above and  
76 below a critical threshold temperature for the onset of the glacier katabatic boundary layer (*KBL*).  
77 The parameterisations that operate as a function of the along-flowline distance have since been  
78 explored by Carturan et al. (2015) and Shaw et al. (2017) on smaller glaciers in different parts of  
79 the Italian Alps. Carturan et al. (2015) found that the original published parameterisations were  
80 sufficient to explain  $T_a$  on small, fragmenting glaciers up to distances of ~2000m. However,  
81 investigation by Shaw et al. (2017) on a small alpine glacier found a pattern of along-flowline  $T_a$   
82 that was better described by an alternative, thermodynamic model approach. This second,  
83 physically-oriented approach was developed by Ayala et al. (2015) based upon modifications of  
84 the original model by Greuell and Böhm (1998) to account for a relative ‘warming effect’ evident  
85 on the termini of some mountain glaciers compared to upper elevations that were fully dominated  
86 by katabatic winds. The modified model (termed ‘ModGB’ in the literature) accounts for the  
87 down-glacier cooling of  $T_a$  at increasing flowline distances due to sensible heat exchange and  
88 adiabatic heating (Greuell and Böhm, 1998). It adds, however, an additional warming factor based  
89 upon on-glacier observations in the lower sections of the glacier (e.g. at the greatest flowline  
90 distances) to account for additional processes of adiabatic warming (Ayala et al., 2015) (Figure  
91 1a). The ModGB approach has been successively applied at other glacier sites around the world  
92 (Shaw et al., 2017; Troxler et al., 2020), though the question of its transferability remains open  
93 (Troxler et al., 2020).

94  
95 Thus the ModGB method operates on the physical principles of the glacier boundary layer  
96 (Greuell and Böhm, 1998) though it corrects for relative warming on the lower portion of glacier  
97 (Ayala et al., 2015). To establish the magnitude of this warming, however, along-flowline data in  
98 the lower portion of the glacier are essential. Because the available distribution of on-glacier  
99 observations is often limited and rarely extends for the entire length of the glacier boundary layer,  
100 this additional correction for warming and the number of physical unknowns of ModGB can lead  
101 to high variability in  $T_a$  estimates on the glacier terminus (Troxler et al., 2020) (Figure 1a). In  
102 contrast to this, the statistical method of Shea and Moore (2010) provides a more simplified



103 estimation that has fewer assumptions and parameters, though it does not explicitly account for  
104 the physical processes on the glacier, especially those that are thought to be the cause of relative  
105 warming for the glacier terminus. It also provides a parameter that more specifically represents  
106 the glacier ‘climatic sensitivity’ of the on-glacier  $T_a$  (defined here as the ratio of changes in  
107 observed  $T_a$  on-glacier to changes in  $T_{aAmb}$ ). Despite its more conceptual nature, because of its  
108 greater generalisability typical of a more simplistic statistical approach, we adopt the Shea and  
109 Moore (2010) method to further investigate along-flowline  $T_a$  in this study.

110  
111 To the author's knowledge, no study has investigated the variability of on-glacier  $T_a$  at different  
112 sites around the world (with the exception of three glaciers considered by Ayala et al., (2015)).  
113 As such the transferability or generalisability of models and/or model parameters remain mostly  
114 unknown, and analysis of individual glacier sites, while beneficial to process understanding, may  
115 not advance the science on how to treat the on-glacier  $T_a$  in models. In this study, we make a step  
116 toward this by utilising new datasets of on-glacier temperature observations on three glaciers of  
117 varying size in the south-east Tibetan Plateau. We analyse the main controls on along-flowline  $T_a$   
118 and its climatic sensitivity and present these new findings in the context of 11 other distributed  
119 on-glacier observations around the world made to date.

120 Specifically we aim to i) understand the variability of  $T_a$  with the along-flowline distances at three  
121 glaciers in the south-east Tibetan Plateau, ii) identify and quantify the climatic sensitivity of on-  
122 glacier  $T_a$  for different meteorological conditions and glacier sizes and iii) parameterise the along-  
123 flowline  $T_a$  using the Shea and Moore (2010) method for the Tibetan glaciers and discuss it in the  
124 context of globally-derived, published datasets of on-glacier air temperatures.

125

## 126 2. Study Site

127 The study glaciers are located in the upper Parlung-Zangbo River catchment in the southeast Tibet  
128 Plateau (29.24°N, 96.93°E - Figure 2), a region characterised by a summer monsoon climate that  
129 typically intrudes via the Brahmaputra Valley (Yang et al., 2011). We present data for three  
130 maritime-type valley glaciers in the wider Parlung catchment: Parlung Glacier Number 4  
131 (hereafter ‘Parlung4’), Parlung Glacier Number 94 (‘Parlung94’) and Parlung Glacier Number  
132 390 (‘Parlung390’). Parlung4 (Figure 2d) is ~10.8 km<sup>2</sup>, north-northeast facing and has an  
133 elevation range of 4659-5939 m a.s.l. (Ding et al., 2017). Glaciers Parlung94 (Figure 2c) and  
134 Parlung390 (Figure 2e) are smaller valley glaciers (2.51 and 0.37 km<sup>2</sup>, respectively) that have  
135 termini at higher elevations (elevation ranges of 5000-5635 and 5195-5469 m a.s.l.,  
136 respectively). The glaciers of the catchment were classified by Yang et al. (2013) as having a  
137 spring-accumulation regime and the largest annual rain season of the entire Tibetan Plateau. The  
138 upper Parlung River catchment has a mean summer (1979-2019) annual air temperature of ~2°C  
139 (at 4600 m a.s.l.), and temperatures in the wider region have been shown to be increasing since  
140 the mid 1990’s (Yang et al., 2013). The glaciers of this region have been shown to be very  
141 sensitive to temperature changes, though with a more elevation-independent mass balance  
142 sensitivity compared to other, continental glaciers of the Tibetan Plateau (Wang et al., 2019). The  
143 accurate estimation of on-glacier temperatures as Tibetan glaciers shrink and fragment (Carturan  
144 et al., 2015) is thus of significant importance for continued modelling efforts. However, to date,  
145 no such studies regarding on-glacier temperature distribution have performed within the Tibetan  
146 Plateau.

147

## 148 3. Data

### 149 3.1. Air temperature observations

150 We present the observations of  $T_a$  from a total of 20 air temperature logger locations (Table 1),  
151 13 of which are situated on-glacier (4680 - 5369 m a.s.l.) and seven off-glacier (4648 - 5168 m  
152 a.s.l.). These stations (hereafter referred to as ‘T-loggers’) observed  $T_a$  at a 2 m height using  
153 HOBO U23-001 temperature-relative humidity sensors (accuracy +0.21°C) within double-  
154 lowered, naturally-ventilated radiation shields mounted on free-standing tripods. The T-loggers  
155 recorded data in 10 minute intervals that are averaged to hourly data for analysis. We identify a  
156 common observation period over the summers of 2018 and 2019 that range from 12<sup>th</sup> July – 18<sup>th</sup>



157 September. For these date ranges, we observe only small data gaps for some T-loggers (Table 1).  
158 We apply the nomenclature of  $T_{X_G}$ , whereby  $X$  refers to the T-logger number on each glacier and  
159  $G$  refers to the glacier number.

160

161 We additionally present  $T_a$  observations at two automatic weather stations (AWS) at elevations  
162  $\sim 4600$  m a.s.l. (off-glacier, henceforth ‘AWS\_Off’) and  $\sim 4650$  m a.s.l. (on Parlung4, henceforth  
163 ‘AWS\_On’) for the same time period (Figure 2). For distributing off-glacier air temperature, we  
164 consider AWS\_Off as our reference station. The AWS  $T_a$  observations are provided by Vaisala  
165 HMP60 temperature-relative humidity sensors (accuracy  $+0.5^\circ\text{C}$ ) also housed in naturally-  
166 ventilated radiation shields.

167

### 168 3.2. Uncertainty of air temperature observations

169 To provide an estimate of observation uncertainty, we compared the hourly divergence of two  
170 naturally-ventilated  $T_a$  observations for the whole period between T4<sub>4</sub> and AWS\_On (Figure 2d),  
171 that are co-located within a few metres of horizontal distance on Parlung4 Glacier. A test of  
172 absolute differences between the two stations resulted in a mean of  $< 0.4^\circ\text{C}$  for all hours ( $n =$   
173  $3312$ ) and  $\sim 0.5^\circ\text{C}$  for the warmest 10% of the hours of ambient temperature at AWS\_Off  
174 (hereafter referred to as ‘P90’ - (Ayala et al., 2015; Shaw et al., 2017; Troxler et al., 2020)). We  
175 find that for these hours (when the *KBL* development is theoretically at its strongest (e.g. van den  
176 Broeke, 1997; Oerlemans and Grisogono, 2002)), that 95% of hourly differences were  $< 1^\circ\text{C}$   
177 (Figure S1). For on-glacier stations at large flowline distances (Figure 2), these large uncertainties  
178 are considered less likely given the good ventilation provided to the sensors within the  
179 *KBL*. While observations at short flowline distances with calm conditions and high incoming  
180 radiation may result in maximum uncertainties up to  $\sim 1^\circ\text{C}$  (Troxler et al., 2020), we apply a  
181  $\pm 0.5^\circ\text{C}$  uncertainty for analysis of distributed  $T_a$ . For the instantaneous differences  $> 1^\circ\text{C}$ , wind  
182 speeds at AWS\_On were  $< 2$  m  $\text{s}^{-1}$ . Wind speeds for P90 conditions were otherwise in excess of  
183  $3\text{--}4$  m  $\text{s}^{-1}$ , though no other observations of on-glacier wind speed are available at higher elevations.  
184

184

### 185 3.3. Meteorological information

186 We obtained information regarding  $T_a$ , incoming shortwave radiation and relative humidity  
187 (AWS\_Off), on-glacier wind speed (AWS\_On) and ‘free-air’ wind speed and direction (ERA5 -  
188 C3S, 2017). We used these data to explore the relationships of hourly on- and off-glacier  
189 temperatures (section 4.2) for different prevailing conditions.

190

### 191 3.4. Elevation information

192 We used the 12.5 m Alos Palsar (ASF DAAC, 2020) digital elevation model (DEM) to provide  
193 elevation information for the catchment (Figure 2b). We utilised this DEM in order to calculate  
194 flowline distances (m) for each glacier from the TopoToolbox functions in Matlab (Schwanghart  
195 and Kuhn, 2010), following Troxler et al, (2020). We note that the methodology for flowline  
196 generation is not currently uniform among all studies of this type (Shea and Moore, 2010; Ayala  
197 et al., 2015; Carturan et al., 2015; Shaw et al., 2017; Bravo et al., 2019; Troxler et al., 2020) and  
198 may produce some differences in the calculated distances close to the lateral borders of the  
199 glaciers. In addition, the generated flowlines may also be dependent upon the quality and  
200 resolution of the DEM available between the aforementioned studies. However, we do not  
201 analyse lateral  $T_a$  variations in this study and consider that the impact of varying methods for  
202 flowline generation to be negligible when assessing observations at a few select points on the  
203 glacier.

204

## 205 4. Methods

206 For this study we use local, off-glacier  $T_a$  data from AWS\_Off for aggregation of on-glacier sub-  
207 groups or for distribution of  $T_a$  in space. Sub-grouping allows one to interpret general causal



208 factors that dictate on-glacier behaviour, whereas the distribution in space allows a direct  
209 comparison of on- and off-glacier temperatures and the effect of the glacier boundary layer. The  
210 following subsections outline the sub-grouping (4.1) and distribution (4.2) methodologies. The  
211 model parameterisations of Shea and Moore (2010) and application to Tibetan and global datasets  
212 are considered in sections 4.3 and 4.4, respectively.

#### 213 4.1. Sub-grouping on-glacier air temperature observations

214 We sub-group our on-glacier observations by 10th and 90th percentiles (P10/P90) of off-glacier  
215  $T_a$  at AWS\_Off (Figure 2a) that have been shown to relate to the development of the glacier  
216 boundary layer (Ayala et al., 2015). Following the methodology of previous studies (Ayala et al.,  
217 2015; Shaw et al., 2017; Troxler et al., 2020), we consider all contemporaneous observations of  
218 on-glacier  $T_a$  at each T-logger that relate to the same hours as the P10/P90 classification at  
219 AWS\_Off. We consider the deviation from a linear relationship of  $T_a$  with elevation and flowline  
220 distance for these subgroups, assessing this ‘linearity’ by use of the coefficient of determination  
221 ( $R^2$ ). For a comparison to previous studies (Petersen and Pellicciotti, 2011; Shaw et al., 2017), we  
222 also report the equivalent on-glacier lapse rate that would be calculated for the above conditions.  
223

#### 224 4.2. Comparison of on- and off-glacier air temperature

225 We extrapolate AWS\_Off  $T_a$  records to the elevation of each on-glacier T-logger (Table 1) to  
226 quantify the  $T_a$  differences within the glacier boundary layer (Figure 1a). We derive an hourly  
227 variable lapse rate between AWS\_Off and off-glacier T-loggers deemed to be independent of the  
228 glacier wind layer, thus excluding those T-loggers in the immediate pro-glacial zones.  
229 Specifically, we use AWS\_Off and T-loggers T1<sub>94</sub>, T2<sub>94</sub> and T1<sub>390</sub> to construct a ‘catchment lapse  
230 rate’ where the origin of the calculated regression must pass through the elevation of AWS\_Off  
231 that acts as the forcing station in this study (see supplementary information, Figure S2). We  
232 consider this as the best available approach to estimate the ambient lapse rate for the catchment.  
233 We compare the hourly estimates of extrapolated off-glacier  $T_a$  ( $T_{aAmb}$ ) with the observations at  
234 each on-glacier T-logger in order to i) understand how large the  $T_a$  offset (bias) is at each site and  
235 how it relates to meteorological conditions and glacier flowline distance; and ii) parameterise the  
236 along flowline climatic sensitivity to  $T_{aAmb}$  following Shea and Moore (2010) (section 4.3).  
237

#### 238 4.3. Estimation of on-glacier climatic sensitivity

239 The Shea and Moore (2010) approach (hereafter ‘SM10’) estimates on-glacier  $T_a$  using  $T_{aAmb}$  at  
240 a given elevation by:  
241

$$242 \quad T_a = \begin{cases} T1 + k2(T_{aAmb} - T^*), & T_{aAmb} \geq T^* \\ T1 - k1(T^* - T_{aAmb}), & T_{aAmb} < T^* \end{cases} \quad (1)$$

243 where  $T^*$  (°C) represents the threshold ambient temperature for the onset of katabatic flow and  
244  $T1$  is the corresponding threshold  $T_a$  on the glacier. Parameters  $k1$  and  $k2$  are the climatic  
245 sensitivities of on-glacier  $T_a$  to  $T_{aAmb}$  below and above the threshold  $T^*$  (Figure 1b and c).  $k1$   
246 and  $k2$  were parameterised in the original study using exponential functions of the along flowline  
247 distance ( $DF$ ):  
248  
249

$$250 \quad k1 = \beta1 \exp(\beta2 DF) \quad (2)$$

$$251 \quad k2 = \beta3 + \beta4 \exp(\beta5 DF) \quad (3)$$



258 where  $\beta_i$  are the fitted coefficients. Following the suggestion of Carturan et al. (2015), we  
259 implement a relation against the flowline that estimates the threshold temperature for onset of  
260 katabatic effects ( $T^*$ ) at a given distance as:  
261

$$262 \quad T^* = \frac{C1DF}{C2 + DF} \quad (4)$$

263  
264

265 where  $C1$  (6.61) and  $C2$  (436.04) are the fitted coefficients of Carturan et al. (2015). We calculate  
266  $k1$  and  $k2$  at each T-logger station using the linear regression of observed  $T_a$  and  $T_{aAmb}$  above  
267 and below  $T^*$  (Figure 1) as derived from equation 4. We note that the parameter  $k2$  holds a greater  
268 significance for modelling  $T_a$  (Figure 1a), as this more closely represents the climatic sensitivity  
269 reported by previous works (Greuell et al., 1997; Greuell and Böhm, 1998; Oerlemans, 2001;  
270 2010), whereas  $k1$  represents the ratio of above-glacier and free-air temperatures without a  
271 katabatic effect that have been shown to relate more closely to  $T_{aAmb}$  (Shea and Moore, 2010;  
272 Shaw et al., 2017). For this study, we therefore pay particular attention to the  $k2$  sensitivities on  
273 the Parlung glaciers and assess their relationship to along-flowline distance.  
274

#### 275 4.3. Global datasets of on-glacier temperatures

276 To explore the generalisability of the SM10 approach and provide context to the findings of the  
277 Parlung catchment, we explore the calculated  $k1$  and  $k2$  parameters for several of the available  
278 distributed on-glacier datasets published to date (Figure S3, Table 2). We subset summer periods  
279 to when all available on-glacier observations are available at a given site. For sites of the Coastal  
280 Mountains of British Columbia ('CMBC' - Shea and Moore, 2010) and Alta Val de La Mare  
281 ('AVDM' - Carturan et al., 2015), we apply the published parameter sets derived from those  
282 authors. For all other sites, we derive  $T_{aAmb}$  from the most locally available off-glacier AWS and  
283 the published lapse rate from the relevant studies (Table 2). In the absence of lapse rate  
284 information, we apply the  $ELR$  ( $-6.5^\circ\text{C km}^{-1}$ ) to extrapolate  $T_a$  to the elevation of the on-glacier  
285 observations.

286 For each glacier site, data are limited to those hours when all stations for that glacier are available  
287 and the  $k1$  and  $k2$  parameters (equation 1) are only calculated when; i),  $>10\%$  of the total hourly  
288 data at a given station is above or below  $T^*$  (to have enough data to calculate  $k2$  and  $k1$ ,  
289 respectively) and, ii) the linear regression to derive each parameter is significant to the 0.95 level.  
290 For those on-glacier stations that do not satisfy the above requirements, we do not calculate the  
291  $k1$  and  $k2$  parameters.  
292

293 Finally, we group the derived  $k2$  sensitivities of the SM10 approach against the climatology that  
294 describes the given glacier(s) location. For this, we consider the mean summer (JJAS or DJFM in  
295 the southern hemisphere) air temperature (MSAT) and the total annual precipitation for the year(s)  
296 of study at each location (Table 2). MSAT is derived from the ERA5 product for the glacier  
297 centroid location and corrected to the mean glacier elevation by the  $ELR$ . However, total  
298 precipitation from ERA5 has been shown to have considerable bias when tested against in-situ  
299 observations (e.g. Betts et al., 2019), and so we provide the best available value from the relevant  
300 literature (Table 2). We note that a full analysis of the local climate is beyond the scope of this  
301 work, though we attempted a generalised analysis in order to link any clear differences in the  
302 global datasets to climatological influences.  
303

## 304 5. Results

### 305 5.1. Variability of on-glacier air temperatures

306 Figure 3 shows the mean  $T_a$  as a function of elevation and flowline distance for the Parlung  
307 glaciers for all conditions and for the warmest 10% of AWS\_Off observations (P90). The average



308 of all hours reveals a generally linear relationship with the glacier elevation (Figure 3a) and  
309 flowline (Figure 3b), resulting in mean on-glacier lapse rate (mean  $R^2$  with elevation) equivalent  
310 to  $-3.0^{\circ}\text{C km}^{-1}$  (0.92),  $-3.7^{\circ}\text{C km}^{-1}$  (0.71) and  $-4.5^{\circ}\text{C km}^{-1}$  (0.81) for Parlung4, Parlung94 and  
311 Parlung390, respectively. For P90 hours ( $n = 312$ ), mean  $T_a$  demonstrates a poorer fit to elevation  
312 and flowline for Parlung4 (mean  $R^2$  with elevation = 0.12, and flowline = 0.20) and Parlung 94  
313 (mean  $R^2$  with elevation = 0.13 and flowline = 0.09). For the small Parlung390 Glacier,  $T_a$  remains  
314 strongly related to elevation ( $R^2 = 0.84$ ) and flowline ( $R^2 = 0.82$ ) under P90 conditions. The  
315 equivalent mean on-glacier ‘lapse rates’ for P90 hours are  $-2.1^{\circ}\text{C km}^{-1}$ ,  $-1.4^{\circ}\text{C km}^{-1}$  and  $-4.1^{\circ}\text{C}$   
316  $\text{km}^{-1}$ . Nevertheless, assuming a calculated  $0.5^{\circ}\text{C}$  uncertainty of the observations for P90  
317 conditions (Figure 3c and d), the mean of observations still lies along a linear fit line. However,  
318 for given hours, the deviation of observations from the linear fit line exceeds  $3^{\circ}\text{C}$  at large flowline  
319 distances ( $> 7000$  m) on Parlung4. In general, 2018 experienced cooler average temperatures at  
320 higher elevations, but in general, there are no marked differences between the two years of  
321 observation when comparing to glacier elevation or flowline (not shown).

322

### 323 5.2. Differences in on- and off-glacier air temperatures

324 Comparing mean on- and off-glacier  $T_a$  reveals the expected behaviour associated with the glacier  
325 ‘cooling effect’ (Carturan et al., 2015) and a greater deviation from the calculated catchment lapse  
326 rate for the warmest conditions (P90, Figure 4), indicating a reduced climatic sensitivity. The  
327 mean  $T_a$  observed at off-glacier T-Loggers supports the selection of those stations used for  
328 catchment lapse rate calculation (green dots in Figure 4) that are further from the potential effects  
329 of the glacier boundary layer (red markers in Figure 4). Following Carturan et al. (2015), we  
330 suggest a potential non-linear behaviour of lapse rates between AWS\_Off and the top of the  
331 flowline for Parlung390, though we lack the off-glacier observations above the flowline origin to  
332 test this (Figure 4b). We therefore utilise a piecewise lapse rate at the point of the highest off-  
333 glacier lapse rate station (T1<sub>390</sub>, red line in Figure 4) to account for the discrepancy between the  
334 estimated and observed  $T_a$  at T6<sub>390</sub>, which is assumed to be near to the flowline origin where  
335 climatic sensitivity is theoretically equal to 1 (i.e. that on-glacier observations =  $T_{a,amb}$ ).

336

337 Figure 5 presents the hourly differences between  $T_{a,Amb}$  and observed  $T_a$  at each site. The  
338 deviation of estimated and observed  $T_a$  theoretically begins at a critical temperature threshold,  $T^*$   
339 (Shea and Moore, 2010) and this effect can be observed at T-logger sites on Parlung94 and  
340 Parlung4, particularly those at greater flowline distances. Coloured by the hourly wind speeds  
341 recorded at AWS\_On, the beginning of the temperature deviations ( $T^*$ ) aligns well with the onset  
342 of katabatic winds on Parlung4 (and only assumed for the other glaciers due to lack of on-glacier  
343 wind observations – Figure 5). Despite being pro-glacial stations, T1<sub>4</sub> and T2<sub>4</sub> reveal a similar,  
344 albeit weaker effect of the glacier boundary layer, possibly due to larger glacier flowline and  
345 sustained effect of the katabatic wind into the pro-glacial area.

346

347 The mean bias offset of along-flowline  $T_a$  using the catchment lapse rate is shown in Figure 6.  
348 For the coolest 10% of hours at AWS\_Off (P10), there is generally minimal offset between  $T_{a,Amb}$   
349 and observed  $T_a$  for the entire dataset. This clearly does not hold true for P90 conditions (Figure  
350 6a), as already established (Figure 4), and offsets of  $T_a$  ( $T_{a,Amb} - \text{observed } T_a$ ) are up to  $5.8^{\circ}\text{C}$  at  
351 flowline distances of  $> 7000$  m on Parlung4. These effects appear to heighten beyond 2000 m  
352 along the flowline (Parlung94), though slight offsets can be witnessed for all glaciers. This is  
353 generally associated with drier conditions, and for hours of greater relative humidity (AWS\_Off),  
354 offsets are small (Figure 6b). Considering ‘free-air’ wind variability provided by ERA5  
355 reanalysis,  $T_a$  offsets are largest for the dominant south-westerly wind direction (85% of hours)  
356 and when free-air wind speeds are smallest (Figure 6c and 6d). However, un-corrected, gridded  
357 wind speeds do not appropriately represent the local ‘free-air’ boundary conditions and thus the  
358 interaction of off-glacier wind speeds and the glacier boundary layer development remain unclear  
359 for these glaciers. For all but the coolest ambient temperatures (Figure 6a), observations at the  
360 greatest flowline distances deviate the most from the estimated values.

361



362 This offset is highly variable in time, however, and related to the prevailing conditions of a given  
363 year (Figure 7). Considering the maximum daily  $T_a$  offsets at the on-glacier T-Logger closest to  
364 the terminus on each glacier (Table 1), we find that Parlung94 and Parlung4 T-loggers have  
365 similar magnitudes of  $T_a$  offsets during the mid-summer months, particularly for 2018 (Figure 7).  
366 These maximum offsets are in clear relation to the incoming shortwave radiation record at  
367 AWS\_Off (correlations of 0.44, 0.60 and 0.80 for Parlung390, Parlung94 and Parlung4,  
368 respectively), which are indicative of warmer ambient conditions (i.e. P90). For Parlung390 this  
369 offset is much smaller, though varies considerably throughout the summer. For 2019, maximum  
370 daily  $T_a$  offsets on Parlung390 steadily increase during July and August then fall close to zero in  
371 September. The bias offsets for Parlung4 and Parlung94, however, remain sizeable (Figure 7).  
372 Because our study period focuses on the core monsoon period (Yang et al., 2011), we do not  
373 observe the influence of monsoon arrival or cessation on the  $T_a$  variability of the Parlung Glaciers.

### 374 375 5.3. Parameterisation of along-flowline air temperatures

376 Figure 8 presents the SM10 parameterisations for the Parlung glaciers in comparison to those  
377 derived for the available distributed  $T_a$  datasets around the world (Table 2). Comparing the  $k1$  and  
378  $k2$  parameters from Tibet to the original parameters of Shea and Moore (2010), a similar  
379 behaviour is observable for both sites up to ~2000-3000 m of flowline distance (red and blue  
380 dashed lines), though there exists a larger variability in the calculated parameters at longer  
381 flowline distances on Parlung4 (Figure 8). Accordingly, the exponential functions that are fitted  
382 to the observations at Parlung glaciers and the original study are notably distinct (Figure 8, Table  
383 3). This behaviour is further highlighted when observing other published or revised datasets for  
384 the context of this work (Figure 8b). A 'global' parameterisation for all sites where down-glacier  
385 decrease in climatic sensitivity is evident (black dashed lines in Figure 8) clearly misrepresents  
386 many of the observations, particularly those at greater flowline distances, balancing the  
387 behaviours reported for different sites.

388 Notably, observations at McCall Glacier, Alaska relate very well to ambient  $T_a$  under cooler  
389 conditions, with all  $k1$  values remaining  $> 0.9$ . Above the  $T^*$  threshold, however, the relationship  
390 of observed and estimated  $T_a$  results in increasing  $k2$  along the flowline, in contradiction to the  
391 majority of the other datasets. Nevertheless, this data also confirms the increased climatic  
392 sensitivity on the glacier terminus (Troxler et al., 2020) as evident with datasets for Tsanteleina  
393 (Shaw et al., 2017), Arolla and Juncal Norte (Ayala et al., 2015). Observations at Parlung4 and  
394 Universidad Glacier (Bravo et al., 2017) emphasise the strong decrease in climatic sensitivity at  
395 large flowline distances (~10,000 m) previously only witnessed from one location on Bridge  
396 Glacier, Canada (Shea and Moore, 2010).

397  
398 Figure 9 shows the  $k2$  parameters plotted against flowline distance, coloured by rankings of  
399 MSAT and precipitation totals (Table 2). The warmest of the investigation sites (during the  
400 measurement years) appear to lie closer along the original SM10 parameterisation until ~4000 m,  
401 whereas deviation of the  $k2$  parameters from this line appears larger for observations at relatively  
402 cold sites (Greve, McCall and Arolla – Figure 9a). The main exception to this is for Juncal Norte  
403 (Petersen and Pellicciotti, 2011; Ayala et al., 2015), which demonstrates a high and rapidly  
404 increasing sensitivity of ambient  $T_a$  at the greatest flowline distances.

405 No clear patterns are visible with relation to mean annual precipitation totals, although the  
406 observations at Juncal Norte are noted as the driest of the study sites (Figure 9b).

407 A clear difference between observations of CMBC and Parlung at large flowline distances is the  
408 distance from the glacier terminus, which suggests a possible difference in processes being  
409 compared between sites. Accordingly, we plot the  $k2$  parameters as a function of the normalised  
410 flowline, adjusted by the total length of glacier under the year(s) of observation (Table 2). A slight  
411 trend toward lower  $k2$  values (lower sensitivity to ambient  $T_a$ ) is observable for relatively warmer  
412 regions, though no clear pattern emerges for MSAT (Figure 9c) or precipitation totals (Figure 9d).  
413 The largest flowline distance observation of the entire dataset (Figure 9a) in fact extends only  
414 ~60% of the total glacier length (Bridge Glacier - CMBC), neither representing the smallest  
415 climatic sensitivity (Figure 8b), nor the increasing climatic sensitivity witnessed at the terminus  
416 of the glacier by other studies (Ayala et al., 2015). In fact, by subjectively grouping glacier sites





417 by the presence of the relative down-glacier cooling (decreasing sensitivity) and warming  
418 (decreasing followed by increasing sensitivity) along the total glacier length, one can observe that  
419 this effect is absent for the both smaller and larger glaciers (Figure 10a), albeit limited by lack of  
420 observations across an entire glacier in most cases. For those glaciers where the pattern of down-  
421 glacier sensitivity increase ('relative warming effect') is evident, it is found only beyond ~70%  
422 of the total glacier flowline distance (Figure 10b – vertical dashed line). Up to this distance, no  
423 increase in  $k_2$  sensitivity is seen from the data.  
424

## 425 6. Discussion

### 426 6.1. Relevance of the findings from Parlung Glaciers

427 The observations of along-flowline  $T_a$  on the glaciers in the Parlung catchment add yet more  
428 evidence of the spatial variability of the glacier cooling and dampening effect (Oerlemans, 2001;  
429 Carturan et al., 2015; Shaw et al., 2017) and the need to appropriately estimate its behaviour for  
430 use in glacier melt models (Petersen and Pellicciotti, 2011; Shaw et al., 2017; Bravo et al., 2019).  
431 It has long since been observed that a static lapse rate is inappropriate for characterising the spatio-  
432 temporal variability of  $T_a$ , both within the *KBL* (Greuell et al., 1997; Konya et al., 2007; Marshall  
433 et al., 2007; Gardner et al., 2009; Petersen and Pellicciotti, 2011) and outside the glacier boundary  
434 layer in adjacent valleys (Minder et al., 2010; Immerzeel et al., 2014; Gabbi et al., 2014; Heynen  
435 et al., 2016; Jobst et al., 2016). Despite this, the lack of locally available observations often  
436 requires modellers to force model simulations with the nearest off-glacier record of  $T_a$  and  
437 extrapolate it based upon the *ELR* value as a default. In the case of Tibetan glaciers, model studies  
438 have often derived static lapse rates between on-and off-glacier stations (Huintjes et al., 2015) or  
439 used static values to extrapolate or downscale  $T_a$  with a correction based upon a single on-glacier  
440 location (e.g. Caidong and Sorteberg, 2010; Yang et al., 2013; Zhao et al., 2014). To the author's  
441 knowledge, this is the first time that such detailed information regarding spatio-temporal  
442 variations in  $T_a$  have been presented for a glacier of the Tibetan Plateau. Because glaciers of the  
443 south-eastern Tibetan Plateau have been shown to be particularly susceptible to increases in  $T_a$   
444 (Wang et al., 2019), accurately parameterising  $T_a$  along glaciers of differing size is highly relevant  
445 for present and future melt modelling attempts. This is especially true where glaciers begin to  
446 shrink or fragment (Munro and Marosz-Wantuch, 2009; Jiskoot and Mueller, 2012; Carturan  
447 et al., 2015) and become more sensitive to ambient air temperatures due to a lack of katabatic  
448 boundary layer development (Figures 6 and 7).  
449

450 The summer monsoon exerts a strong control on the energy and mass balance of Tibetan glaciers  
451 (Yang et al., 2011; Mölg et al., 2012; Zhu et al., 2015). Although our dataset spanned two  
452 summers of only the core monsoon period for this region (Yang et al., 2011), we have shown that  
453 the sensitivity of the glacier to external temperature changes (shown by  $T_a$  bias offsets) has a  
454 sizeable temporal variability that can be controlled by the monsoon weather conditions (such as  
455 ambient air temperature, humidity and incoming radiation) and can sometimes be independent of  
456 the glacier size (Figure 7). Whilst we cannot determine the impact of monsoon timing and  
457 intensity upon the climatic sensitivity of these glaciers with the current dataset, we are able to  
458 determine that the observed relationship to flowline distance is consistent to that of other regions  
459 of the world (Figure 8). Future work on Tibetan glaciers should attempt to extend monitoring to  
460 the pre-monsoon period to identify if a seasonal onset for the changing glacier climatic sensitivity  
461 can be defined, and how the monsoon may affect it. Particular focus should be given to understand  
462 the local meteorological conditions for each glacier, as this may explain some of the variability  
463 in  $T_a$  offset values, and why they may sometimes be independent of the along-flowline distance  
464 (Figure 7).  
465

### 466 6.2. Parameterising glacier climatic sensitivity

467 In this study, we discuss the climatic sensitivity of on-glacier  $T_a$  based upon observations above  
468 a threshold ambient temperature for the onset of katabatic conditions ( $T^*$ ). This sensitivity to  
469 ambient temperature during relatively warm conditions, indicated by the  $k_2$  parameter of Shea  
470 and Moore (2010)(Figure 1), demonstrates a generally consistent behaviour between the T-logger



471 observations of Parlung glaciers and those where this model had been previously implemented  
472 (Shea and Moore, 2010; Carturan et al., 2015). It also resulted in a similar parameterisation, albeit  
473 with a slightly greater sensitivity to the ambient temperature (i.e. larger  $k_2$  values - Figure 8b).  
474 Whilst the newly presented dataset for the Parlung catchment provides an important confirmation  
475 of the climatic sensitivity for some Tibetan glaciers, further studies of individual glaciers can  
476 provide only local parameterisations for climatic sensitivity that may not be applicable to other  
477 sites. Accordingly, we have made here a first attempt at combining many of the published datasets  
478 regarding distributed  $T_a$  on mountain glaciers around the world (Table 2) to examine the potential  
479 for generalisability of a model accounting for climatic sensitivity (Figure 8).

480

481 We found a sizeable spread in the climatic sensitivities of  $T_a$  for the on-glacier datasets considered,  
482 though a consistently rapid decrease of sensitivity along glacier flowlines is found for most sites  
483 up until ~2000-3000 m of distance (Figure 8b). While localised meteorological and topographic  
484 factors likely interact to explain the spread of sensitivities at small flowline distances (Figure 8b),  
485 the results suggest that small glaciers with flow lengths < 1000 m would reflect a 0.7-0.8  
486 sensitivity to changes in  $T_{aAmb}$ . Beyond this distance, the climatic sensitivities notably follow  
487 one of two patterns; a continued, albeit less rapid decrease in sensitivity (more generally following  
488 the model proposed by Shea and Moore (2010)), or a tendency toward increasing sensitivity at  
489 the largest flowline distances (more related to those findings of the 'ModGB' model - Figure 1a).  
490 With reference to the relative  $T_a$  differences among only on-glacier observations, these have been  
491 termed as down-glacier 'cooling' or 'warming', respectively for many past studies (Ayala et al.,  
492 2015; Carturan et al., 2015; Shaw et al., 2017; Troxler et al., 2020). Whilst the former is generally  
493 associated with relatively warmer regions of study (Figure 9), such as the Canadian Rockies (Shea  
494 and Moore, 2010) or Universidad Glacier (Bravo et al., 2017), no strong relationship of the  
495 climate setting exists between these sites to explain the magnitude of the climatic sensitivity (i.e.  
496 the strength of the glacier cooling and dampening effect) nor the observed increases in climatic  
497 sensitivity on glacier termini (Ayala et al., 2015; Shaw et al., 2017; Troxler et al., 2020).

498

499 Interestingly, we noted that the most distant station observation used to derive the  
500 parameterisation by Shea and Moore (2010) was located only around 60% of the total glacier  
501 flowline distance (Bridge Glacier - Figure 10), whereas data presented by other studies, provided  
502 observations up to the glacier terminus (Greuell and Böhm, 1998; Ayala et al., 2015; Shaw et al.,  
503 2017; Troxler et al., 2020), therefore potentially parameterising different effects of the glacier  
504 boundary layer. It has been suggested that observations at large flowline distances (such as that  
505 on Parlung4 or Bridge Glacier) represent a segment of the boundary layer where the near-surface  
506 layer becomes highly insensitive to the ambient free-air temperature fluctuations (point '3' in  
507 Figure 1a and d). This phenomenon has been shown to be sustained over large fetch distances by  
508 an increasing depth of the glacier wind layer (van den Broeke et al., 1997; Greuell and Böhm,  
509 1998; Shea and Moore, 2010, Jiskoot and Mueller, 2012). However, as air parcels travel down-  
510 glacier toward the glacier terminus (point '4' in Figure 1a and d), they potentially encounter warm  
511 air entrainment due to a divergent boundary layer (Munro, 2006), up-valley winds (Pellicciotti et  
512 al., 2008; Oerlemans, 2010; Petersen and Pellicciotti, 2011) or heating from debris-covered ice at  
513 the terminus (Brock et al., 2010; Shaw et al., 2016; Steiner and Pellicciotti, 2016; Bonekamp et  
514 al., 2020). These are effects of the glacier boundary layer that the ModGB model was designed  
515 to account for, though we did not explicitly test this within our study due to a requirement for  
516 more data and a greater number of parameters and assumptions (Shaw et al., 2017). The strength  
517 of this so called along-glacier 'warming effect' could therefore be governed by local topography  
518 (adjusting the boundary layer convergence or divergence) or the total glacier flowline distance  
519 and the large fetch of a cool air parcel overcoming the competing effect of warm, up-valley winds  
520 (Figure 1d - as seen at T24 in Figure 5).

521

522 By subjectively grouping glaciers by the presence of the observed increase in climatic sensitivity  
523 and normalising the flowline distance of the observations by the total flowline for each glacier,  
524 we identify that the relative increases in climatic sensitivity begin at ~ 70% of the total flowline  
525 distance (Figure 10). A smaller climatic sensitivity can be observed for larger glaciers (Figure



526 10a), which is consistent with the development of the *KBL* over a large fetch (Greuell and Böhm,  
527 1998; Shea and Moore, 2010), though the length itself indicates nothing clear about why greater  
528 climatic sensitivity exists for some glacier termini (Figure 10b).

529  
530 The clear outlier of these datasets is Juncal Norte Glacier in Chile (Figure 8b). It is interesting to  
531 note that Juncal Norte is the only reported case in the literature on  $T_a$  variability where the warmest  
532 hours of the afternoon correspond to the dominance of an up-valley, off-glacier wind (Pellicciotti  
533 et al., 2008; Petersen and Pellicciotti, 2011). Counter to the typical role of the dominant, down-  
534 glacier wind layer for these warmest afternoon hours (Greuell et al., 1997; Greuell and Böhm,  
535 1998; Strasser et al., 2004; Jiskoot and Mueller, 2012; Shaw et al., 2017; Troxler et al., 2020), up-  
536 valley winds on Juncal Norte seemingly erode the along-flowline reduction in climatic sensitivity  
537 (along-flowline cooling) up to a distance along the flowline where it is theoretically at its  
538 maximum (point ‘3’ in Figure 1). Evidence from other glaciers suggest that this point is close to  
539 upper observations for Juncal Norte at ~70% of the total flowline (Figure 10b), though further  
540 observations would be required to test this.

### 541 6.3. Future directions for researching air temperatures on glaciers

542 A limitation of our work is the dependency of the derived ‘global’ climatic sensitivities (Figure  
543 8b) to the available off-glacier data and the published lapse rates to extrapolate them to the  
544 relevant elevations on-glacier. In our case, we are able to identify a potentially non-linear lapse  
545 rate of  $T_a^{Amb}$  for the highest elevations over Parlung94 and Parlung390 (Figure 4). Although we  
546 cannot confirm this without off-glacier observations above the top of the flowline (Carturan et al.,  
547 2015), we are able to well constrain ambient air temperature distribution using hourly  
548 observations at several off-glacier locations to derive the best possible ‘catchment lapse rate’. For  
549 other datasets (Table 2), we rely upon the available off-glacier data and lapse rates that are not  
550 derived in a consistent manner. The derivation of flowline distances from the DEM are also not  
551 consistent between the prior studies (Shea and Moore, 2010; Carturan et al., 2015; Shaw et al.,  
552 2017; Bravo et al., 2019; Troxler et al., 2020), and may hold some small influence on the derived  
553 parameterisations (Table 3), particularly at lateral locations on the glacier (not explored here),  
554 that can be subject to different micro-meteorological effects (van de Wal, 1992; Hannah et al.,  
555 2000; Shaw et al., 2017). Equally, the uncertainty of the actual observations (e.g. section 3.2) is  
556 hard to clearly define due the variable instrumentation (sensors and radiation shielding), on-  
557 glacier location and local topographic and micro-meteorological effects of each study site (Table  
558 2). Further work on a unified model of estimating  $T_a$  should need to address these issues, perhaps  
559 with further, dedicated analyses.

560  
561 In our study, we apply the parameterisation of Carturan et al. (2015) to derive along-flowline  
562 values of the theoretical onset of the *KBL* ( $T^*$ ). While these values appear appropriate for our case  
563 studies (based upon manual inspection), they were derived for a more limited number of total  
564 observations. We experimented with a static  $T^*$  value of 5°C in order to test the sensitivity of our  
565 analysis to the assumptions of  $T^*$ , though found a minimal change in our derived  $k2$  sensitivities  
566 (not shown).

567  
568 Finally, in this study we assess climatic sensitivity based upon ambient air temperatures above  
569 this  $T^*$  threshold. We identify, however, that this is partly different to the climatic sensitivity  
570 presented by earlier works (Greuell et al., 1997; Greuell and Böhm, 1998; Oerlemans, 2001;  
571 2010), which considered all hours of the on-glacier observations when comparing to extrapolated  
572 off-glacier  $T_a$ . In some instances, over estimation of on-glacier  $T_a$  also for cooler conditions may  
573 produce a consistent ‘all-hour’ climatic sensitivity value (i.e. where  $k1$  and  $k2$  sensitivities are  
574 similar - Figure 1b). However, ignoring separate effects ( $k1$  and  $k2$ ) due the rise of the *KBL*  
575 (Figure 1c, Figure 5) arguably over-simplifies the glacier’s climatic sensitivity and therefore does  
576 not aptly describe the two behaviours separated by an onset event (Shea and Moore, 2010; Jiskoot  
577 and Mueller, 2012). Accordingly, we caution somewhat the direct comparison of the climatic  
578 sensitivity presented here and that of previous works, though consider the use of  $k2$  to be an  
579 appropriate indicator of climatic sensitivity for this work going forward. As previously  
580



581 mentioned, we have considered the approach of Shea and Moore (2010) to be a more  
582 generalizable method for calculating glacier climatic sensitivity and thus estimating on-glacier  $T_a$ .  
583 However, the competing effects of glacier katabatic and up-valley winds need to be incorporated  
584 to address the challenges that less simplistic methods (i.e. ModGB) were designed for.

585  
586 Based upon the findings of this work, we recommend that future research i) attempt to standardise,  
587 where possible, the measurement and comparison of off- and on-glacier air temperature,  
588 potentially exploring more the use of artificially-ventilated radiation shields that are less prone to  
589 heating errors (Georges and Kaser, 2002), ii) instrument glaciers of varying size in the same  
590 catchment to explore the relative importance of glacier size and local meteorological conditions  
591 (Figure 7), and iii) model the detailed interactions of air flows on the glacier termini using, for  
592 example, large eddy simulations (Sauter and Galos, 2016; Bonekamp et al., 2020) in order to  
593 identify possible drivers of the observed increase in climatic sensitivity for certain glaciers (point  
594 '4' in Figure 1).

## 595 7. Conclusions

596 We presented a new dataset of distributed on-glacier air temperatures for three glaciers of  
597 different size in the south-east Tibetan Plateau during two summers (July - September). We  
598 analysed the along-flowline air temperature distribution for all three glaciers and compared them  
599 to the estimated ambient temperatures derived from several, local off-glacier stations. Using this  
600 information, we parameterised the along-flowline climatic sensitivities of these glaciers using the  
601 method proposed by Shea and Moore (2010) and presented the results in the context of several  
602 available distributed on-glacier datasets to date. The key findings of this work are:

- 604
- 605 1. For our Tibetan case study, on-glacier air temperatures at short flowline distances are  
606 more climate sensitive (i.e. demonstrate a relationship with off-glacier air temperature  
607 that is closer to 1). We therefore confirm earlier evidence regarding the high sensitivity  
608 of small glaciers (flowline distances < 1000 m) to external climate, and thus future  
609 warming.
  - 610 2. The largest offsets between observed on-glacier and estimated off-glacier air  
611 temperatures are found for the warmest off-glacier hours, during drier, clear sky  
612 conditions of the summer monsoon period.
  - 613 3. Above the established onset of the katabatic boundary layer, climatic sensitivity to  
614 ambient temperature decreases rapidly up to ~2000-3000 m along the glacier flowline.  
615 Beyond this distance, both the Tibetan glaciers and other datasets of the literature show  
616 a slower decrease of climatic sensitivity.
  - 617 4. A parameterisation for the climatic sensitivity of the Tibetan study glaciers implies a  
618 smaller boundary layer effect than the existing parameterisation of Shea and Moore  
619 (2010). The climatology of a given region may influence the magnitude of the glacier's  
620 climatic sensitivity, though no clear relationships with the climatology of the glacier sites  
621 are found, thus suggesting the stronger role of local meteorological or topographic effects  
622 on the along-flowline pattern of  $T_a$  variability.
  - 623 5. The terminus of some glaciers remain associated with other warm air processes, possibly  
624 due to boundary layer divergence, warm up-valley winds or debris cover heating. We find  
625 that these effects are evident only beyond ~70% of the total glacier flowline distance,  
626 although further work is required to explain this behaviour. A better understanding of  
627 temperature variability for this lower 30% is highly important as most of the summer  
628 melting will occur for this sector of the glacier.

629  
630 In summarising the findings from all available distributed on-glacier datasets to date, we identify  
631 some key directions for future work on this subject. This includes comparing local influences of  
632 glacier size and micro-meteorology and standardising measurement practices, where possible, to  
633 aid the conclusions for a generalised model of on-glacier air temperature estimation.

634



635 **Acknowledgements**

636 Funding for the instrumentation of the Parlung catchment was provided by NSFC project  
637 (91647205 and 4191101270) and Newton Advanced Fellowship (NA170325). Á. Ayala  
638 acknowledges support from a FONDECYT project (number 3190732) and C. Bravo from the  
639 ANID Becas Chile PhD scholarship program. F. Pellicciotti acknowledges an ERC Consolidator  
640 Grant: ‘RAVEN’ (Rapid mass loss of debris covered glaciers in High Mountain Asia, grant  
641 agreement no. 772751). The authors kindly acknowledge the sharing of global datasets or  
642 parameters provided to aid this analysis, explicitly M. Nolan (McCall Glacier), J. Pomeroy, D.  
643 Pradhananga and the Global Water Futures Programme (Peyto Glacier), P. Smeets and IMAU,  
644 Utrecht (Pasterze Glacier), DGA, Chile (Universidad Glacier and Greve Glacier) and L. Carturan  
645 (AVDM, Italy). E. Ludewig is thanked for the provision of off-glacier temperature records at  
646 Sonnblick station, Austria. Additionally we recognise the hard work involved in obtaining and  
647 sharing all of the datasets acquired and the acknowledgements of those works.

648 **Author contributions**

649 TES and WY discussed and designed the research plan with Parlung data provided by WY and  
650 CZ. Additional data and analysis was provided by AA and CB. TES wrote the manuscript with  
651 scientific input from all co-authors.

652 **Data availability**

653 Calculated flowlines and climatic sensitivities are available at the following Zenodo repository:  
654 <http://doi.org/10.5281/zenodo.3937777>

655 **Competing Interests**

656 The authors declare that they have no conflicting interests.

657

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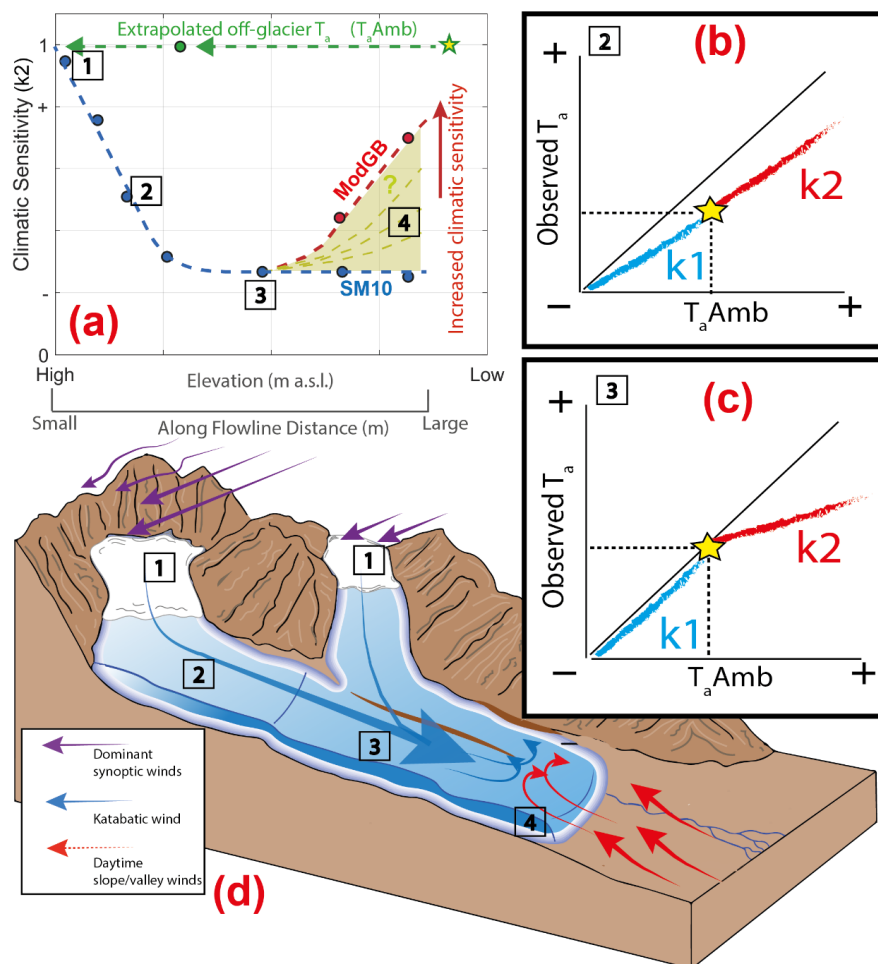
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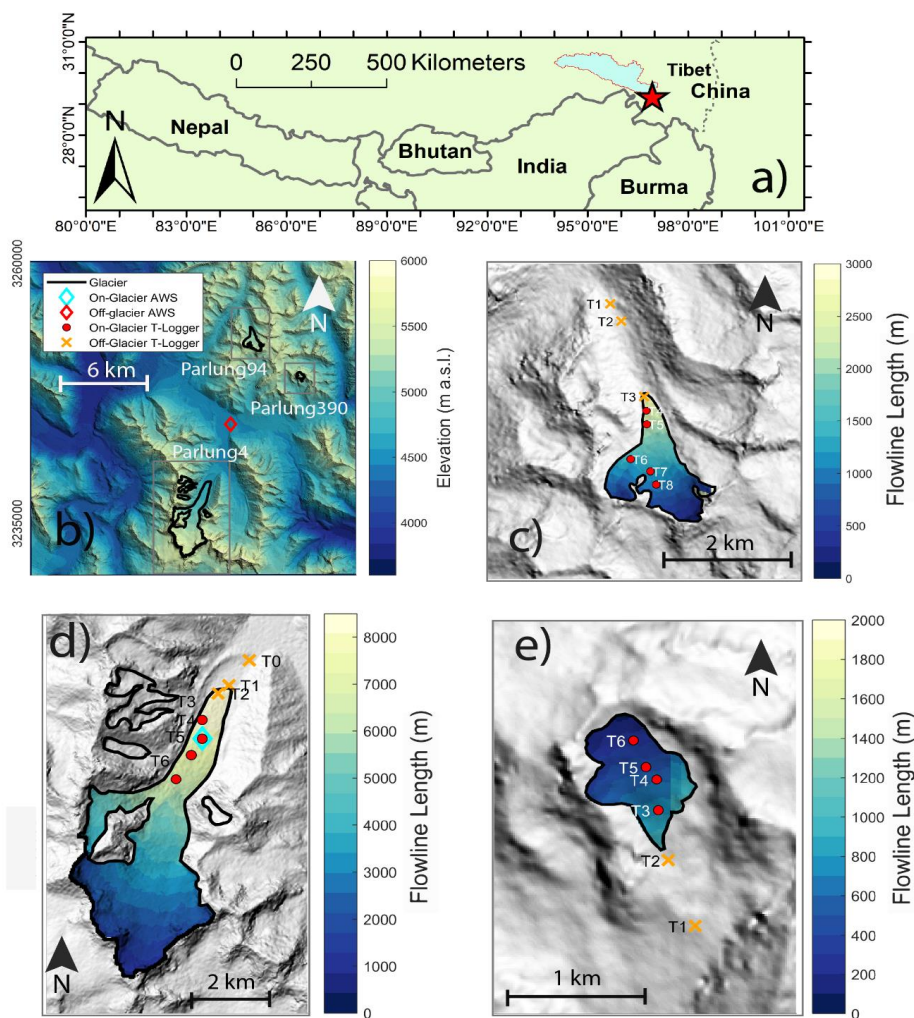
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**Figures**



844  
 845 *Figure 1: A schematic diagram to describe the climatic sensitivity of on-glacier air temperature ( $T_a$ ) to the*  
 846 *extrapolated ambient temperature ( $T_a^{Amb}$ ) at given elevations/flowline distances on a mountain glacier.*  
 847 *Points 1-4 indicate locations of interest in the context of the current science for this topic that are linked*  
 848 *between panels. Panel (a) indicates the along-flowline 'k2' climatic sensitivities to  $T_a^{Amb}$ , considering*  
 849 *down-glacier decrease in sensitivities and the observed differences in the models of SM10 and ModGB for*  
 850 *glacier termini (see text). While ModGB does not explicitly include the k2 parameter, its approach is similar*  
 851 *to considering an increasing climatic sensitivity to  $T_a^{Amb}$  (see Ayala et al., 2015). The green line in panel*  
 852 *(a) indicates the local off-glacier lapse rate to estimate  $T_a^{Amb}$  using off-glacier observations at varying*  
 853 *elevations (green dot). Panels (b) and (c) represent the differences of k1 and k2 sensitivities observed in*  
 854 *the data at different theoretical locations on the glacier (see Figure 5, for examples on Parlung glaciers),*  
 855 *the latter of which shows the theoretical parameterisation presented by Shea and Moore (2010). Panel (d)*  
 856 *represents an idealised case of katabatic and valley/synoptic wind interactions that potentially dictate the*  
 857 *along-flowline structure of on-glacier climatic sensitivity and thus  $T_a$  estimation.*

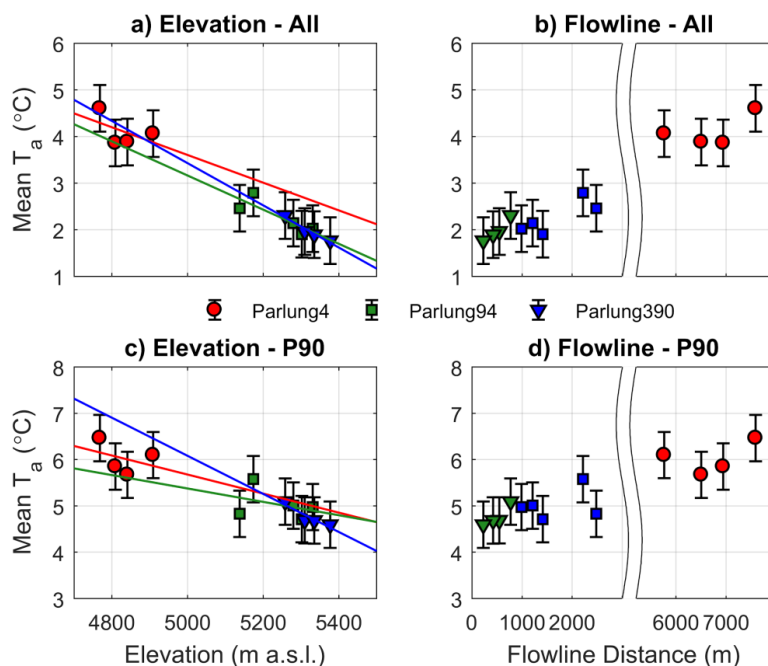


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859 *Figure 2: The location of Parlung catchment in Tibet (a) and a map of the Parlung catchment (b) with the*  
860 *study glaciers, Parlung 94 (c), Parlung4 (d) and Parlung390 (e). Off-glacier and on-glacier AWS and T-*  
861 *Logger locations are shown (without glacier number suffix). (a) shows the elevation of the catchment (DEM*  
862 *source: Alos Palsar) and (b-d) show the calculated flowline distances based upon TopoToolbox (scales*  
863 *vary).*

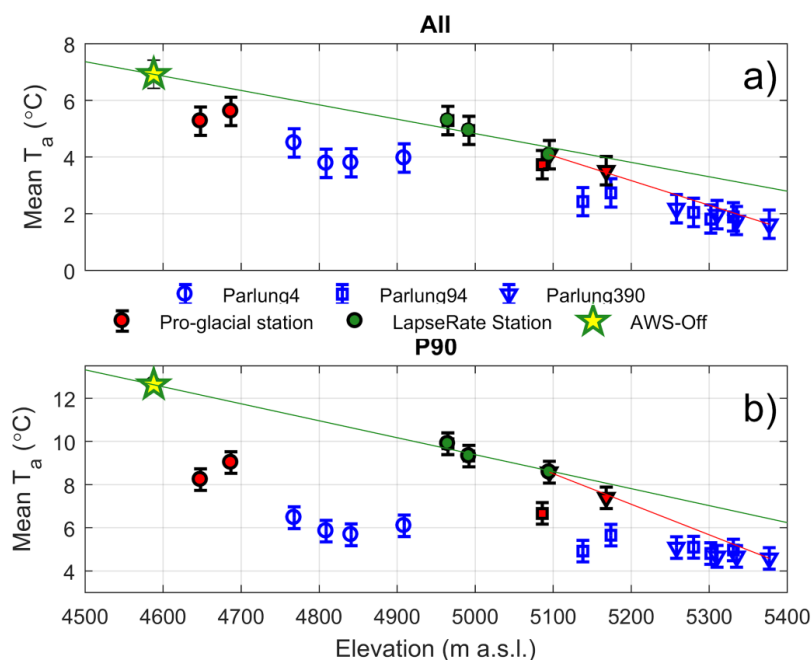
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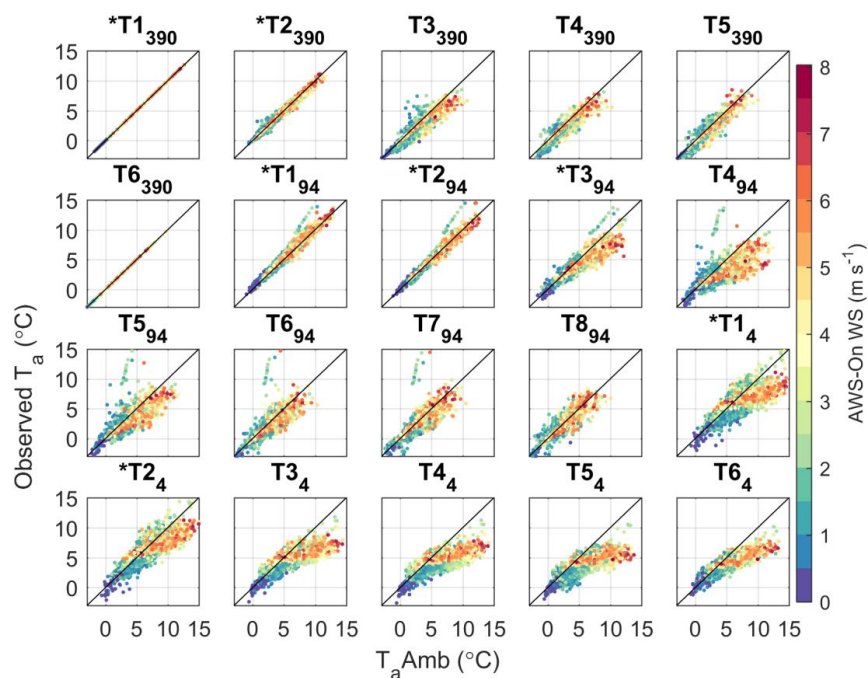
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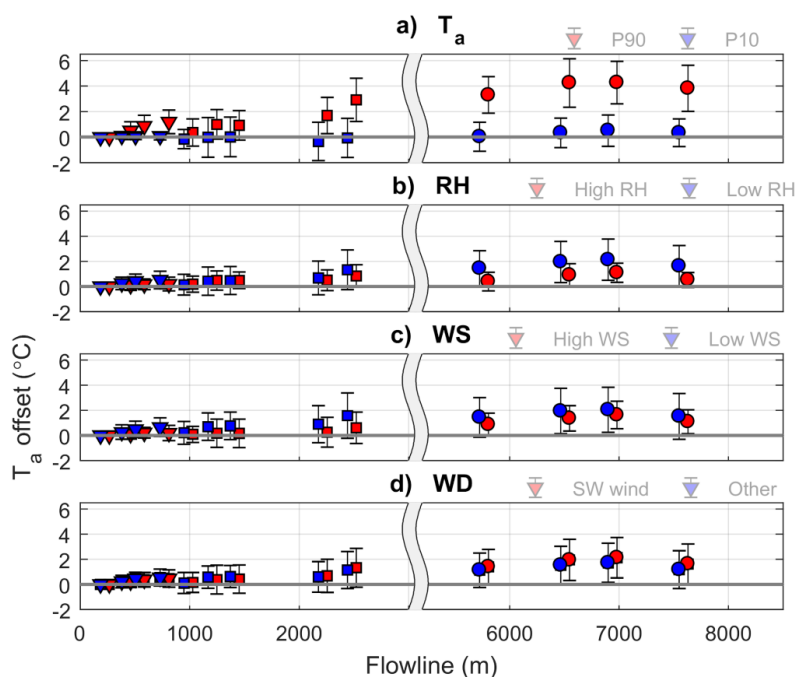
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868 *Figure 3: The elevation-mean  $T_a$  and uncertainty (errorbar) for (a) all hours and (c) P90 hours ( $n = 312$ ).*  
869 *Panels (b) and (d) are the equivalent plots against flowline distance. Coloured lines show the linear fit*  
870 *against elevation ('lapse rate') to each glacier. An x-axis scale break is used in (b) and (d) for clarity.*  
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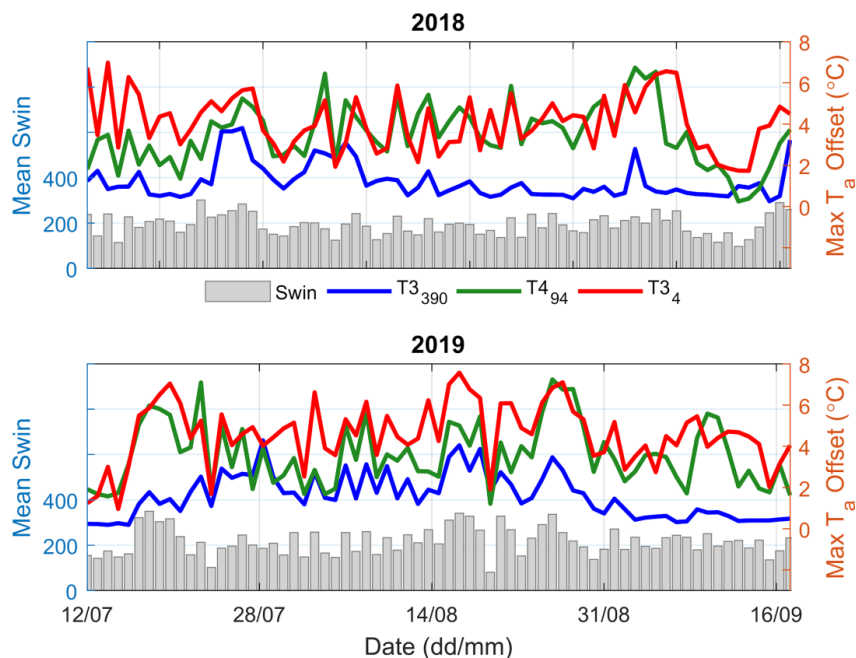
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 874 Figure 4: The mean  $T_a$  against elevation for all hours (a) and P90 hours (b), where blue markers are on-  
 875 glacier T-Loggers, red markers are pro-glacial T-Loggers and green circles denote off-glacier T-Loggers  
 876 used to construct an hourly variable 'catchment lapse rate' (green line), extrapolated from AWS\_Off  
 877 (star). The red line indicates the piecewise lapse rate above the elevation of T1\_390 to lapse  $T_a$  to the top  
 878 of the flowline. A 0.5°C uncertainty is shown by the errorbar for each station (not applied to the lapse  
 879 rate for neatness).



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 881 *Figure 5: Estimated ( $T_{aAmb}$ ) vs Observed  $T_a$  at each T-Logger location (including off-glacier T-*  
 882 *Loggers). Individual, hourly values are coloured by the observed wind speeds at AWS\_On (Parlung4). No*  
 883 *on-glacier wind speed data exist for Parlung94 and Parlung390, so the coloured markers are only*  
 884 *assumed for those glaciers from the parlung4 wind speed data. \* denotes stations that are off-glacier.*  
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 891 *Figure 6: The mean and standard deviation (error-bars) of hourly  $T_a$  bias offsets (estimated-observed)*  
 892 *along the glacier flowline. Each panel depicts hourly grouping by (a) off-glacier  $T_a$  at AWS\_Off (P90 is  $\geq$*   
 893 *10.5 °C and P10 is  $\leq 3.5$  °C), (b) off-glacier RH at AWS\_Off (high is  $> 90$  % and low is  $< 70$  %), (c)*  
 894 *wind speed from ERA5 (high =  $> 2.5$  m  $s^{-1}$  and low =  $< 0.7$  m  $s^{-1}$ ) and (d) dominant wind direction from*  
 895 *ERA5 (Southwest wind direction is considered as 180-270°). Marker shapes show the different glaciers,*  
 896 *as in Figure 3 and 4.*  
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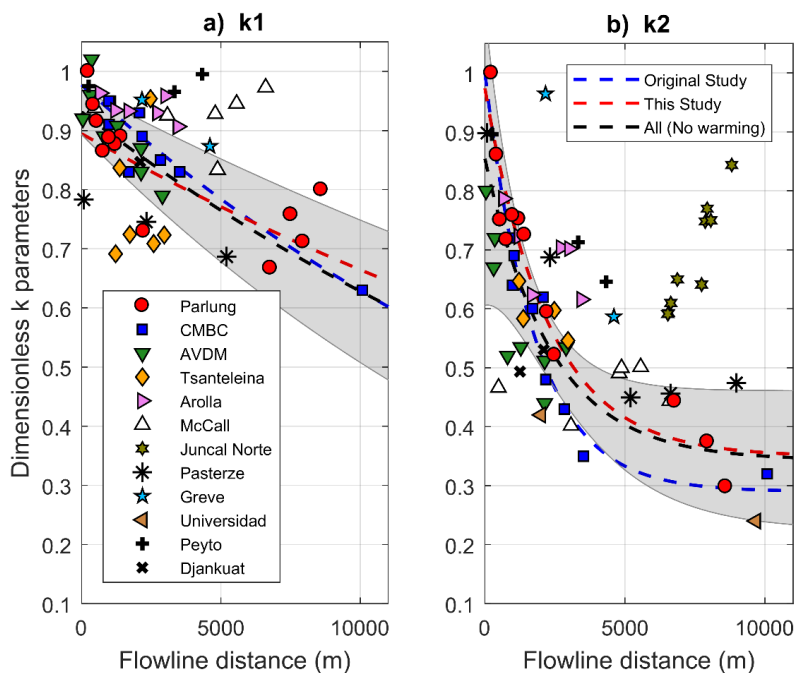


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900 *Figure 7: Maximum daily  $T_a$  offsets (estimated - observed) at the most distant along-flowline T-Loggers on*  
901 *each glacier for (top) and 2019 (bottom). Mean daily incoming shortwave radiation at AWS\_Off is shown*  
902 *by the grey bars.*

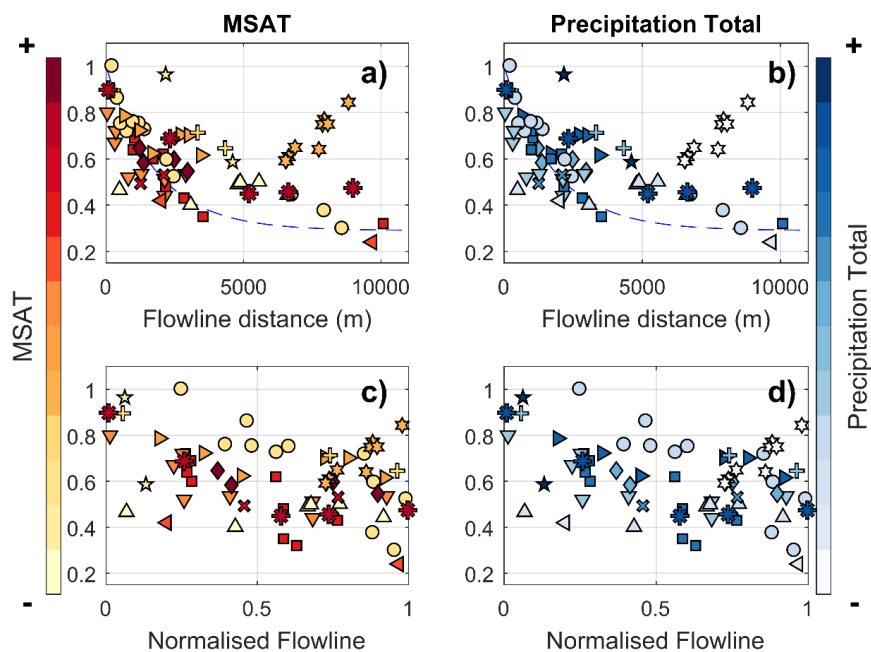
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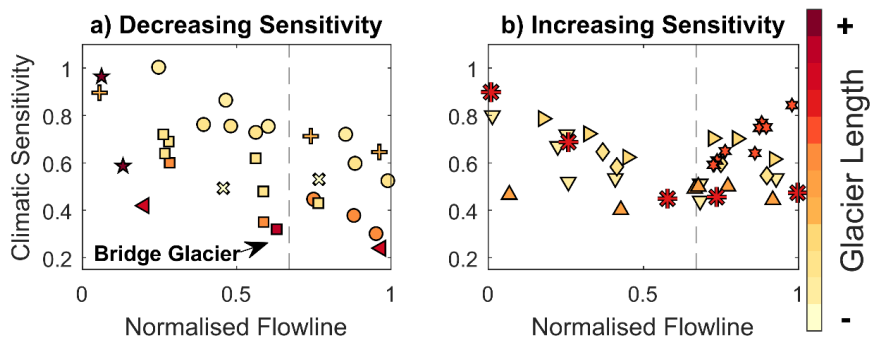
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905 *Figure 8: The calculated  $k_1$  and  $k_2$  sensitivities as a function of the flowline distance of each observation*  
 906 *on the Parlung glaciers (red circles) and other, global datasets (Table 2). The dashed blue and red lines*  
 907 *show the fitted exponential parameterisation of Shea and Moore (2010) and this study, respectively. The*  
 908 *dashed black line and shaded area denotes the equivalent parameterisation for all observations where a*  
 909 *large increase in sensitivity on the glacier terminus ('warming effect') is absent (explicitly excluding data*  
 910 *from McCall, Juncal Norte and Djankuat). The shaded area represents the 95% confidence interval of this*  
 911 *fit line.*



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913 *Figure 9: The  $k_2$  sensitivities as a function of flowline distance (top) and a normalized distance, considering*  
914 *the total flowline distance for the year of study (bottom). The individual glaciers of grouped studies*  
915 *(Parlung, CMBC and AVDM) are separated and normalized by the individual glacier length (symbols as*  
916 *in Figure 8). Glaciers are coloured by rankings of the mean summer air temperatures (MSAT - left) and*  
917 *precipitation total (PT - right). The original parameterisation is retained in the top panels.*

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933 *Figure 10: The  $k_2$  sensitivity along the normalized flowline compared to total glacier length (colour bar)*  
934 *and subjectively grouped by the evidence of a relative warming effect (increasing climatic sensitivity)*  
935 *toward the glacier terminus.*

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965 **Tables**  
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967 *Table 1: Details of each AWS/T-Logger station used in this analysis including the calculated flowline*  
 968 *distances.*

Station	Latitude	Longitude	Elevation (m a.s.l.)	Flowline (m)	on/off glacier
AWS_Off	29.314	96.955	4588	-	off
AWS_On	29.500	97.009	4649	-	off
T1 <sub>390</sub>	29.348	97.022	5095	-	off
T2 <sub>390</sub>	29.352	97.020	5168	-	off
T3 <sub>390</sub>	29.354	97.0202	5258	770	on
T4 <sub>390</sub>	29.356	97.020	5310	544	on
T5 <sub>390</sub>	29.357	97.019	5335	420	on
T6 <sub>390</sub>	29.359	97.018	5377	224	on
T1 <sub>94</sub>	29.621	97.218	4965	-	off
T2 <sub>94</sub>	29.417	96.99	4992	-	off
T3 <sub>94</sub>	29.635	96.975	5086	-	off
T4 <sub>94</sub>	29.596	97.065	5138	2481	on
T5 <sub>94</sub>	29.56	97.067	5174	2215	on
T6 <sub>94</sub>	29.466	97.023	5302	1411	on
T7 <sub>94</sub>	29.434	97.080	5280	1208	on
T8 <sub>94</sub>	29.399	97.097	5331	988	on
T1 <sub>4</sub>	29.271	96.968	4690	-	off
T2 <sub>4</sub>	29.368	96.935	4769	-	off
T3 <sub>4</sub>	29.298	97.168	4809	8589	on
T4 <sub>4</sub>	29.298	97.168	4809	7940	on
T5 <sub>4</sub>	29.496	97.126	4841	7505	on
T6 <sub>4</sub>	29.403	97.068	4909	6765	on

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Table 2: The details of each site where distributed on-glacier air temperatures are available. Elevation ranges and mean summer air temperatures (MSAT) are reported for the year of investigation. Precipitation totals (mm – ‘PT’) was obtained upon cited literature.

Site	Lat	Lon	Year(s)	Elevation	MSA	PT	$T_a$ Data Reference
				m .a.s.l.	T °C	mm	
Parlung (Tibet)	29.24	96.93	2018-2019	4600-5800	2.19	679	This Study
CMBC (Canada)	50.32	-122.48	2006-2008	1375-2898	10.29	1113	Shea and Moore (2010)
AVDM (Italy)	46.42	10.62	2010-2011	2650-3769	7.94	784	Carturan et al. (2015)
Tsanteleina (Italy)	45.48	7.06	2015	2800-3445	13.76	805	Shaw et al., (2017)
Arolla (Switzerland)	45.97	7.52	2010	2550-3520	7.28	1663	Ayala et al. (2015)
McCall (USA)	69.31	-143.85	2004-2014	1375-2365	-2.28	500	Troxler et al. (2020)
Juncal Norte (Chile)	-33.01	-70.09	2007-2008	2900-5910	6.58	352	Ayala et al. (2015)
Greve (Chile)	-48.88	-73.52	2015-2016	0-2400	-0.1	12000	Bravo et al. (2019)
Pasterze (Austria)	47.09	12.71	1994	2150-3465	12.66	2761	Greuell and Böhm, (1998)
Universidad (Chile)	-34.69	-70.33	2009-2010	2463-4543	8.24	474	Bravo et al. (2017)
Peyto (Canada)	51.66	-116.55	2011	2260-3000	2.94	800	Pradhananga et al. (2020)*
Djankuat (Russia)	43.20	42.77	2017	3210-4000	12.13	950	Rets et al. (2019)

983 \*paper not yet submitted

984 Table 3: The coefficients of the original SM10 model and those fit to the  $k1$  and  $k2$  sensitivities on the  
 985 Parlung glaciers and all glaciers where no warming effect was evident (see Figure 10).

Model	$k1 = \beta1 * \exp(\beta2 * DF)$	$k2 = \beta3 + \beta4 * \exp(-\beta5 * DF)$
CMBC (Shea and Moore, 2010)	$\beta1 = 0.977$ $\beta2 = -4.4e-5$	$\beta3 = 0.29$ $\beta4 = 0.71$ $\beta5 = 5.6e-4$
Parlung	$\beta1 = 0.894$ (0.805,0.983) $\beta2 = -2.972e-5$ (-5.543e-5,-4.0e-6)	$\beta3 = 0.349$ (0.241,0.456) $\beta4 = 0.624$ (0.492,0.757) $\beta5 = 4.4e-3$ (1.7e-4,7.2e-4)
All (no increased sensitivity on glacier terminus)	$\beta1 = 0.923$ (0.886,0.96) $\beta2 = -3.375e-5$ (-5.543e-5,-4.0e-6)	$\beta3 = 0.343$ (0.225,0.46) $\beta4 = 0.511$ (0.38,0.642) $\beta5 = 4.2e-3$ (1.5e-4,6.9e-4)

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