

1 **Debris cover effects on energy and mass balance of Batura Glacier in the Karakoram over the**
2 **past 20 years**

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15 Abstract:

16 The influence of supraglacial debris cover on glacier mass balance in the Karakoram is noteworthy. However,
17 understanding of how debris cover affects the seasonal and long-term variations in glacier mass balance through
18 alterations in the glacier's energy budget is incomplete. The present study coupled an energy-mass balance model
19 with heat conduction within debris layers on debris-covered Batura Glacier in Hunza valley, to demonstrate the
20 influence of debris cover on glacial surface energy and mass exchanges during 2000-2020. The mass balance of
21 Batura Glacier is estimated to be -0.262 ± 0.561 m w.e. yr⁻¹, with debris cover accounting for a 45% reduction in
22 the negative mass balance. Due to the presence of debris cover, a significant portion of incoming energy is utilized
23 for heating debris, leading to a large energy emission to atmosphere via thermal radiation and turbulent sensible
24 heat. This, in turn, reduces the melt latent heat energy at the glacier surface. We found that the mass balance exhibits
25 a pronounced arch-shaped structure along the elevation gradient, which is associated with the distribution of debris
26 thickness and the increasing impact of debris cover on the energy budget with decreasing elevation. Through a
27 comprehensive analysis of the energy transfer within each debris layer, we have demonstrated that the primary
28 impact of debris cover lies in its ability to modify the energy flux reaching the surface of the glacier. Thicker debris
29 cover results in a smaller temperature gradient within debris layers, consequently reducing energy reaching the
30 debris-ice interface. Over the past two decades, Batura Glacier exhibited a trend toward less negative mass balance,
31 likely linked to a decrease in air temperature and reduced ablation in areas with thin or sparse debris cover.

32

33 1 Introduction

34 Karakoram Glaciers have maintained a relative stable status under atmospheric warming, compared with other
35 High Mountain Asia (HMA) glaciers over past 30 years (Zemp et al., 2019; Nie et al., 2021; Gardelle et al., 2012),
36 a phenomenon which has been referred to as the “Karakoram Anomaly” (Hewitt, 2005). However, due to the
37 influence of topographical and supraglacial features, the rate of glacier change across this region exhibits a distinct
38 spatial heterogeneity. Notably, supraglacial debris plays a key role in mass change on many glaciers in the
39 Karakoram. Over the past three decades, a discernible expansion of supraglacial debris has been observed
40 throughout the Karakoram region (Xie et al., 2023), achieving a notable coverage of 21% in areas such as the Hunza
41 river basin (Xie et al., 2020). Ever since Hewitt (2005) identified the inhibitory effect of supraglacial debris on melt,
42 particularly below 3500m, as a possible explanation for the "Karakoram Anomaly", mapping the changes in the
43 extent and mass changes of debris-covered glaciers has been the focus of several recent studies (e.g., Mölg et al.
44 (2018), Azam et al. (2018), Xie et al. (2020)).

45 Until now, the direct assessment of debris impact on Karakoram glaciers has been limited to a few glaciological
46 measurements conducted over short periods. Mihalcea et al. (2008) modeled debris-covered ice ablation across the
47 ablation area of the Baltoro glacier, employing a distributed approach that calculated conductive heat flux through
48 the debris layer. However, their study lacked a thorough analysis of the debris effect on ice melt. Recently, Huo et
49 al. (2021b) conducted advanced research on the Baltoro glacier, presenting a model that comprehensively
50 characterizes ablation dynamics, considering temporally-linked radiative forcing, surface geomorphological
51 evolution, and gravitational debris flux. They emphasized the role of system couplings and feedbacks between
52 surface morphology, melt, and debris transport, revealing an overall increase in ablation due to high-frequency
53 topographic variations leading to a larger area with thin debris cover. At a larger scale, such as the Central
54 Karakoram, Minora et al. (2015) reported a noticeable difference in melt rates between debris-covered and debris-
55 free ice, utilizing an enhanced temperature index model. Furthermore, by conducting a comparative modelling study
56 of ice melt with and without debris cover for one ablation season in 2004, Collier et al. (2015) estimated that debris
57 cover reduced ablation by approximately 14% in the Karakoram. They attributed this significant reduction to
58 insulation by thick debris cover exceeding increases in melt under thin debris. Additionally, Groos et al. (2017)
59 confirmed that debris influences the anomalous behavior of glaciers in the Karakoram using a surface mass balance
60 model. They emphasized that debris is not the sole driver, however; factors such as favorable meteorological
61 conditions and the timing of the main precipitation season also contribute. Consequently, the distribution of debris

62 holds strong potential for affecting atmosphere–glacier feedbacks and glacier ablation in this region, warranting
63 more comprehensive exploration of the intricate dynamics of mass and heat exchange within the debris in the
64 Karakoram.

65 Supraglacial debris up to a few centimeters thickness generally increases melt due to lowered albedo and
66 increased heat absorption at the surface (Collier et al., 2014), while thicker debris cover typically suppresses the
67 melt rate through insulation (Østrem, 1959; Nicholson and Benn, 2006; Bisset et al., 2020). These contrasting effects
68 have been demonstrated by many recent studies (Gardelle et al., 2012; Nuimura et al., 2017; Basnett et al., 2013;
69 Fujita and Sakai, 2014). The reduction of ablation associated with increasing debris thickness down glacier can lead
70 to an inverted mass-balance elevation profile on the debris-covered ablation zone, which has profound implications
71 on the evolution of a glacier under a warming climate (Banerjee, 2017). Some field studies have also identified
72 diverse effects on melt rates of debris cover with different thickness in Karakoram; one particular finding showed
73 that thin debris cover, e.g. 0.5 cm in thickness, does not accelerate ice melting in this region (Muhammad et al.,
74 2020). However, some remote sensing based research proposed that while thick debris typically inhibits the melt
75 rate, the overall ablation on glaciers extensively covered in debris is still significant (Kääb et al., 2012). These
76 findings imply that understanding of the process and feedback mechanisms governing ablation of debris-covered
77 glaciers in this region is still incomplete. Therefore, it is important to quantify not only the amplitude of melt under
78 time-variable debris cover but also its role in the “Karakoram Anomaly” by assessing the thermal properties of
79 debris layers of different thickness.

80 Field glaciological and meteorological observations on glaciers in the Karakoram are limited by logistical and
81 political constraints (Mayer et al., 2014; Mihalcea et al., 2008). Consequently, a significant knowledge gap exists
82 for debris thickness and its thermal properties as well as the complex coupling of meteorology with heat exchange
83 over glaciers and in debris layers. A limited number of previous melt process investigations under debris layers,
84 e.g., Juen et al. (2014), Evatt et al. (2015), Muhammad et al. (2020), supported by remote sensing observations and
85 climate reanalysis data, have enabled physically-based numerical modeling to provide insight into thermal dynamics
86 within supraglacial debris. For example, Huo et al. (2021b) provided new insights into the relationships between
87 ablation dynamics, surface morphology and debris transport, while Collier et al. (2015) developed understanding of
88 how debris cover affects the atmosphere–glacier feedback processes during the melt season. However, despite these
89 advancements, certain aspects remain insufficiently addressed. Specifically, the seasonal variations and long-term
90 changes in melt patterns, along with the manner in which debris cover exerts its influence on such variations, have

91 not been comprehensively studied. Understanding these dynamics is essential not only for establishing the physical
92 basis of the “Karakoram Anomaly” but also for quantifying the extent to which debris cover contributes to this
93 phenomenon. In this study, we applied an energy-mass balance model coupled with heat conduction within debris
94 layers on Batura Glacier in Hunza valley, Karakoram to demonstrate the influence of debris cover on glacial melt.
95 We aim to: (1) reconstruct the long-term mass balance history of the Batura Glacier, a representative debris-covered
96 glacier in the region; and (2) numerically estimate the distributed ice melt rate under the spatially-heterogeneous
97 supraglacial debris of the Batura Glacier. By enhancing our understanding of glacier mass balance behavior and its
98 relationship to debris cover energy budgets in the Karakoram over the last two decades, this research adds
99 significantly to existing knowledge in this field.

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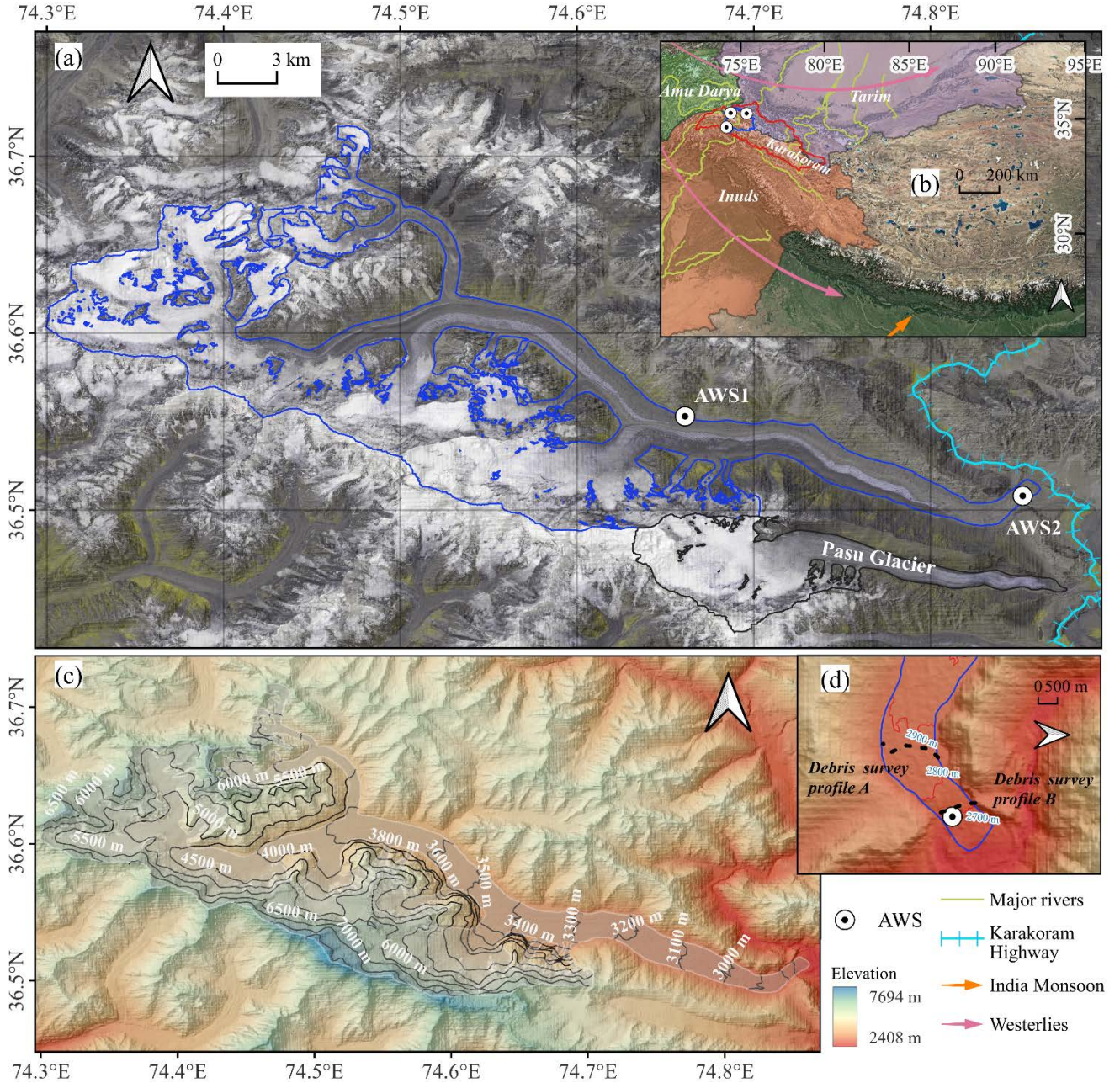
101 2 Study site

102 The Batura Glacier, located in northwest Karakoram, stands as one of the most prodigious valley-type glaciers
103 in the lower latitudes, extending over a length of more than 50 km and encompassing an expansive area exceeding
104 310 km² (Xie et al., 2023) (Figure 1). Approximately 24% (~76 km²) of the glacier's area is covered with debris
105 (Xie et al., 2023), while its thickness in the part below 3000 m a.s.l. surpasses 50 cm (Gao et al., 2020). Due to the
106 heavy debris cover, Batura Glacier presents a hummocky topography and a concave longitudinal surface profile.
107 Because of the large difference in density between ice and debris, the heavily debris-covered glacier section has
108 higher hydrostatic pressure at the glacier bottom (Gao et al., 2020).. Influenced by the prevailing Westerlies, the
109 Batura Glacier receives abundant snowfall (exceeding 1000 mm w.e. at altitudes above 5000 meters) in the high-
110 altitude region (Lanzhou Institute of Glaciology and Geocryology, 1980). In addition, the interaction of the South
111 Asian monsoon and Karakoram vortex cause localised cooling over Karakoram, leading to a low air temperature in
112 summer (Dimri, 2021; Forsythe et al., 2017). As observed by (Lanzhou Institute of Glaciology and Geocryology,
113 1980), the Batura glacier is characterized by a relatively low average annual air temperature compared to observed
114 glaciers in Tianshan and Himalayas, particularly near the annual snowline, where temperatures close to, or below,
115 0 °C endure throughout the year, averaging approximately -5°C annually. The glacier displays a rapid flow velocity,
116 with a maximum rate reaching up to 517.5 m yr⁻¹, facilitated by a high rate of mass turnover, and undergoes frequent
117 periods of advance and retreat, while remaining devoid of any surging events (Bhambri et al., 2017).

118 Since the comprehensive investigation on Batura Glacier conducted by Lanzhou Institute of Glaciology and
119 Geocryology during 1974-1975, there has been a scarcity of systematic observations and studies on this glacier.

120 Contemporary investigations of Batura Glacier primarily utilize remote sensing observations, focusing on the
 121 glacier dynamics and long-term mass balance, e.g. Rankl and Braun (2016), Wu et al. (2021). There is a challenge
 122 in understanding glacier ablation, associated secondary hazards such as glacier floods, and the contribution of
 123 glacier runoff to river replenishment.

124



125 **Figure 1** Study area. (a) Image of Batura Glacier in 2019 (Synthesized using Sentinel-2 data). (b) Geographic
 126 location of Batura Glacier, with the red line outlining the Karakoram, and the blue line outlining the Hunza valley
 127 within which Batura Glacier is situated. The three weather stations labeled are Khunjerab, Ziarat, and Naltar. (c)
 128 Surface topography of Batura Glacier. (d) Measurement profiles of debris thickness.

130 3 Data and methods

131 3.1 Data

132 3.1.1 Observations

133 An automatic weather station (AWS 1, 74.661° E, 36.550° N, 3390 m) was set up at Batura Glacier on 23
134 September 2013 by the Northwest Institute of Eco-Environment and Resources, Chinese Academy of Sciences
135 (Figure 1a) and has been in continuous operation since then (Figure 1a). Meteorological variables observed at the
136 station are maximum/minimum wind speed and direction, maximum/minimum air temperature, relative humidity,
137 atmospheric pressure, upward and downward long- and shortwave radiations and precipitation, recorded on a daily
138 basis. In this study, we use data from AWS1 in the period 23 September 2013 to 9 May 2018 for the bias correction
139 of HAR v2 (High Asia Refined) reanalysis data (Wang et al., 2020) (see section 3.1.2) and for the accuracy
140 assessment of the energy and mass balance simulations. The second AWS (AWS 2, 74.851° E, 36.506° N, 2664 m)
141 was set up in August 2019 by Yunnan University on a debris-covered part of the tongue of the Batura Glacier. The
142 AWS2 records the same climatic factors as AWS1, but it doesn't measure precipitation. We use data from AWS 2
143 between 1 September 2019 to 25 November 2020 to evaluate the reliability of parameters for energy balance in the
144 debris-covered area. The technical specifications for the sensors used in both AWSs are detailed in Table S1. We
145 additionally used daily maximum/minimum temperatures and precipitation from stations at Khunjerab, Ziarat, and
146 Naltar in the Hunza Valley (Figure 1b) covering the period from January 1, 1999 to December 31, 2008, provided
147 by Water and Power Development Authority (WAPDA), Pakistan, to assess the accuracy of HAR in the Hunza basin.

148 The debris thickness at the terminus of the Batura Glacier (2014) was surveyed by WAPDA and provided by a
149 research group at COMSATS University Islamabad of Pakistan. Additionally, we collected measurements of debris
150 thickness at six sample points near AWS 2 during fieldwork in 2019.

151

152 3.1.2 Reanalysis data

153 The HAR reanalysis data is a product derived from the dynamical downscaling process using the Weather
154 Research and Forecasting (WRF) model. The driving data for the first version is FNL (Final) Operational Global
155 Analysis data, while the second version uses ERA5-atmospheric (0.25°) data (Wang et al., 2020). Compared to the
156 first version, the second version expanded the spatial range of the simulation and extended the time range and will
157 continue to receive updates (see Wang et al. (2020)). In the production of the meteorological variables, the dynamic
158 assimilation of downscaled results was achieved using satellite products and ground observations such as wind

159 speed, wind direction, temperature, and geopotential height. This process significantly improved the accuracy and
160 credibility of the downscaling simulation. Notably, the HAR product has shown great potential in reflecting regional
161 water vapor transport processes (Curio et al., 2015) as well as spatial heterogeneity and seasonal variations in
162 precipitation and temperature (Maussion et al., 2014).

163 3.1.3 Other data

164 The geodetic mass balances for Batura Glacier generated by Brun et al. (2017), Wu et al. (2020), Shean et al.
165 (2020), and Hugonnet et al. (2021) were utilized to validate the energy and mass balance simulation results. These
166 mass balance data were derived from elevation differences with some assumptions such as ice density, etc. With the
167 exception of the five-year mass balance (2000-2020) produced by Hugonnet et al. (2021), the other data only show
168 the long-term mass balance status after 2000. Time ranges for all mass balance data can be found in Figure 3. The
169 30 m resolution DEM from the Shuttle Radar Topography Mission (SRTM) was used to generate required terrain
170 factors, while the glacier boundary was defined using the most recent delineation published by Xie et al. (2023).

171 3.2 Methods

172 3.2.1 The physically-based energy-mass balance (EMB) model

173 The EMB model for snow and ice is a distributed model that combines surface energy processes with a
174 subsurface evolution scheme for snow and ice (COSIPY v1.3) which was developed by Sauter et al. (2020). Details
175 of the model relating to applied parametrizations, physical principles and technical infrastructure have been
176 described in Huintjes et al. (2015b), Sauter et al. (2020) and (Arndt and Schneider, 2023). In common with previous
177 energy balance models, the surface energy budget is defined as the sum of the net radiation, turbulent heat fluxes
178 (including sensible heat flux q_{sh} and latent heat flux q_{lh}), conductive heat flux (q_g), sensible heat flux of rain (q_{rr})
179 and melt energy (q_{me}) (Eq.1). The net radiation is the sum of the net shortwave radiation calculated from incoming
180 shortwave radiation ($q_{sw_{in}}$) and surface albedo (α), incoming longwave radiation ($q_{lw_{in}}$) and outgoing longwave
181 radiation ($q_{lw_{out}}$). To link the surface energy balance to subsurface thermal conduction, the snow/ice surface
182 temperature ($T_{s_{si}}$) is defined as an upper Neumann boundary condition. The penetrating scheme of shortwave
183 radiation is based on Bintanja and Van (1995).

$$184 \quad q_{me} = q_{sw_{in}}(1 - \alpha) + q_{lw_{in}} + q_{lw_{out}} + q_{sh} + q_{lh} + q_{rr} + q_g \quad (1)$$

185 The glacier melt is solved using q_{me} and penetrating shortwave radiation, while the sublimation is solved
186 using q_{lh} . Combined with the snowfall and refreezing of meltwater (or rain), the total mass balance of the glacier
187 surface can be calculated (Eq.2). The sum of subsurface melt (m_{sub}) due to penetrating shortwave radiation energy

188 and the refreezing of meltwater (or rain) (refreeze), is defined as the internal mass balance. The internal ablation
 189 occurs when temperature at a specific layer reaches the melting temperature (T_m). Internal meltwater, in combination
 190 with infiltrated surface meltwater, can be stored in the snow layers. Once a layer becomes saturated, meltwater will
 191 drain into the next layer until the liquid water content within all layers is less than a defined ratio, or else the
 192 meltwater runs off when it reaches the lowest model layer. In this process, a part of the meltwater refreezes when
 193 the temperature at a layer is less than T_m . Full details for resolving mass and energy budgets in the EMB can be
 194 found in Sauter et al. (2020).

$$195 \quad mb = \left(\frac{q_{me}}{L_m} + \frac{q_{lh}}{L_v} + \text{snowfall} \right) + (m_{sub} + \text{refreeze}) \quad (2)$$

196 where L_m is the latent heat of ice melt and L_v is the latent heat of sublimation or condensation.

197 The debris energy balance is calculated according to the model of Reid and Brock (2010), and the reader is
 198 referred to their paper for a detailed description of the model. The sum of energy fluxes at the surface is essentially
 199 the same as Eq. 1, but because debris does not melt, the debris surface temperature ($T_{s,d}$) is assumed to change such
 200 that these fluxes sum to zero:

$$201 \quad q_{swin}(1 - \alpha) + q_{lw_{in}}(T_{s,d}) + q_{lw_{out}}(T_{s,d}) + q_{sh}(T_{s,d}) + q_{lh}(T_{s,d}) + q_{rr}(T_{s,d}) + q_g(T_{s,d}) = 0 \quad (3)$$

202 The circularity in solving for $T_{s,d}$ is resolved using a numerical Newton-Raphson method (Eq. 4). Conduction
 203 through the debris is then calculated using a Crank-Nicholson scheme with intermediate temperature layers for a
 204 set depth, and boundary conditions determined by the newly calculated $T_{s,d}$ and the temperature at the debris-ice
 205 interface, which is assumed to stay at zero (Eq. 5). The ablation rate is determined from the conductive heat flux to
 206 the first (uppermost) ice layer, found using the temperature gradient between the lowest debris layer and the ice (Eq.
 207 6). The detailed solution processes for Eq. 4~6 can be found in Figure 2 and Appendix materials in Reid and Brock
 208 (2010).

$$209 \quad T_{s,d}(n+1) = T_{s,d}(n) - \frac{fun(T_s(n))}{fun'(T_s(n))'} \quad (4)$$

210 where, $T_{s,d}(n)$ and $fun(T_{s,d}(n))$ refer to the temperature and the total energy flux at nth debris layer. The
 211 termination condition for this solution is set as $T_s(n+1) - T_s(n) < 0.01$.

$$212 \quad q_G = -k_d \left(\frac{dT_s}{dz} \right) \approx k_d \frac{T_{s,d}(N-1) - T_m}{h} \quad (5)$$

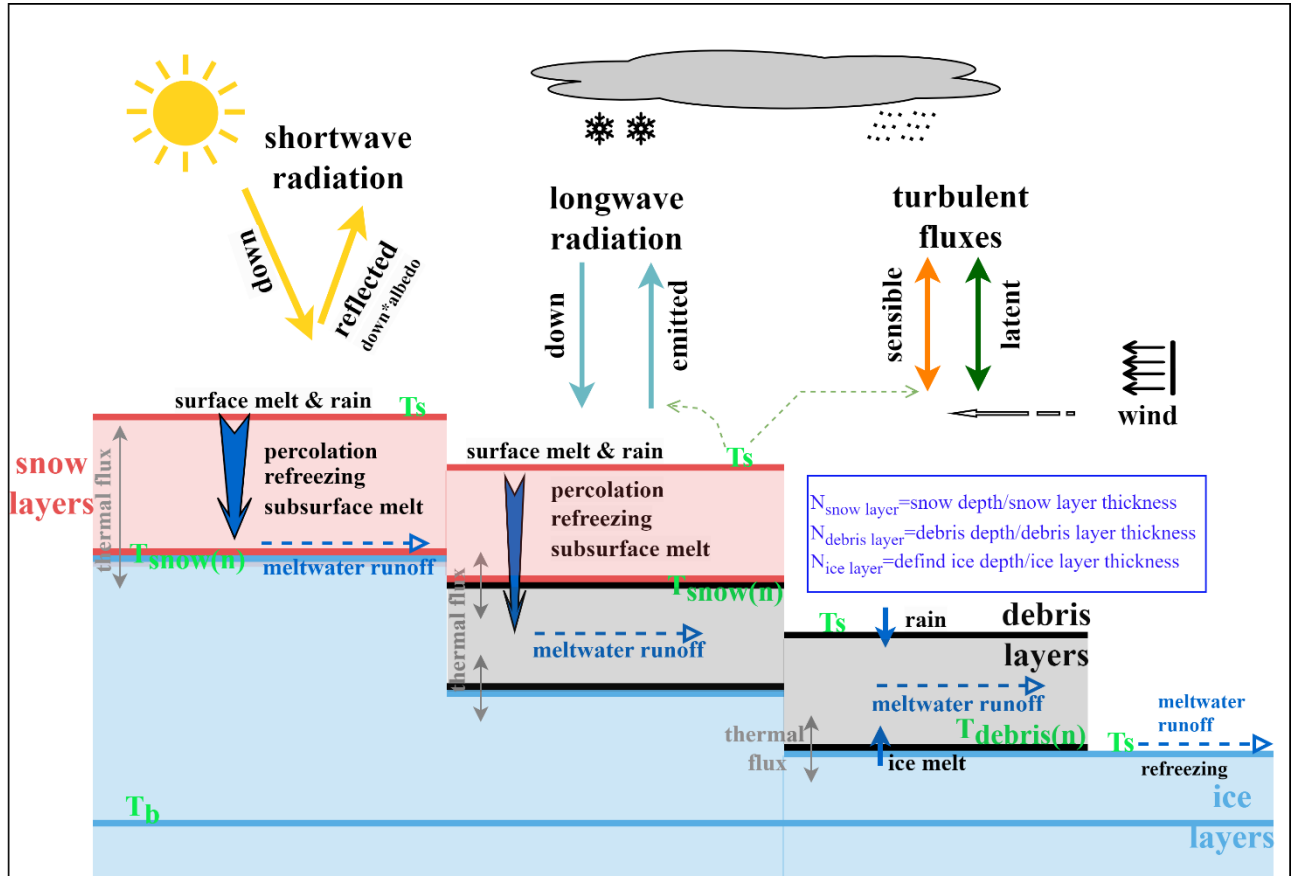
$$213 \quad Melt_{deb} = \frac{q_G}{\rho_i L_f} \quad (6)$$

214 where, h represents the thickness of each layer, n represents nth debris layer, N represents the number of
 215 calculation layers, T_m represents the melting temperature of ice, and k_d is the thermal conductivity of

216 supraglacial debris. $Melt_{deb}$ refers to the ablation rate of ice at the debris interface.

217 In the model run, the initialization of the model was firstly conducted using the defined parameters. The most
218 important in this step was the establishment of the temperature profile, which was initialized with air temperature
219 (T_a) and bottom temperature (T_b) by using linear interpolation. The second step involved recalculating the
220 temperature profile, involving two scenarios: (1) In debris-free areas, the temperature profile was calculated entirely
221 according to the COSIPY. Initially, the temperature profile was computed without considering the impacts of
222 refreezing or subsurface melt but factoring in temperature increase due to penetrating radiation. If a snow/firn pack
223 is present, the densification of the dry snow pack was calculated using an empirical relation (Herron and Langway,
224 1980). After densification, the available surface and subsurface meltwater percolated downward, with a small
225 amount retained in each layer. Subsequently, the temperature changes resulting from refreezing of meltwater were
226 computed, updating the subsurface layer temperature. In debris-covered areas, when snow presented, the snow-
227 debris interface temperature was first obtained using the snow layer temperature update scheme of the COSIPY
228 model. This temperature was then set as the debris surface temperature. By defining the debris-ice interface
229 temperature as zero, the debris layer temperature was then calculated using Eq. 5. In the absence of snow, the model
230 employs the debris layer temperature update scheme described by Reid and Brock (2010). The third step involved
231 using the surface temperature obtained from the second step, combined with glacier surface meteorological
232 parameters, to calculate the surface energy balance and surface melt. The primary physical processes of the model
233 are illustrated in Figure 2. In this study, a two-year spin-up was implemented to allow the model to adapt to the
234 surrounding conditions (Huintjes et al., 2015a).

235



236

237 **Figure 2** General scheme of the model used in the current study with fluxes and physical processes. T_s represents
 238 surface temperature, solved for using the heat conduction equation. The solution process varies depending on the
 239 different surface cover conditions of the glacier. T_s is a crucial variable linking the energy exchange between the
 240 glacier and the atmosphere. T_s is primarily used to calculate sensible heat flux and emitted longwave radiation.

241 Reflected shortwave radiation is mainly determined by surface albedo. In the case of snow cover, the albedo
 242 changes continuously with snowmelt and densification. $T_{snow(n)}$ represents the temperature of the nth snow layer,
 243 reflecting the energy flux at the snow-ice interface or snow-debris interface. $T_{debris(n)}$ represents the temperature of
 244 the nth debris layer, reflecting the energy flux at the debris-ice interface. These two variables are important for
 245 characterizing the internal energy balance of the glacier.

246 In the model, the layers of snow, debris, and ice were dynamically calculated based on their individually
 247 specified thicknesses. Considering that the temperature of the ice layer does not change with increasing thickness
 248 below a certain depth in glaciers, a depth of 10 m for the ice layer was set, following Huintjes (2014). As ice
 249 temperature cannot exceed 0 °C, the boundary conditions at snow-debris interfaces were configured similarly,
 250 following an analogous scenario that the temperature of snow-debris interface remains below 0 °C (Giese et al.,
 251 2020). Based on this, we made the assumption that any rain or snowmelt water does not refreeze within the debris

252 layer, and the infiltration of such water does not alter the temperature of the debris layer. The temperature boundary
253 condition at the debris-ice interface follows Reid and Brock (2010), ensuring that the temperature of debris-ice
254 interfaces remains below 0 °C. For the lower boundary condition (bottom temperature), values referenced from
255 Huintjes (2014) are employed, derived from observational data. To prevent ice layer temperatures from exceeding
256 freezing level, a heating mechanism is applied to the ice layer above the bottom layer, directing above-freezing
257 energy into the melting process.

258 In this study, the model simulations were conducted using a high-performance server, equipped with dual
259 Intel Xeon CPU E5-2687W processors (48 threads), 768 GB of RAM, and dual Quadro P6000 (24G) GPUs for
260 acceleration. We conducted simulations that compared scenarios with and without supraglacial debris on the Batura
261 Glacier to assess the influence of debris on mass balance.

262 3.2.2 Model setup and input data

263 In this study, HAR v2 data were used to drive the model to simulate the energy and mass balance of the Batura
264 Glacier from 2000 to 2020. The meteorological variables in HAR v2 selected to meet the requirements of the energy
265 balance simulation include precipitation, air temperature at 2 m, wind speed (u - and v - components at 10 m),
266 atmospheric pressure, specific humidity, downward shortwave radiation, and cloud cover. The 10 m wind speed was
267 converted to 2 m using an empirical formula provided by Allen et al. (1998), while specific humidity was converted
268 to relative humidity using the formula given by Bolton (1980) utilizing the 2 m air temperature and atmospheric
269 pressure. Air temperature was calibrated at the basin scale using a gridded bias factor. The gridded bias was
270 interpolated by the nearest-neighbor method, with the bias at each station calculated between the observed and HAR
271 temperatures. After correction, a small bias range of $\pm 1^\circ\text{C}$ was observed between HAR temperature and station
272 temperature, with a Pearson correlation coefficient of 0.98. Details regarding the precipitation calibration can be
273 found in Appendix A1. Due to lack of observations for other variables, no further validation before statistical
274 downscaling was conducted at the basin scale in this study. However, minor adjustments were applied for
275 downscaled other variables. These adjustments were made using scale factors calculated through the least squares
276 method, considering the downscaled results and observed values at the two stations on Batura glacier.

277 We utilized the data from Rounce et al. (2021) based on an inversed energy balance modeling procedure to
278 calculate debris thickness inputs. The debris thickness with a 100 m resolution is resampled to 300 m using an
279 inverse distance weighted interpolation method to match the simulation resolution. We validated the simulated
280 debris thickness using observed data, which showed an average deviation of 6 cm. However, it should be noted that

281 the Rounce et al. (2021) results significantly underestimated the debris thickness at certain locations near the
282 terminus of the glacier. For instance, at AWS2, the observed debris thickness was approximately 1.13 m, whereas
283 the inverted thickness was only 0.47 m.

284 The simulation was conducted at a spatial resolution of 300m and a temporal step of 1 day. The primary
285 meteorological drivers, such as precipitation and temperature, were calibrated using data from meteorological
286 stations. We employed statistical methods to downscale all meteorological inputs to a resolution of 300 m (for more
287 details, please refer to the supplementary material). The simulation grid was constrained using the glacier boundaries
288 from Xie et al. (2023), and no ice flow dynamic adjustments for the glacier were considered. In this study, we also
289 conducted a simulation on the debris-free Pasu Glacier situated adjacent to the Batura Glacier to make a comparative
290 study of mass and energy balance. We assumed that Pasu Glacier experiences similar climatic conditions to Batura
291 Glacier. The physical parameters used for this simulation are identical to those from AWS1 on Batura Glacier (see
292 the Section 3.2.2) and we compared the simulated mass balance with the geodetic mass balance to test the extension
293 of these parameters.

294

295 3.2.3 Parameters calibration/ validation

296 In this study, we used value ranges for most parameters which have been acquired from empirical equations,
297 large extent observations, or physical process simulations in previous studies e.g., Reid and Brock (2010), Mölg et
298 al. (2012), Hoffman et al. (2016), Zhu et al. (2020), and Sauter et al. (2020). Since the model is very complex, it
299 was necessary to constrain the number of calibrated parameters to limit the modeling effort. Through sensitivity
300 analysis at AWS1, we identified four parameters that have significant impacts on simulating mass balance: ice
301 albedo and roughness length of ice, which constrain ice melting through the radiative and turbulent energy fluxes,
302 respectively; and firn albedo and roughness length of firn, which control the snow evolution processes. By adjusting
303 these parameters within a specific step range, our goal was to achieve the closest match between simulated albedo
304 and longwave radiation and their observed values using a self-defined $RMSE_{score}$. The $RMSE_{score}$ is calculated
305 as Eq.7.

$$306 \quad RMSE_{score} = \sum_{k=1}^n \sqrt{\frac{1}{m} \sum_{i=1}^m (obs_std_{k,i} - sim_std_{k,i})^2} \quad (7)$$

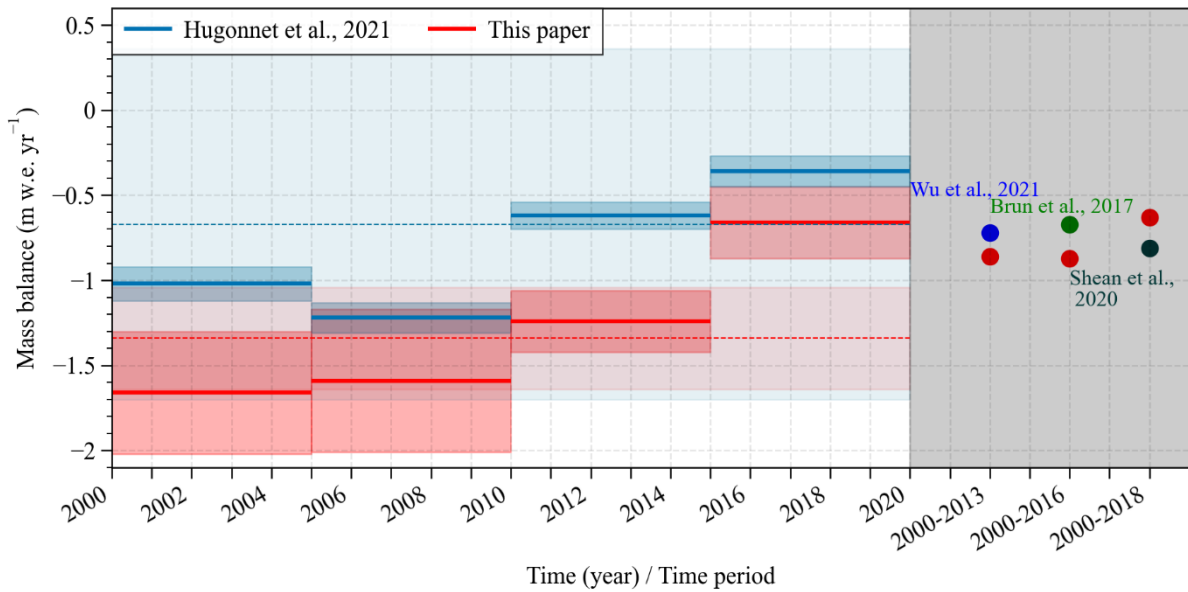
307 Where n represents the number of variables, obs_std_k and sim_std_k represent the standardized observed and
308 simulated values of k th variable. The standardization is achieved through min-max normalization. For the purpose
309 of comparison, the final $RMSE_{score}$ is presented as a standardized result ranging from 0 to 1. A smaller $RMSE_{score}$

310 indicates better performance of the model. By comparing the $RMSE_{score}$, we can easily determine the optimal
311 values for calibrating the parameters (Figure S1). The final determined values for the selected parameters are show
312 in Table S2. With these parameters, the RMSE between simulations and observations on albedo and outgoing
313 longwave radiation are 0.09 and 18.93 W/m², respectively, and there is a high degree of correlation between
314 observations and simulations on annual variations, with Pearson correlation coefficients (r) of 0.83 for albedo and
315 0.86 for outgoing longwave radiation (Figure S2). After determining the primary parameters, we fine-tuned some
316 independent parameters such as albedo timescale, albedo depth scale, temperature threshold of rain/snow ratio,
317 ensuring a comparable level of simulated mass balance with geodetic mass balance. The simulated mass balance
318 agrees well with the geodetic mass balance, with an average bias of 0.27 m w.e. In particular, there is a strong
319 agreement between the results from Hugonnet et al. (2021) and our simulations in terms of the trend observed from
320 2000 to 2020 (Figure 3). This indicates that the parameters used in our study can reliably estimate the mass and
321 energy budget.

322 A point simulation at AWS2 was conducted to calibrate and validate the parameters required to simulate energy
323 balance in debris layers. Following Giese et al. (2020), we evaluated the model parameters by optimizing the
324 agreement between the simulated surface temperature and the surface temperature recorded by AWS 2 (the
325 temperature probe is buried ~ 2 centimeters below the debris surface). The parameters calibrated at AWS1 were
326 applied unchanged to AWS2, with adjustments only made to the debris thermal conductivity and debris albedo
327 during the simulation process. The calibration process can be observed in Figure S3. Figure 4 depicts the
328 comparative analysis of the observed station temperature and the simulated temperature, using the optimized values
329 for debris thermal conductivity and albedo, revealing a strong consistency between the two over time, with a
330 correlation coefficient of 0.87, although there is a tendency to underestimate the temperature in late summer and
331 autumn, and overestimate temperature in late winter. The correlation of observed and simulated temperature for the
332 annual cycle is 0.96, while the RMSE during the simulation period is 0.86 °C.

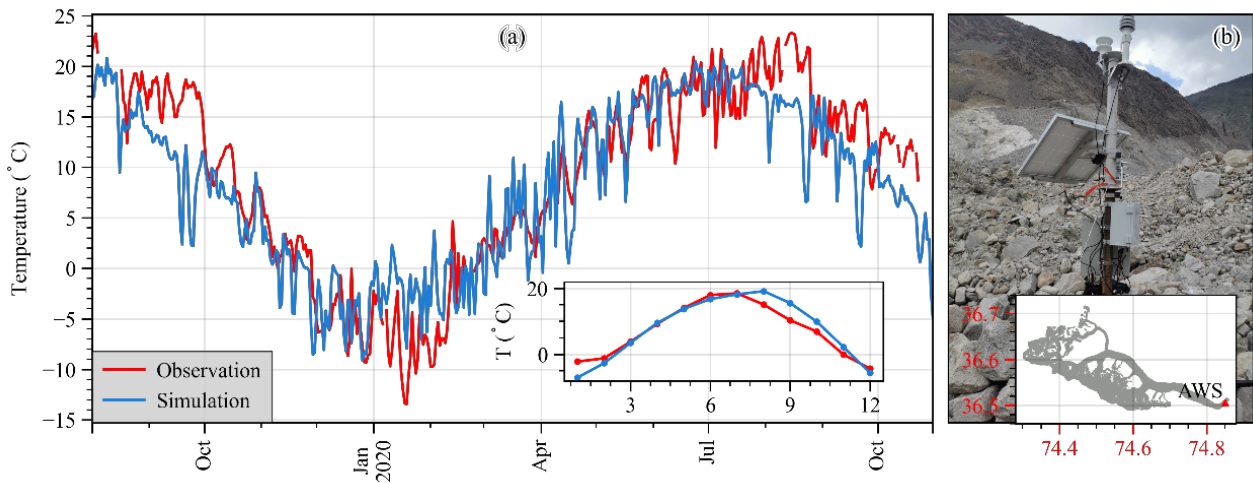
333 The parameter evaluation process at AWS 2 supports the applicability and scalability of the parameters
334 calibrated at AWS1 to other parts of the glacier. Based on the final parameters determined (Tables S2 and S3), the
335 simulated mass balance for the entire glacier is estimated to be -0.23 m w.e. yr⁻¹ (2000-2016). This value closely
336 aligns with the geodetic mass balances derived from remote sensing (-0.18 m w.e.yr⁻¹, spanning the years 2000-
337 2016, Brun et al. (2017); -0.39 m w.e.yr⁻¹, covering the years 2000-2009, Bolch et al. (2017); and -0.24 m w.e.yr⁻¹,
338 covering the years 2000-2014, Wu et al. (2020)). This further supports the robustness of parameter transfer across

339 the glacier.



340

341 **Figure 3** Comparison of simulated and geodetic mass balance over different time periods. To assess the performance
342 of our model, we compared the simulated mass balance with estimates derived from geodetic observations. However,
343 it is important to acknowledge that this approach introduces a degree of dependence between the two results since
344 some model parameters were calibrated using the geodetic mass balance.



345

346 **Figure 4** (a) Observed and simulated surface temperature at AWS 2. (b) Photograph and location of AWS 2 on
347 Batura Glacier. AWS 2 collects data at both daily and hourly intervals, this study utilizes daily records for
348 analysis.

349

350 4 Results and discussions

351 4.1 Glacier climatic-mass-balance dynamics and corresponding energy budgets

352 4.1.1 Energy budgets

353 During 2000-2021, the surface net radiation of the Batura Glacier accounted for the largest proportion of total
354 energy heat flux (46%), followed by sensible heat flux (23%). Latent heat flux (-18%) and conductive heat flux
355 (17%) demonstrated a similar magnitude of contribution to the total energy heat flux, albeit with opposite sign
356 (Table 1).

357 The net shortwave radiation accounted for 85% of the total energy influx (77 W/m^2), while sensible heat
358 constituted 15% (14 W/m^2). Regarding energy sink components, net longwave radiation contributed 57% (52 W/m^2),
359 melt heat 20% (18 W/m^2), latent heat 12% (11 W/m^2), and conductive heat 11% (10 W/m^2). In terms of the energy
360 components that contribute to glacial mass loss, sublimation latent heat accounted for approximately 38%, while
361 the energy for snow/ice melting constituted 62%. For the Batura Glacier, roughly 32% (29 W/m^2 out of 91 W/m^2)
362 of the surface energy influx was consumed by glacier mass loss, a proportion similar to that of Muztag Ata No.15
363 Glacier, which is also situated in the Westerly influenced area (30%, 26 W/m^2 out of 89 W/m^2) (Zhu et al., 2017).
364 However, it is worth noting that the melting heat of the Batura Glacier was significantly higher than that of Muztag
365 Ata No.15 Glacier ($\sim 2 \text{ W/m}^2$), possibly due to differences in surface debris cover between the two glaciers.

366 During the period of accumulation, a notable proportion of 73% of the energy influx of the Batura Glacier was
367 expended through net longwave radiation, with 15% of the energy utilized for snow/ice sublimation, leaving the
368 remaining portion dedicated to thermal conduction within the debris cover or snow layer. In contrast, throughout
369 the ablation season, the energy influx was mostly from net shortwave radiation, specifically amounting to 133 W/m^2 .
370 The conductive heat flux exhibited by the Batura Glacier diverged significantly from debris-free glaciers, such as
371 the Guliya ice cap (Li et al., 2019). In the Batura Glacier, a considerable portion of the energy influx at lower
372 elevations was absorbed by the debris cover, resulting in higher surface temperatures compared to the lower layers,
373 thus yielding heat transfer towards the debris-ice interface. Conversely, in the accumulation area, the primary source
374 of energy was dedicated to heating the snow layer. It became evident that during the ablation season, the debris
375 cover assumed a more prominent role, ultimately leading to an overall negative thermal conduction.

376

377

378

379

380 **Table 1** The energy budget on Batura Glacier. lw_{in} and lw_{out} denote incoming and outgoing longwave
 381 radiation, sw_{in} and sw_{out} denote incoming and outgoing shortwave radiation, sh and lh represent the
 382 sensible heat flux and latent heat flux, g represents conductive heat flux, and me represents melt energy. All
 383 values are expressed in W/m^2 .

Periods	lw_{in}	lw_{out}	sw_{in}	sw_{out}	Net lw	Net sw	Net radiation		sh		lh		g		me
							—	%	—	%	—	%	—	%	
Annual average	212	-264	249	-172	-52	77	25	42	14	23	-11	18	-10	17	18
Ablation (6-9)	231	-293	345	-212	-62	133	71	65	-7	6	-15	14	-16	15	33
Accumula tion (10-5)	202	-249	187	-153	-48	34	-12	19	32	52	-10	16	-8	13	0

384

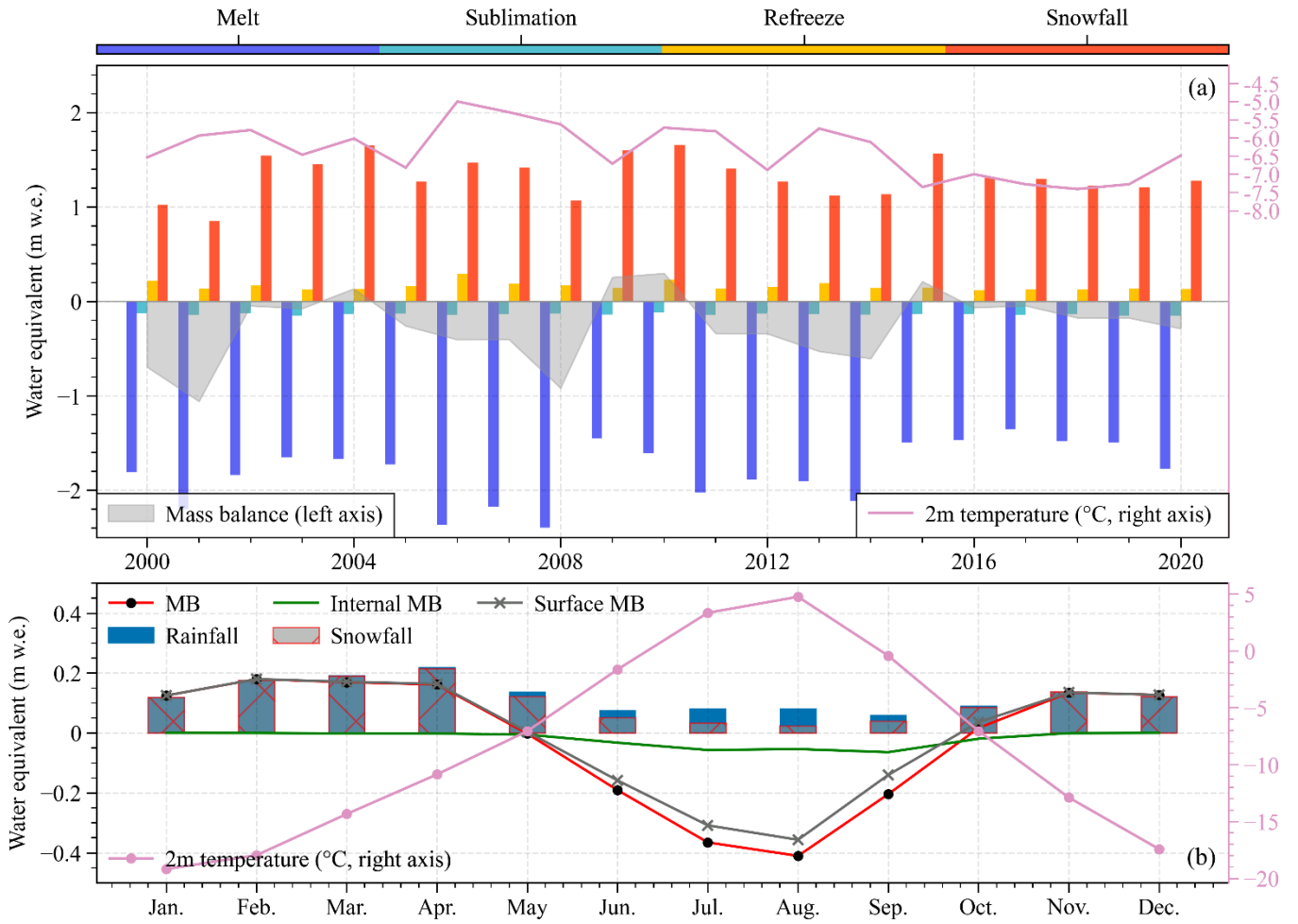
385 4.1.2 Mass balance history

386 The results from the EMB model show that the average mass balance of the Batura Glacier during the studied
 387 period was -0.262 ± 0.561 m w.e. yr^{-1} (Table 2). The glacier experienced its highest positive mass balance in 2010
 388 (0.32 m w.e. yr^{-1}) and its greatest negative mass balance in 2001 (-1.19 m w.e. yr^{-1}). Snowfall was the primary source
 389 of glacier mass gain, accounting for 89% of the total mass gain. Refreezing mitigated the internal melting caused
 390 by radiation penetration and contributed to 11% of the mass accumulation. Glacier melting constituted 92% of the
 391 mass loss, while sublimation/evaporation, which exhibited minimal interannual variability, contributed only 8% to
 392 the mass loss.

393 The model simulations show a decline in glacier ablation after 2008, accompanied by a decrease in the absolute
 394 magnitude of the mass budget over the study period (Figure 5a). Independent measurements of thinning rates at the
 395 glacier terminus measured by ground-penetrating radar, declined from 4.58 m yr^{-1} between 1974-2000 to 0.59 m yr^{-1}
 396 after 2000 (Gao et al., 2020), implying a similar reducing trend in surface melt rate, which further supports the
 397 EMB results. The striking decrease in thinning rates at Batura Glacier for the periods 1974-2000 and 2000-2017,
 398 and decline in modeled ablation since 2008 might be linked to regional climate fluctuations. Previous studies based
 399 on station observations have indicated a notable cooling trend in the upper Indus River basin during the summer
 400 months, particularly in July, September, and October, from 1995 to 2012 (Hasson et al., 2017). Moreover, there was
 401 a lack of long-term warming during the winter months over the same period (Hasson et al., 2017). Forsythe et al.

402 (2017) suggested that the summer temperature in the Karakoram was relatively low and exhibited a decreasing trend
403 due to the influence of the Karakoram vortex (KV). This influence may have contributed to the notably higher
404 positive accumulated temperatures pattern observed from 1970 to 2000 compared to those recorded after 2000, as
405 shown in Figure 4b of Forsythe et al. (2017). Our analysis on air temperature in the Hunza basin from 1980~2020,
406 utilizing ERA5 data, corroborates these findings (Figure S4).

407 As shown in Figure 5b, the variations in internal mass balance and surface mass balance are generally
408 consistent throughout the year, both showing a negative mass balance from June to September. During this period,
409 there was a high shortwave radiation and, consequently, a great amount of shortwave radiation penetrated into
410 snow/ice. This increased ablation resulted from penetration radiation, coupled with relatively high temperature,
411 reducing the rate of refreezing, and thus causing a negative internal mass balance. The mass budgets in May and
412 October were transitional between accumulation and ablation periods. The seasonal pattern on mass balance
413 observed in this study is generally similar to that of the Siachen Glacier, East Karakoram presented by Arndt and
414 Schneider (2023). Both glaciers exhibit a characteristic of winter/spring accumulation. However, the modeled
415 meltwater during the ablation season was found to be significantly lower for Siachen Glacier compared to Batura
416 Glacier. It is worth noting that Arndt and Schneider (2023) did not consider the impact of supraglacial debris cover
417 on glacier melt, which is known to be substantial (Agarwal et al., 2016). Even without considering the debris cover,
418 the mass balance of Siachen Glacier, as indicated by Arndt and Schneider (2023), can still remain in equilibrium,
419 largely depending on the precipitation and temperature driving data. On the other hand, in the simulation study
420 conducted by Kumar et al. (2020), Siachen Glacier exhibited a negative mass balance during the same period, with
421 the average temperature and precipitation being higher than those used by Arndt and Schneider (2023). This suggests
422 that simulation results can be considerably influenced by model inputs, and this will be discussed in Section 4.5.



423

424

Figure 5 Interannual (upper panel) and mean monthly (lower panel) characteristics of the glacier-wide average of mass components on Batura Glacier over the study period. The MB denotes mass balance. The 2m temperature is obtained from the simulated results.

426

427

Table 2 Mean values of the mass balance components of Batura glacier over 2000 to 2020.

	Mass balance	Snow accumulation	Surface melt	Refreezing	Sublimation
Values (m w.e. yr ⁻¹)	-0.262±0.561	1.325±0.174	1.613±0.394	0.162±0.125	0.136±0.005
Proportion of mass gain (loss) (%)	—	89	(92)	11	(8)

428

429

Over the study period, the glacier demonstrated a positive rate of annual mass balance change of 0.023 m w.e.

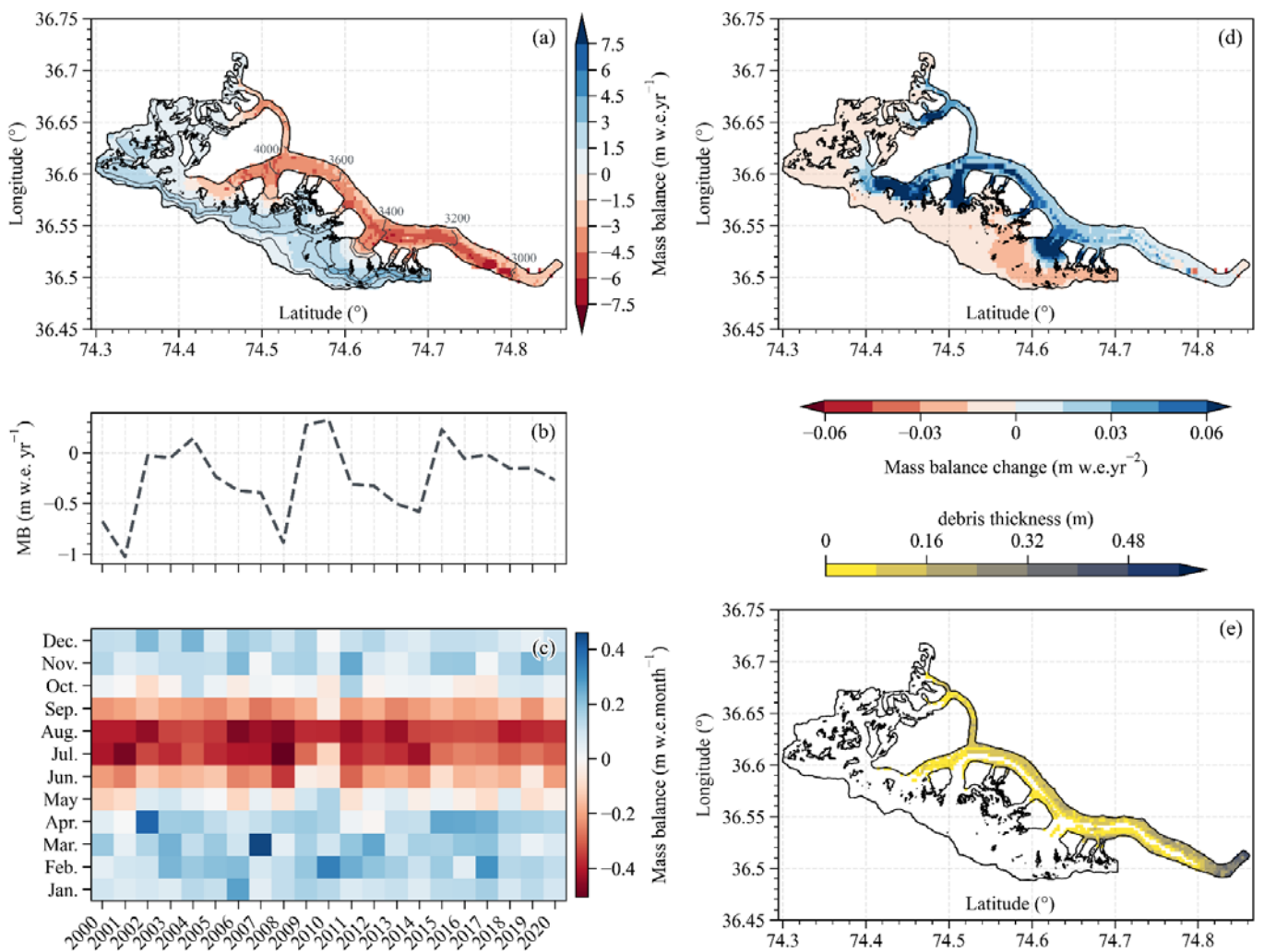
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yr⁻², indicating the glacier's mass balance was becoming less negative and approaching equilibrium between 2000-

431 2020 (Figure 6a, b and d). Particularly noteworthy is the trend of decreasing mass loss across the ablation zone,
432 which is particularly pronounced in the junction where debris cover and bare ice intersect and the tributary where
433 debris cover is thin or absent (Refer to debris cover in Figure 6e), which indicates a reduction in melt (Figure 6b).
434 Given the rate of mass balance change over time (reduction of melt) is highest in these areas, the mass changes in
435 these areas probably have a large impact on the trend of decreasing negative mass balance.

436 Across the entire accumulation zone, a slight decrease in mass gain over the 2000-2020 period was observed,
437 with a more pronounced reduction in mass gain observed on the southern flank of the accumulation area, likely
438 associated to diminished winter snowfall. From a mass budget perspective, the glacier's mass balance appears to be
439 approaching equilibrium, likely due to the reduced melting during the months of June and July (Figure 6c). For
440 instance, in years characterized by a positive mass balance, such as 2010, the duration of mass accumulation in
441 spring extended, accompanied by minimal mass loss during June and July. The glacier's mass balance generally
442 followed a cyclic pattern spanning roughly five-seven years. The mass balance has become more negative after
443 2016, possibly indicating a phase of reduced snow accumulation gain (Figure 6c).

444



445
 446 **Figure 6** Spatial distribution of the annual mass balance over the 2000-2020 period (a). Time series of modeled
 447 annual (b) and monthly (c) mass balance from 2000-2020. Spatial distribution of the annual mass balance change
 448 rate over the 2000-2020 period (d). Spatial distribution of debris thickness (e)

449 4.2 Energy and mass budgets along the altitudinal profile

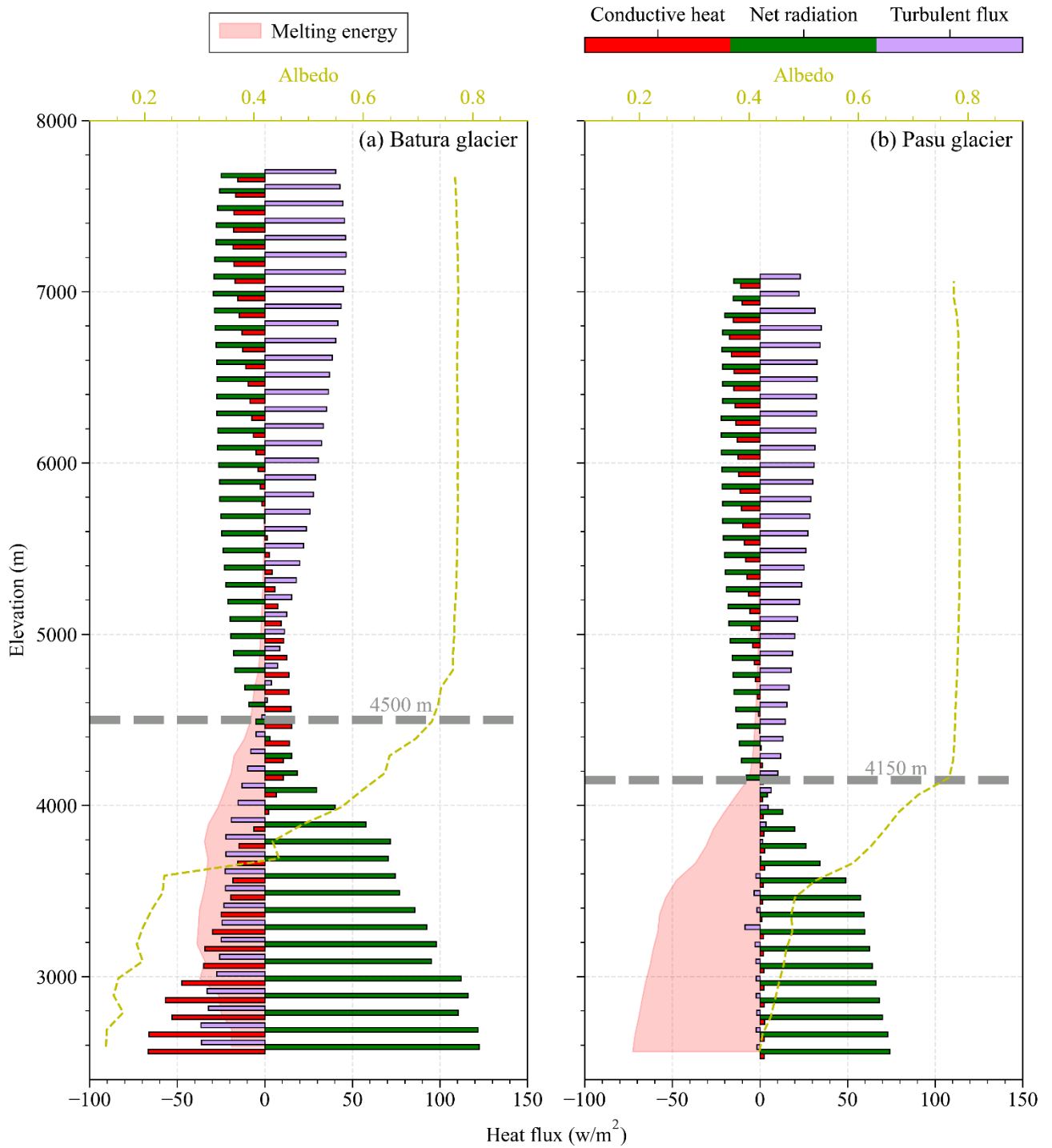
450 A significant heterogeneity of mass balance was observed in the Batura Glacier. The mass gain in the glacier
 451 accumulation zone can reach up to almost 2 m w.e., whereas terminus melting exceeded 4 m w.e. between 3000-
 452 3800 m, with the maximum melting of 4.6 m w.e. occurring within the elevation range of 3350-3450 m. Mass
 453 balance exhibited discernible altitude-dependent distribution, whereby the most substantial melting was observed
 454 not at the terminus but rather in the range between 3000 and 3400 m (Figure S5a).

455 A comparative analysis was performed to understand the variations in mass balance across different elevation
 456 zones between Batura Glacier and Pasu Glacier. The equilibrium line altitude (ELA) of the Batura Glacier (4500 m)

457 was significantly higher than that of the Pasu glacier (4150 m). Below the ELA, both glaciers exhibit gentle overall
458 slopes, leading to high receipt of solar shortwave radiation. As shown in Figure 7, the net radiation of the Batura
459 Glacier was significantly larger than that of the Pasu glacier, primarily attributable to surface albedo disparity. The
460 Pasu Glacier's surface primarily comprises firn or ice, whereas the Batura Glacier is largely covered with fragmented
461 rocks with associated lower albedo. Evidently, the melt energy for the Batura Glacier is less than that of the Pasu
462 Glacier, chiefly due to heat conduction between debris layers, which absorb a substantial amount of energy. Overall,
463 the Batura Glacier demonstrated an "arch-shaped" melt energy pattern from its terminus to the ELA, in sharp
464 contrast to the "slope-increasing" pattern exhibited by the Pasu Glacier. This altitude-dependent spatial energy
465 distribution pattern also affects that of the glaciers' melt (Figure S5).

466 Within the regions spanning from the ELA to the zones of maximum snow accumulation (Batura: 4500-5400
467 m, Pasu: 4150-5400 m), glacier mass accumulated rapidly due to significantly heavy snowfall (Figure S5). Turbulent
468 heat exchange intensifies within this altitude range, with melt energy approaching zero. A modest amount of melting
469 resulted in mass accumulation within the snowpack through refreezing (Figure S5). At altitudes exceeding 5200 m,
470 net radiation, turbulent exchange, and conductive heat flux did not demonstrate significant variations. Net radiation
471 was dominated by longwave radiation, and the snow's surface temperature surpassed the air temperature. The glacier
472 acted as an energy source, transferring energy to the atmosphere to maintain energy balance. While the maximum
473 snowfall on the Batura Glacier was similar to that on the Pasu Glacier, the accumulating area was larger. For instance,
474 in the region above 7000 m, up to 1 m w.e. of snowfall was observed on the Batura Glacier (Figures S5). Changes
475 in precipitation not only induced net radiation variations due to snow albedo feedback but also triggered outgoing
476 longwave radiation and sensible heat variations through alterations in surface temperature. This trait aligned with
477 some of the other glaciers in this area, as well as some glaciers in the West Kunlun and Pamir (Li et al., 2019; Zhu
478 et al., 2017; Bonekamp et al., 2019). However, the Batura Glacier exhibited more negative mass balance compared
479 to these glaciers including the Pasu glacier (The geodetic mass balance, as reported by Brun et al. (2017), is -0.01
480 ± 0.05 w.e.m yr⁻¹, while the simulated mass balance in this study is 0.01 ± 0.26 w.e.m yr⁻¹, both for the period from
481 2000 to 2016.).

482



483

484 **Figure 7** Altitudinal distribution of the primary energy balance components for (a) Batura Glacier and (b) Pasu
 485 glacier.

486

487 4.3 Impact of debris cover on glacier mass balance

488 Our findings revealed that the presence of supraglacial debris led to a notable 45% reduction in negative mass

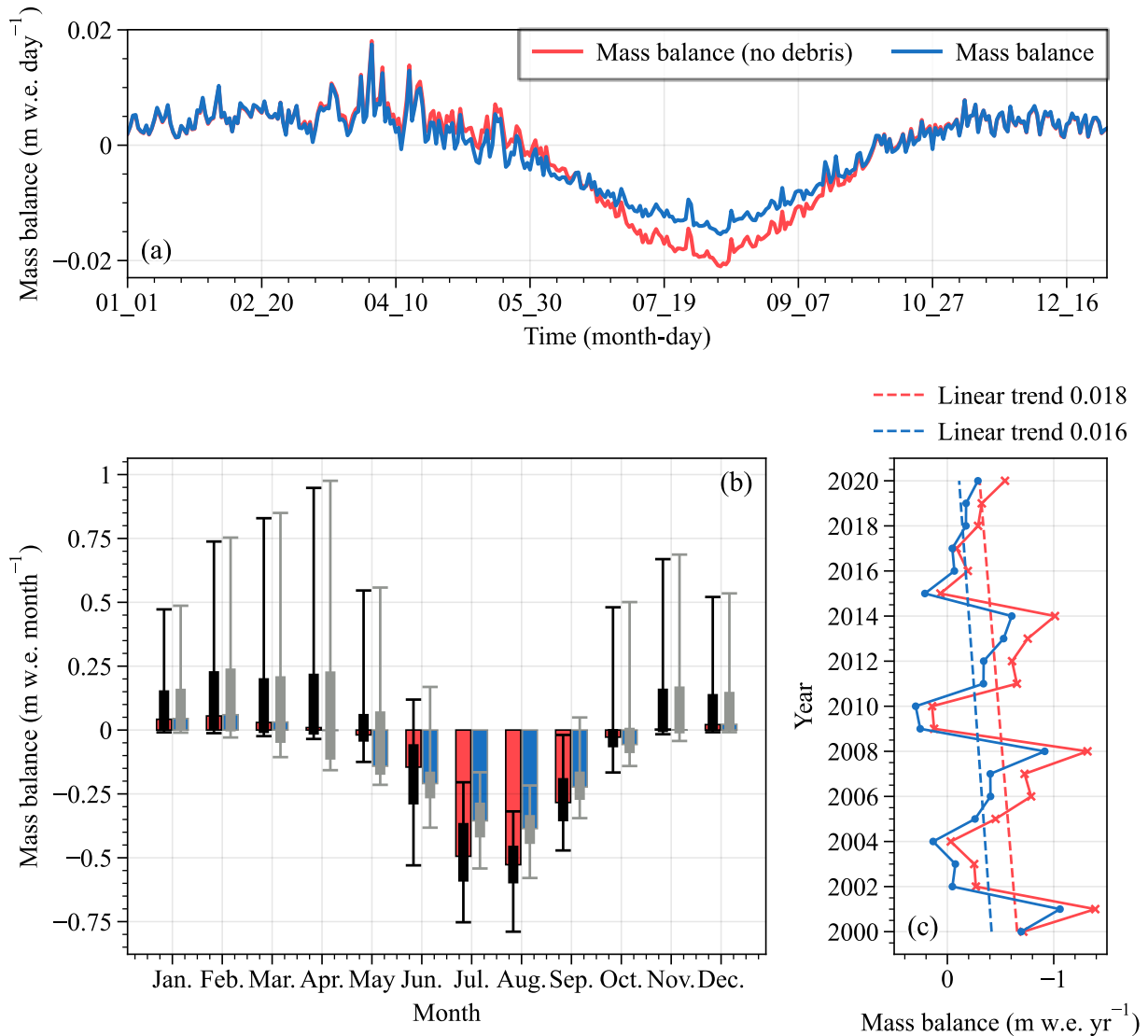
489 balance of the Batura Glacier. Specifically, in the absence of debris, the mass balance exhibited a value of -0.48 m
490 w.e. yr^{-1} , whereas with the inclusion of debris, this value decreased to $-0.26 \text{ m w.e. yr}^{-1}$, likely due to the insulating
491 effect of debris on melt rate. In contrast, a similar modeling experiment conducted in the Karakoram found that the
492 Baltoro Glacier experienced a reduction in ablation by approximately 35% when debris was excluded (Groos et al.,
493 2017). Moreover, glaciers in the Central Karakoram National Park, Pakistan, showed a 24% decrease in modeled
494 ablation when debris was excluded (Minora et al., 2015). It's important to note that these contrasting findings with
495 respect to the impact of debris cover on glacier mass balance in the Karakoram can be attributed to differences in
496 the models employed, their configurations, and the thickness distribution of debris cover. The latter directly impacts
497 the spatial characteristics of sub-debris melting intensity (Compagno et al., 2022).

498 On a daily or monthly basis, the impact of supraglacial debris on the Batura Glacier manifested most
499 prominently during the ablation season, as depicted in Figure 8a and b. On an interannual scale, supraglacial debris
500 had a significant impact on mass balance of the Batura Glacier; however, it did not induce alterations in its overall
501 temporal fluctuations or trends (Figure 8c). This was mainly because the simulation process did not include the
502 influence of changes in the debris cover distribution over time on mass balance.

503 The debris had a significant protective effect, effectively mitigating glacier ablation. This effect was most
504 pronounced in August, a period characterized by high air temperatures. During May and June an extensive snow
505 cover blanketed the Batura Glacier. When supraglacial debris is included in energy balance processes, the snow
506 layer absorbed a greater amount of heat from the atmosphere through thermal conduction, thereby leading to
507 accelerated melting. As the snow progressively melted and the debris became exposed, the surface albedo
508 experienced a rapid decline spanning from July to October. This transition resulted in the debris absorbing a greater
509 portion of incoming shortwave radiation, much of which is returned to the atmosphere as emitted longwave radiation
510 or sensible heat, consequently yielding a reduction in the melting energy available (Figure 8b). Statistical analysis
511 revealed that when supraglacial debris was not considered, the average net radiation decreased by 14 W/m^2 . The
512 most substantial reduction was observed in May, with a reduction of approximately 20 W/m^2 .

513

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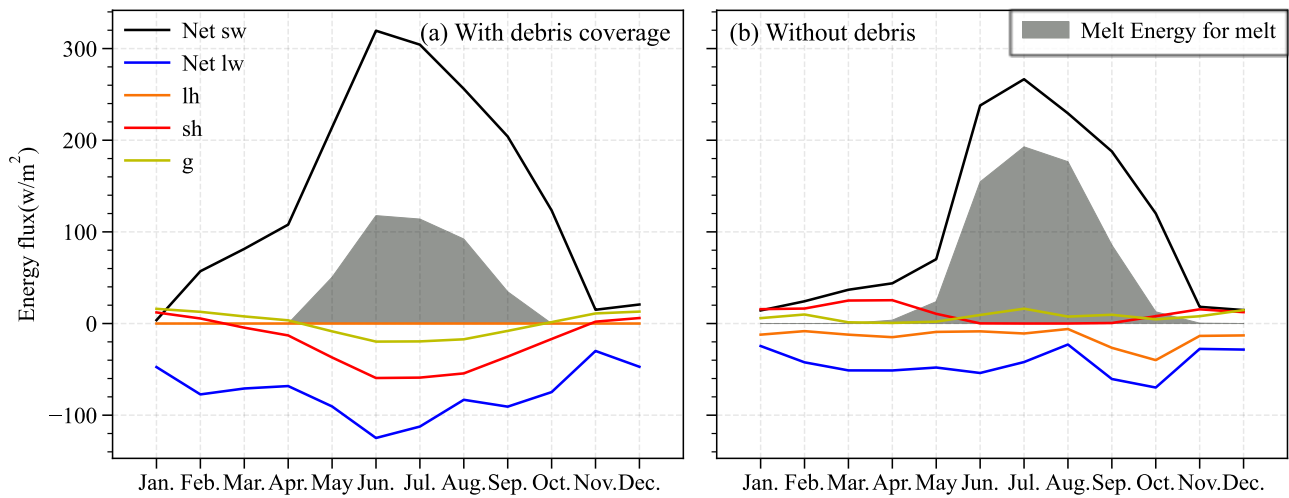


515
 516 **Figure 8** The difference between modeled mass balance with (blue lines and bars) and without debris cover (red
 517 lines and bars): (a) daily mass balance; (b) monthly mass balance; and (c) annual mass balance trend.

518 4.4 The energy controls of sub-debris melt

519 We conducted additional investigations to understand how the supraglacial debris affect the ice ablation. In the
 520 case of the Batura Glacier, the presence of supraglacial debris reduces the average albedo of the glacier, thereby
 521 increasing net shortwave radiation. Notwithstanding the observed augmentation in net radiation, an attenuation in
 522 melt was recorded. To investigate the impact of debris on energy-driven melting, this study conducted a statistical
 523 analysis of the energy balance for scenarios with and without debris coverage in the specific area characterized by
 524 the presence of debris (Figure 9). The results indicated that while the presence of debris did amplify the net radiation
 525 income, the available energy for melting is reduced by the sum of longwave radiation emission, sensible heat, and

526 thermal conduction within the debris (an average decrease of 25 W/m²).



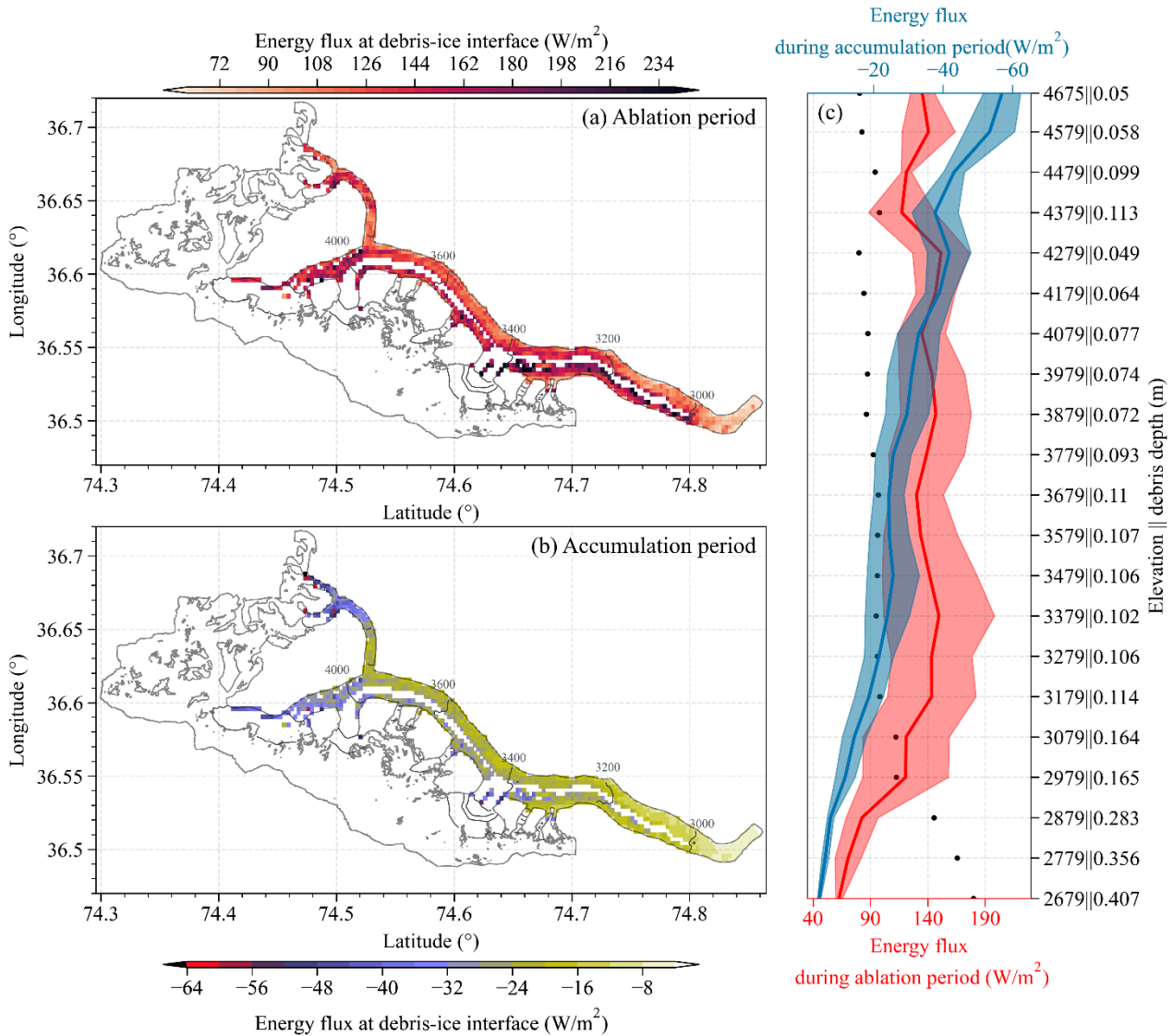
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528 **Figure 9** Annual cycles of energy budget components (a) with and (b) without debris coverage for the currently
529 debris-covered area on Batura Glacier.

530 During the ablation season (June to September), when accounting for the presence of debris, the glacier's
531 energy income, represented by net shortwave radiation, witnessed an augmentation of 61 W/m². Meanwhile, the
532 energy output increased by 116 W/m², comprising net longwave radiation (50 W/m²), sensible heat (42 W/m²), and
533 conductive heat (24 W/m²). Consequently, this led to a reduction of 45 W/m² in latent heat of melt (sublimation heat
534 of the debris layer, which was not considered when deducting the 11 W/m² for sublimation heat without debris cover)
535 (Figure 9). In light of these observations, it can be concluded that the influence of debris cover on glacier melt is
536 twofold. Firstly, it reverses the net direction of turbulent heat fluxes at the glacier surface. Secondly, it alters the
537 heat flux reaching the glacier through thermal conduction. The former aspect primarily emanates from the heating
538 of the debris layer due to shortwave radiation, causing the debris temperature to surpass the atmospheric temperature.
539 Consequently, the glacier transfers heat to the atmosphere, effectively acting as an energy source. This finding aligns
540 with earlier research results, as exemplified by Steiner et al. (2018) and Nicholson and Stiperski (2020). Regarding
541 the second aspect, we conducted an analysis that considered the thermal conduction occurring within both the debris
542 and ice layer, as well as the energy equilibrium within each layer. When the heat gained from net radiation was
543 conducted within the debris layers (the radiation penetration of the debris was neglected), it could be consumed to
544 heat the debris, thereby satisfying the energy balance within and between the debris layers.

545 At the interface between debris and ice, heat exchange exhibits pronounced seasonal variations, with notable
546 altitudinal gradients, particularly during the accumulation period (Figure 10). In the ablation season, a debris layer
547 is very quickly warmed by solar radiation before cooling back close to zero after sunset. The temperature of surface

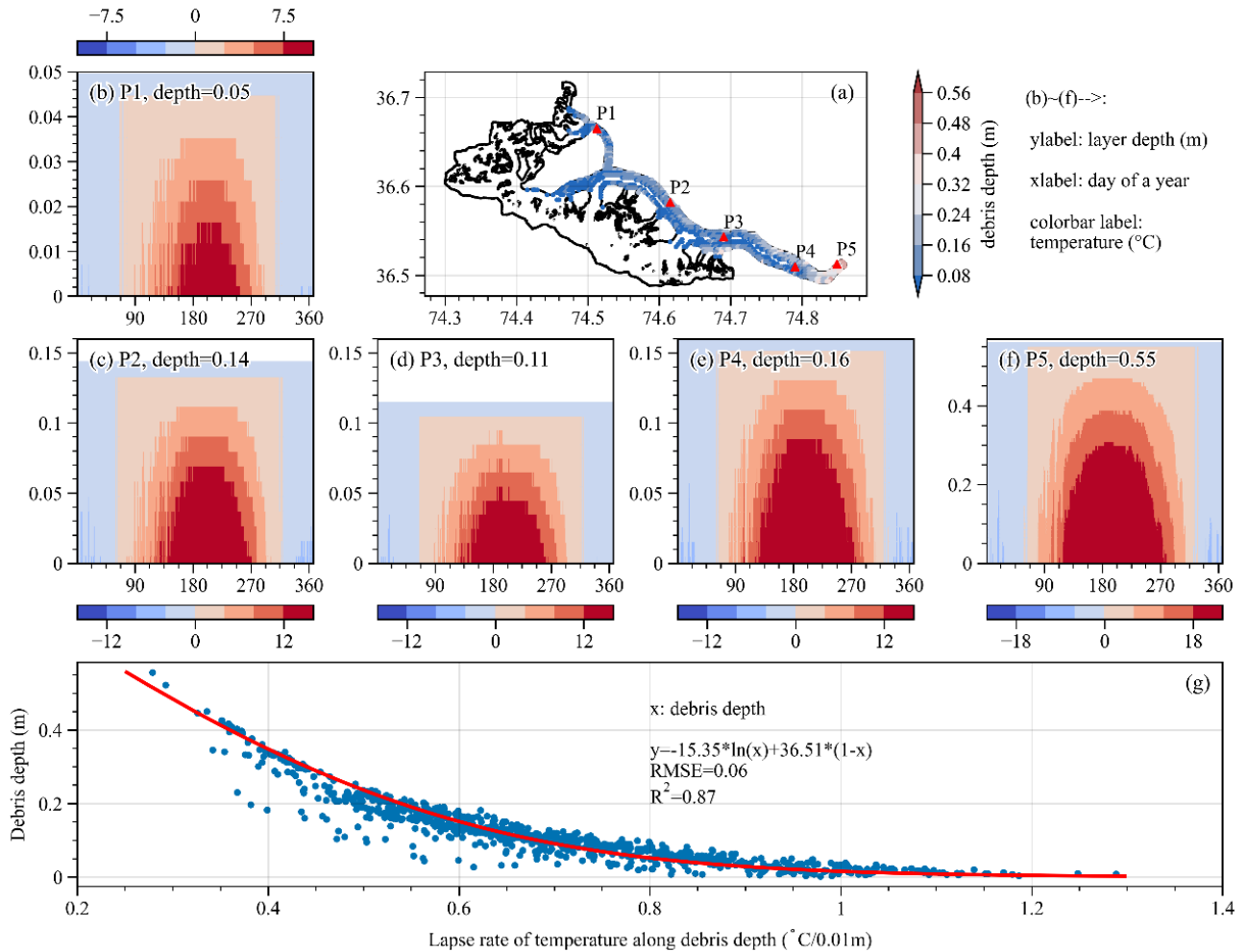
548 debris rises, transferring heat into the interior of the debris (Reid et al., 2012). However, the energy reaching the
549 debris-ice interface is predominantly influenced by the thickness of the debris layer. Below 2900 m, where the debris
550 thickness exceeds 20 cm, the energy at the debris-ice interface is less than 90 W/m². At altitudes above 3200 m
551 where the debris thickness is less than 11 cm, the energy at the debris-ice interface increases to 140 W/m² (Figure
552 10). At these altitudes the debris thickness remains relatively constant, and correspondingly, the energy flux at the
553 debris-ice interface exhibits minor fluctuations. Despite Collier et al. (2015)'s suggestion that near-surface air
554 temperature is generally a stronger driver of melt rates below debris, our findings from the energy at the debris-ice
555 interface, in conjunction with Figure S6, imply that this relationship may not hold true during the ablation season
556 in high-altitude regions. During the accumulation season, the energy at the debris-ice interface is negative, with the
557 glacier transferring heat to the debris layer. This significantly affects the upwelling longwave radiation and sensible
558 heat flux at the debris surface. Thinner debris layers result in more heat transfer from the glacier to the debris (Figure
559 10b). In contrast to the ablation period, the energy at the debris-ice interface steadily increases with altitude during
560 the accumulation season. This difference may be attributed to snowfall causing substantial variations in the surface
561 energy balance process during the accumulation period compared to the ablation season. Overall, altitudes below
562 2900 m are identified as the less sensitive zone for Batura Glacier's ablation. Conversely, the areas where debris
563 cover and bare ice intersect emerge as highly sensitive zones for melting, with the average thickness of debris in
564 these regions being less than 2.3 cm.



565
 566 **Figure 10** Spatial distribution of the mean energy flux at the debris-ice interface during ablation (a) and
 567 accumulation (b) periods. An elevation-dependent distribution of the debris-ice energy flux in each season is
 568 shown in (c).

569 The process of heat conduction within the debris was clearly illustrated in our study through an analysis of
 570 temperature changes within debris of varying thicknesses (Figure 11). During the ablation season, for thinner debris
 571 (Figure 11b, location P1), achieving a stable ice surface at zero °C necessitates a temperature difference of 2.5°C
 572 within the uppermost 0.015 m (comprising 3 layers), with an average temperature decrease of 1.7°C per 0.01 m
 573 increment. Conversely, in the case of thicker debris (Figure 11f), with a depth of 0.2 m (20 layers), the temperature
 574 alteration amounts to 8°C, accompanied by a vertical temperature gradient of 0.4°C per 0.01 m increment.
 575 Consequently, with respect to the upper layers, thin debris is more likely to conduct a greater amount of heat. At the
 576 interface between the surface ice and overlying supraglacial debris, the temperature difference at P1 (0.035-0.045

577 m) was 2.5 °C with a vertical gradient of 2.5 °C/0.01m. At P5 (0.42-0.55 m), the vertical gradient of
 578 was 0.61 °C/0.01m (Figure 11). This indicates that in areas covered by thin supraglacial debris, more energy was
 579 transferred from the debris to the glacier, resulting in a greater amount of latent heat being released by the glacier.



580
 581 **Figure 11** Temporal variations of debris temperature across different depths throughout a year. Temperature
 582 profiles at specific points in (a) are displayed in (b)~(f). The relationship between temperature lapse rate and
 583 debris depth is presented in (g).

584 When the thickness of the debris is comparable, the vertical temperature gradient within the debris exhibits a
 585 corresponding similarity (P2, P4), except for slight deviations primarily observed at the surface. These variations
 586 are primarily attributed to discrepancies in both air temperature and surface temperature of the debris between the
 587 two points. Throughout the accumulation period, net shortwave radiation remained limited, leading to low
 588 temperatures and causing the debris temperature to either reach or drop below freezing point. As a result, the rate
 589 of heat conduction process decelerated, thereby mitigating the influence of the debris on glacier melting.

590 To quantify the relationship between the thickness (x) of the debris layer and the vertical temperature gradient

591 (y), we computed the average temperature gradient for individual pixels within the debris-covered area during the
592 ablation period and conducted regression analysis (Figure 10g). According to Eq. 8, an increase in debris layer
593 thickness corresponds to a reduction in the vertical temperature gradient. Combined with Eq. 4 & 5, the heat
594 conduction to the interface between the debris layer and the glacier will also decrease, leading to diminished
595 availability of latent heat for glacier melting. As the thickness of the debris layer approaches minimal values, the
596 heat originating from a temperature difference of approximately 20°C is used for melting. This fundamentally
597 quantifies the impact of debris cover thickness on melt and further explains the differences in mass balance shown
598 in Figure S5.

$$599 \quad y = -15.35\ln(x) + 36.5(1 - x) \quad (8)$$

600

601 4.5 The potential uncertainties and limitations

602 The parameter settings significantly influence simulation results. Of the six calibration parameters, the
603 simulation results are highly sensitive to firn albedo, ice roughness length, and debris albedo (Figure S1 and S3).
604 The largest changes are observed when varying the debris albedo. When the debris albedo decreases to 0.1
605 (approximately a 23% change in albedo from the calibrated value), the melt increases by about 3.4%. With a 100%
606 increase in debris albedo (0.26), the melt decreases by approximately 14%. This magnitude of sensitivity is
607 consistent with the findings of Giese et al. (2020) on Changri Nup Glacier in the Himalayas. The calibrated
608 parameters ice and firn roughness length lie on the margin of the range, implying that a larger range may be
609 beneficial or that a parameter not considered in calibration is not chosen optimally. However, extending the limits
610 of these parameters would result in physically unrealistic values. Due to the complexity of the model, we did not
611 calibrate all parameters. Instead, we identified the aforementioned six parameters through sensitivity analysis.
612 Besides the calibrated parameters, certain factors, such as the rain and snow separation threshold influence the
613 simulated mass balance. In this study, we constrained these parameters using geodetic mass balance.

614 Apart from the model-inherent parameters, the model's input dataset presents considerable challenges during
615 calibration and introduces uncertainty into the results (Arndt and Schneider, 2023). While HAR data has been
616 applied in glacier mass balance simulation studies (e.g., Huintjes et al. (2015b) and Groos et al. (2017)), its
617 applicability in the Karakoram mountains remains uncertain (Groos et al., 2017) due to the majority of ground
618 validation being conducted on the Tibetan Plateau (Maussion et al., 2014). Additionally, uncertainties can also be
619 introduced by the calibration methods and downscaling schemes of the climatic factors, as evident from the

620 comparison of our study with results from Groos et al. (2017). Initially, Groos et al. (2017) downscaled HAR Version
621 1 data to 30 m resolution using interpolation for glacier mass balance simulations in the Karakoram. In this study,
622 we first calibrated temperature and precipitation in HAR Version 2 using station observations and then employed
623 statistical downscaling to achieve a 300m resolution for energy balance research, incorporating radiative
624 downscaling that accounts for complex topography. While both results of Groos et al. (2017) and this study compare
625 well with station observations, discrepancies exist in temperature and precipitation on Batura Glacier. For example,
626 Groos et al. (2017) reported a temperature of 5.0 °C during the ablation season at ~4,060 m a.s.l., while this study
627 recorded 1.7°C at the same elevation. Annual precipitation for Batura Glacier is ~960 mm in this study compared
628 to 1059 mm in Groos et al. (2017). These differences resulted in significant spatial disparities between the two
629 simulated results (Figure 5a of this study and Figure 6 of Groos et al. (2017)). Although the multi-year average mass
630 balance in this study aligns more closely with geodetic mass balance compared with that of Groos et al. (2017), it
631 remains challenging to determine which result can better capture the spatial characteristics of glacier mass balance
632 due to a lack of knowledge about meteorological conditions in high-altitude glacierized regions and insufficient
633 characterization of surface features like ice cliffs and supraglacial ponds in both models. Therefore, as highlighted
634 by Collier et al. (2013), this uncertainty can only be minimized through additional high-altitude observations and
635 more reliable downscaling approaches, such as dynamic downscaling.

636 The spatial resolution of a glacier model can impact simulation results, particularly in debris-covered areas. To
637 investigate this effect, we conducted comparative simulations with varying resolutions on a small section of the
638 Batura Glacier terminus. We used the 300 m resolution simulation from this study as the benchmark. When
639 increasing the resolution to 100 m (matching the debris data resolution), the average debris thickness showed a
640 minimal difference of 0.01 m compared to the 300 m resolution thickness. However, the spatial distribution of debris
641 thickness exhibited significant discrepancies, especially at the glacier margins (Figure S7a, b). Notably, subsurface
642 melt rate decreased by 2.2% compared to the benchmark (Figure S7e). Since debris albedo was set as a constant
643 value, net radiation remained relatively unchanged. However, the surface temperature decreased by 0.17°C (Figure
644 S7f), accompanied by a 1.9% reduction in sensible heat flux (Figure S7i) and a 2.7% decrease in conductive heat
645 transfer within the debris layer (Figure S7j). These findings demonstrate that while spatial resolution influences the
646 energy fluxes and ablation of debris-covered glaciers, its primary impact lies in the spatial distribution (Figure S7c,
647 d) with minimal effect on average values. This spatial variation primarily stems from the differences in debris
648 thickness captured at varying resolutions. Given the limitations of the employed debris thickness data (Rounce et

649 al., 2021), we cannot definitively conclude if higher resolution simulations yield results closer to reality.
650 Additionally, the computational cost of high-resolution simulations is substantial. Therefore, this study utilized a
651 coarser grid to capture the overall energy and mass balance characteristics of the glacier. However, the potential for
652 more realistic outcomes with reliable high-resolution debris thickness data is undeniable.

653 The main limitation of the model lies in the absence of parameterization for the impact of glacier surface
654 features on melting, such as ice cliffs and supraglacial ponds. This omission may lead to an underestimation of the
655 ice melt rate across debris-covered areas, as observed amplifying effects of supraglacial lakes and ice cliffs on
656 glacial melt (e.g., Tedesco et al. (2012), Miles et al. (2016), and Buri et al. (2021)) are not considered. Supraglacial
657 ponds and lakes efficiently transfer heat into glacier ice due to their low surface albedo and active convection.
658 Simulations by Miles et al. (2018) indicated that ponds may contribute to 1/8 of total ice loss in the Langtang Valley,
659 Nepal. Modeling by Huo et al. (2021a) also suggested a substantial increase in ice loss on the Baltoro Glacier in the
660 Karakoram due to the intervention of supraglacial ponds. Supraglacial ice cliffs influence glacier ice melt by creating
661 a direct ice-atmosphere interface with low albedo and exposure to high emissions of longwave radiation from
662 surrounding debris-covered surfaces (Buri et al., 2016). According to Buri et al. (2021), neglecting ice cliffs in
663 Langtang Valley would result in a mass loss underestimation of $17\% \pm 4\%$ for debris-covered glacier tongues. In
664 most glaciers, interactions generally exist between ice cliffs and ponds/lakes (Buri et al., 2021; Huo et al., 2021a).
665 Therefore, future research should incorporate parameterization for these elements to better understand their impact
666 on glacier melting. However, in the absence of sufficient observations, a limited representation of ponds and ice
667 cliffs in the parameterization of model can introduce additional uncertainty in glacier-wide energy fluxes (Miles et
668 al., 2016).

669

670 5 Conclusions and outlook

671 This study presented a comprehensive investigation into the relationships between supraglacial debris cover,
672 energy fluxes, and mass balance dynamics on the Batura Glacier in the Karakoram. Through simulation analysis,
673 we propose that the presence of debris on the glacier surface effectively reduces the amount of latent heat available
674 for ablation by absorbing solar radiation and preventing it from reaching the ice surface, which creates a favorable
675 condition for the Batura Glacier's relatively low negative mass balance. Furthermore, the glacier's mass budget has
676 shown a decreasing trend in (negative) magnitude between 2000 and 2020, primarily due to a reduction in ablation,
677 especially in areas with thin debris cover and debris-free parts of the ablation area, which outweighs the relatively

678 smaller reduction in snowfall accumulation. More detailed findings and outcomes of the study are concluded as
679 follows.

680 (1) The Batura Glacier exhibits substantial spatial heterogeneity in mass balance distribution along its
681 elevation gradient. Altitudinal dependence was influenced by the presence of debris cover, resulting in the
682 most intense melting occurring between 3000 and 3400 m, with a reversal of the ablation gradient below
683 3000 m due to the greater insulation by thicker debris on the lower portion of the glacier.

684 (2) Our simulations revealed that supraglacial debris cover exerted a notable influence on glacier mass balance.
685 Including debris cover in the energy balance model led to a 45% reduction in the magnitude of the negative
686 mass balance of the Batura Glacier (with debris: -0.26 m w.e. yr^{-1} , without debris: -0.48 m w.e. yr^{-1}). This
687 reduction was particularly prominent during the ablation season, highlighting the significance of debris
688 cover in mitigating glacier ablation.

689 (3) The role of debris cover in altering energy exchange was multifaceted. Debris cover enhances net radiation
690 income by reducing albedo but also promotes thermal transfer, which warms the debris and leads to a
691 higher rate of energy transfer to the atmosphere through longwave emission and sensible heat, thereby
692 reducing available melt energy compared with bare ice. This intricate interplay modified the glacier's
693 response to energy budgets, ultimately affecting its mass balance.

694 (4) Our investigation into the effects of debris thickness on temperature gradients within the debris layer
695 reveals a fundamental connection between debris thickness and its influence on melt processes. Thicker
696 debris layers engender reduced temperature gradients, leading to reduced latent heat available for glacier
697 melting.

698 This study significantly advances our understanding of energy and mass interaction on debris-covered glaciers
699 in the Karakoram. However, in addition to the previously discussed impact of ponds and ice-cliffs on ice ablation,
700 future work should also address the evolution of supraglacial debris thickness and glacier dynamics. These factors
701 exert a significant influence on the energy reaching the glacier surface (Compagno et al., 2022; Huo et al., 2021b).
702 Finally, this paper identified that the mass balance of Batura Glacier became less negative in the period 2000-2020,
703 most likely due to a decrease in air temperature over the same period. This result supports wider findings associated
704 with the "Karakoram anomaly" and this phenomenon warrants further discussion and investigation.

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708 **Declaration of competing interest**

709 The contact author has declared that none of the authors has any competing interests.

710

711 **Data/Code availability**

712 HAR dataset is available from Institute of Ecology Chair of Climatology website at https://www.klima.tu-berlin.de/index.php?show=daten_har2&lan=en. Meteorology and ablation observations. Glacier surface elevation difference of Wu et al. (2021) is available upon request from the authors, the elevation difference produced by Hugonnet et al. (2021), Shean et al. (2020), and Brun et al. (2017) are available at <https://doi.org/10.6096/13.>, from National Snow and Ice Data Center (NSIDC) at <https://nsidc.org/data/highmountainasia> and from PANGAEA website at <https://doi.pangaea.de/10.1594/PANGAEA.876545>. The KGI datasets are available from the National Cryosphere Desert Data Center of China at <https://doi.org/10.12072/ncdc.glacier.db2386.2022>. The observations collected by this research are available upon reasonable request from the authors. The COSIPY used in this study is available on GitHub at <https://github.com/cryotools/cosipy>. The code developed for calculating energy and mass balance on supraglacial debris is available upon request from the authors. The coupled model will be publicly available once some technical issues are fixed.

723

724 **Author contribution**

725 Yu Zhu: Conceptualization, methodology, model development, writing original draft, writing review & editing.
726 Shiyin Liu: Conceptualization, supervision, project administration, funding acquisition. Ben W. Brock: Supervision,
727 model development, writing review & editing. Lide Tian: Supervision, project administration. Ying Yi: Validation,
728 formal analysis, writing original draft. Fuming Xie: Investigation, visualization. Donghui Shangguan: Investigation.
729 Yiyuan Shen: Formal analysis, visualization.

730

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739

740 **Appendix A Correction and downscaling of the model Inputs**

741 **A1 Adjusting of precipitation**

742 Numerous research endeavors have elucidated notable biases in precipitation observations within and in the
743 vicinity of the Hunza river basin. For instance, Winiger et al. (2005) discovered a noteworthy discrepancy, with
744 precipitation at altitudes surpassing 5000 m exhibiting sixfold or more intensity compared to lower altitudes, as

745 deduced from station observations. Similarly, Tahir et al. (2011) ascertained a dissimilarity between runoff and
 746 observed precipitation, with Dainyor station recording a runoff of 750 mm/yr but a mere 100 mm/yr of observed
 747 precipitation. This asymmetry was also discerned in the neighboring region (Immerzeel et al., 2009). To make a
 748 more accurate precipitation input for the simulation, we consulted the method proposed by Wortmann et al. (2018)
 749 to rectify the precipitation data. This method entails the calibration of precipitation through the calculation of the
 750 calibration factor $f_c(H)$, as expressed by the following equation:

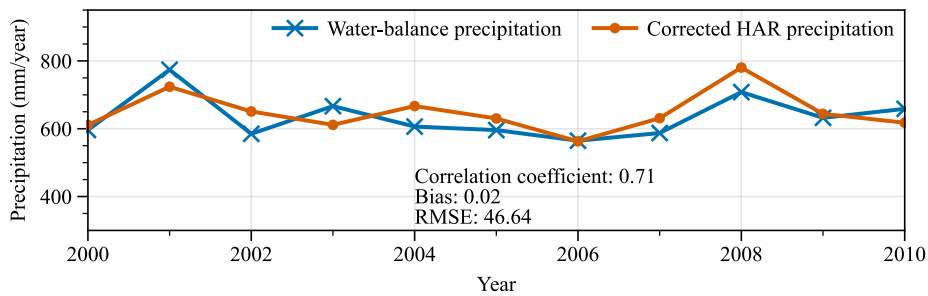
$$751 \quad f_c(H) = (c - 1) \exp \left\{ - \left[\frac{P_{LR}}{(c-1)*100} \right]^2 * (H - H_{max})^2 \right\} + 1 \quad (A1)$$

752 Where c represents the calibration factor, H_{max} represents the maximum elevation at which precipitation
 753 occurs, P_{LR} signifies the elevation correction factor for precipitation. These parameters are determined using the
 754 linear relationship proposed by Immerzeel et al. (2012), and the range of values for the determination is derived
 755 from existing studies. The linear relationship can be expressed as follows:

$$756 \quad \begin{cases} P_T = P_{HAR} * [1 + (H - H_{ref}) * P_{LR} * 0.01] & H_{ref} < H < H_{max} \\ P_T = P_{HAR} * [1 + ((H_{max} - H_{ref}) + (H_{max} - H)) * P_{LR} * 0.01] & H > H_{max} \end{cases} \quad (A2)$$

757 Where H_{ref} denotes the reference elevation, which corresponds to the elevation at which the observed
 758 precipitation closely matches the actual precipitation. P_{HAR} and P_T represent HAR precipitation and calibrated
 759 precipitation. We determined H_{max} and P_{LR} by approximating the calculated P_T based on the water balance
 760 equation (Eq. A3) (Figure A1), with the range of values for H_{max} and P_{LR} referencing the priori studies. In the
 761 Eq.3, ET uses MODIS evapotranspiration products, R takes the runoff from the watershed outlet observation
 762 station (Dainyor station), and TWS takes the average of GLDAS and GRACE solutions.

$$763 \quad P_T - ET - R - TWS = 0 \quad (A3)$$



764
 765 **Figure A1** Comparison between corrected precipitation and precipitation calculated by water balance equation.

766 **A2 Downscaling of the model inputs**

767 In order to achieve the desired level of precision for mass balance simulation on a glacier scale, this study

768 downscaled HAR reanalysis data from 10 km to 300 m by using statistical methods. Special attention was given to
 769 the impacts of topography, slope, and aspect on meteorological factors during this process. The SRTM DEM with a
 770 spatial resolution of approximately 30 meters was utilized to obtain topographic features. In order to effectively
 771 represent topographical features on a glacier scale while maintaining optimal computational efficiency during the
 772 energy balance simulation process, the target grid size was set at 10 times the SRTM DEM (~300 m).

773 Based on water balance at basin outlet, the precipitation was first calibrated using remote sensing data and
 774 station observations to obtain the altitude gradient and maximum precipitation altitude (Supplementary Methods).
 775 After calibration, the altitude gradient of precipitation throughout the Hunza river basin was determined to be
 776 0.18%/m. The maximum precipitation altitude of the Batura glacier was 4900 m. Then, the precipitation was
 777 downscaled at a resolution of 300 m for the Batura glacier by applying the Eq.1 provided in the Supplementary.
 778 Incoming shortwave radiation was downscaled by using the radiative transfer equation (Eq.4) on sloping surfaces.
 779 The details in solving this equation can be found in publication of Ham (2005). The correlation coefficient of
 780 incoming shortwave radiation before and after downscaling is 0.91, with an RMSE of 26, indicating the
 781 parameterization-based downscaling enables a more refined representation of spatial characteristics while
 782 preserving the original characteristics and trends of the data.

$$783 \quad R_{gs} = R_b \left(\frac{\cos(\phi) \cos(i) + \sin(\phi) \sin(i) \cos(\gamma - \alpha)}{\cos(\phi)} \right) + R_d \quad (4)$$

784 In the above equation, R_d represents scattered radiation, which is solved using a modified Gompertz function
 785 that quantifies the relationship between horizontal total radiation (R_{gh}) and clear sky index (CI) (Wohlfahrt et al.,
 786 2016); CI is determined based on radiation duration, while R_{gh} is initialized as R_{gs} ; R_b denotes direct incident
 787 radiation and is calculated by subtracting R_d from R_{gh} ; ϕ and γ represent solar zenith angle and azimuth angle
 788 respectively, which can be obtained using parameterization schemes proposed by Wohlfahrt et al. (2008); i denotes
 789 the angle between the slope and horizontal plane, while α represents the azimuth angle of the slope.

790 Temperature, relative humidity, wind speed, and air pressure were downscaled using altitude gradient obtained from
 791 HAR data. Cloud cover was downscaled refer to the scheme of ERA5 (Muñoz Sabater, 2019). Owing to the absence
 792 of meteorological observations required for computing altitude gradient, the altitude gradient over a broader region
 793 (Karakoram Mountains), which encompasses the study area, were determined using HAR data to minimize errors.
 794 The altitude gradient for 2 m air temperature was calculated to be -0.0054 °C/m, while that for 2 m wind speed was
 795 0.00078 m*s⁻¹/m. The rate for 2 m relative humidity was 0.014 %/m, and that for atmospheric pressure was -0.044
 796 hPa/m.

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