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3	Supplementary Information for
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6	Disappearance of the Last Tropical Glaciers in the Western Pacific Warm
7	Pool (Panua Indonesia) Appears Imminent
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38 **Supplementary Text**

39

40 Determination of changes in ice surface area

Background image (Fig. 3A) was derived from

41 42 http://eoimages.gsfc.nasa.gov/images/imagerecords/79000/79084/grasberg tm5 2009301 43 lrg.jpg. Details of five satellite images used to outline and quantify the surface area of the Puncak Jaya glaciers (Fig. 3B and C) are in Table S1. The 2002 image was extracted 44 45 from Google Maps as a flat image (stereoscopic 3D mode turned off). All images were obtained as orthorectified products. To correct for residual relief displacement not 46 47 removed through orthorectification, IKONOS and WorldView-2/3 images were manually 48 co-registered with respect to Planet's May 2016 image using clearly identifiable 49 topographic features (a similar approach for the same area was used in (1)). For area 50 calculations, the five images were mapped in a Universal Transverse Mercator (UTM) 51 projection (zone 53 S) using the World Geodetic System 1984 (WGS84) datum. All 52 image processing and area calculations were performed using the QGIS software (version 53 3).

54

55 Simulation of glacier volumetric changes

56 Volumetric changes from 2017 to 2030 were simulated with a spatially distributed 57 model that calculates mass balance (in meters water equivalent or m.w.e.) at daily time 58 steps on each grid-cell flagged as ice or snow, according to the Randolph Glacier 59 Inventory (RGI) (2). Ablation (Ab) is computed from daily positive temperatures while 60 accumulation (Ac) utilizes temperature thresholds to determine the amount of solid 61 precipitation falling over the glacierized area:

62

$$Ab = DDF \times T_a \text{ if } T_a \ge 0^{\circ}C$$
 Eq. 1

$$Ab = 0 \text{ if } T_a < 0^{\circ}C$$
 Eq. 2

$$Ac = Pp \quad if \ T_a < T_{sf}$$
 Eq. 3

$$Ac = Pp\left(\frac{T_r - T_a}{T_r - T_{sf}}\right) \text{ if } T_{sf} < T_a < T_r$$
 Eq. 4

$$Ac = 0 \text{ if } T_a \ge T_r$$
 Eq. 5

63

DDF is the Degree Day Factor (m $d^{-1} \circ C^{-1}$), which converts air temperature into 64 ablation; T_{sf} , is the temperature threshold for accumulation while and T_r is the threshold 65 for rain. T_{sf} and T_r were assumed to be 0°C and 2°C. Computation of daily melt utilized 66 67 different DDFs: (a) snow on grid-cells that had accumulation the previous day and 68 otherwise (b) for ice.

69

Modeled elevation changes were converted into volume (in km³) and transformed 70 71 into a ratio relative to an initial ice volume computed from a volume-area scaling (3-4) and estimated as 0.13 km³. This class of semi-empirical model was selected following 72 73 recommendations (5) regarding quality and quantity of available data and the length of

calibration periods for model evaluation. Here, the climatic data available and the
relatively flat topography allowed us to infer that (a) most, if not all, ablation occurs as
melt, and (b) relatively short windows for calibration and validation are available from
1997 to 2016.

78

The ASTERGDEM (6) (30-meter grid-cell size) provided the regional
topographic representation. Time series for precipitation and temperature were derived
from available observations and downscaled global climate model simulations (see
below).

83

84 Mass balance model calibration

85 The model was calibrated by simulating periods with available stake 86 measurements and we sought optimal combinations of lapse rates and DDFs that 87 reproduced the surface thinning. One thousand combinations were tested using lapse rates 88 provided by Fig. S3 and published DDFs (7). The model reproduced observed thinning 89 with a relatively small bias. Optimal parameters were identified using a quasi-zero 90 temperature lapse rate and precipitation lapse rate similar to the mean detected between 91 DISP and ALP (Fig. S3). Average DDF for snow (ice) was fairly consistent (more 92 variable) among the simulated periods. Only the May 2016 – November 2016 simulation 93 showed a logical pattern of higher DDF for snow versus ice, thus the simulated 94 parameters for this period were employed in future change simulations (Table S3). This 95 optimal combination of parameters suggested that temperatures do not change 96 significantly over the ice fields, in sharp contrast with the average lapse rates for the pair 97 DISP-ALP (Fig. S3). To test the sensitivity of the results to other realistic lapse rates, two 98 additional temperature lapse rates were utilized: 4.6°C/km, corresponding to the mean

between GRS-DISP and GRS-ALP, and the environmental lapse rate (6.5°C/km).

100

101 The climatic records for mass balance model input

102 Precipitation and temperature records around the study area are short and 103 discontinuous (Table S2) and were combined to construct a relatively continuous time 104 series for the highest elevation weather station (ALP). Daily lapse rates were computed 105 between each pair of stations to examine the relationship with elevation. Lapse rates for precipitation vary considerably, with some extreme values well above +150 mm/km 106 107 while for temperature they are more concentrated around the mean (close to the 108 environmental lapse rate for the pair DISP-ALP, Fig. S3). ALP reconstruction used a 109 multilinear regression with GRS and DISP as predictors. Linear regressions of the daily values between each pair of records explain 45% to 66% of the variance for precipitation 110 111 and 50% to 66% of the variance for temperature (Fig. S6). The reconstructed ALP record 112 (henceforth, rALP) covers 88% of the period 1999-2016 for both variables, a 113 significantly longer time series than the original ALP record (< 75% for precipitation and 114 < than 60% for temperature), therefore maximizing the number of calibration periods 115 utilized for the modeling. From the remaining gaps (12% for 1999-2016), 0.5% and 0.7% are isolated records of precipitation and temperature, respectively; these records were 116 117 filled by linear interpolation using values from the previous and subsequent days. Fig. S7 118 shows that ALP and rALP time series are significantly correlated and the Nash–Sutcliffe 119 model efficiency coefficient is close to 1 for both variables. Simulation of future volume

change was performed for the period 2017-2030 using input from climate modelprojections, described in the next section.

121

123 **Downscaling of global climate projections for the region**

Southeast Asia Regional Climate Downscaling (SEACLID) / Coordinated 124 125 Regional Climate Downscaling Experiment-Southeast Asia (CORDEX-SEA) project (8-126 11) (http://www.ukm.edu.my/seaclid-cordex, http://www.apn-127 gcr.org/resources/items/show/1886) provides output from regional atmospheric model 128 simulations over Southeast Asia that were forced by output from global climate model 129 experiments conducted by the Coupled Model Intercomparison Project Phase 5 [CMIP5 130 (12)]. The horizontal grid spacing of the regional model output is 25 km. The data were 131 downloaded from the Earth System Grid node of Linköping University (https://esg-132 dn1.nsc.liu.se/search/esgf-liu/) and derived from CORDEX-SEA dataset (as in (11)). 133 Daily estimates of precipitation and 2-meter temperature from five models and for three 134 CMIP5 experiments (historical, RCP4.5, and RCP8.5) (Table S4) were used. Time series 135 for the Puncak Jaya ice fields were obtained from bilinear interpolation of the gridded 136 model output, and then downscaled using the cumulative distribution function (CDF)-Transform (CDFt) method (13). The 1999-2005 period was used for calibration while the 137 2006-2016 period was used for validation. The bounds of the calibration period were 138 139 dictated by the beginning of the rALP record (1999) and the end of the CMIP5/CORDEX 140 historical runs (2005). Validation included comparison between the 2006-2016 141 downscaled time series (RCP4.5 and RCP8.5) with rALP. Following (13), the skill of the 142 models was assessed using the Cramér-von Mises test with $\alpha = 0.05$, in which a p-value 143 greater than this significance level means that downscaled and observed time series are 144 not statistically different. The CSIROMK36-REGCM4 model exhibits the best overall 145 skill for both RCP scenarios (highest *p*-values). All other downscaled time series were less consistent with observations (see Table S4 for details). The model predicts that under 146 147 practically all scenarios glacier shrinkage will continue unabated, leading to total ice loss 148 no later than 2026 (Fig. S8). It is important to remember that the model had been tuned 149 during a period of accelerated thinning (Table S3) based on the available meteorological 150 data (GRS, DISP and ALP stations) in the highland, which in itself infers a somewhat 151 anomalously high amount of energy on the ice surface. As a tunable value given the scarcity of data, the rate of 0.1°C/km indicates that the melting conditions were more or 152 153 less similar across the study area. After considering the data for the model, the elevation 154 of 4366 m was used for the pole site, which is just within 4 meters of the nominal 155 elevation of the ALP station (~4400 m). That also supports the low lapse rate, because 156 the elevation is practically the same. Even with a more accurate ALP elevation, the result 157 $(\sim +/-30 \text{ m})$ would not be significantly different. The hypsometric distributions 158 demonstrate that the elevation range of the glacierized region is ~400 m (except 159 ASTERGDEM, where less than 10% of the area is above that range, Fig S4). Therefore, 160 it is essentially flat and thus justifies a low lapse rate as well.

161

Figure S8 shows somewhat erratic performances of IPSL and especially MPI. Different runs of RegCM4 performed by CORDEX-SEA members were statistically downscaled and forced, in turn, by different model output from GCMs (see Table S4).

165 After performing the statistical downscaling, the bias of both combinations, IPSL-

166 RegCM4 and MPI-RegCM4, relative to observations, was observed during the validation 167 period (2006-2016). IPSL-RegCM4 shows a negative bias (colder than observations) in 168 the RCP8.5 while there is a positive bias in RCP4.5 (warmer than observations). This 169 would explain less melting in the RCP8.5 scenario. For MPI-RegCM4, the bias is 170 noticeable in precipitation, with RCP8.5 being almost 5 five times larger than RCP4.5, 171 and thus during the simulation any temperature below freezing implies much more 172 accumulation in RCP8.5. All other models have consistent biases in both scenarios and 173 thus provide a more coherent behavior of glacier mass balance. One of the co-authors 174 (Alfonso Fernández) has been working over the last year with data from the RegCM4 175 model (the same model utilized by CORDEX-SEA members to dynamically downscale 176 the global models to the versions used in the paper) for downscaling MPI for Chile (i.e., 177 the same combination MPI-RegCM4 from Table S4). Inconsistency in precipitation was 178 found which is similar to the Puncak Jaya case where the RCP8.5 (more pessimistic 179 scenario) produces more precipitation than RCP4.5. In Puncak Jaya case, RCP8.5 shows 180 a larger positive bias (2.72 mm/day) relative to RCP4.5 (0.6 mm/day). In both cases 181 (Puncak Jaya and Chile), this is counterintuitive as drier conditions should be expected in 182 the RCP8.5 as most models predict, despite large interannual variability. This problem is 183 so important to consider that the institute producing the MPI-RegCM4 simulations for 184 Chile had to add a disclaimer message on their website (http://simulaciones.cr2.cl/). 185 These two coincidental results with the combination MPI-RegCM4 suggest that one of 186 the models (MPI or RegCM4) might be generating some biases.

187 188

Ice core sample analysis

189 Ice cores were cut into discrete samples, melted, and analyzed for oxygen and hydrogen isotope ratios (δ^{18} O, δ D, respectively), and insoluble dust and major anion and 190 191 cation concentrations. All measurements were conducted in BPCRC laboratories. δ^{18} O 192 and δD were measured using a Thermo Finnigan mass spectrometer and a Picarro cavity 193 ring-down spectroscopy analyzer. Analyses of identical samples using both machines 194 agreed well (14). The concentrations of insoluble dust particles from 0.63 to 20 μ m in 195 diameter were measured using a Beckman-Coulter Multisizer 4. The concentrations of the major cations $(NH_4^+, Na^+, K^+, Mg^{2+}, \text{ and } Ca^{2+})$ and anions $(SO_4^{2-}, NO_3^-, F^-, \text{ and } Cl^-)$ 196 197 were determined using a Dionex ICS-3000 ion chromatograph. Descriptions of the 198 sample handling procedures and data collection are in (14).

199

Core D1 was cut into three sets of 1156 co-registered samples (each ~2.8 cm long) which were analyzed for stable isotopes, insoluble dust, and major ions. Tritium (³H) concentrations for 68 samples (Core D1) were measured at the Division of Climate and Environmental Physics, Physics Institute, University of Bern, Switzerland to locate the 1960s thermonuclear bomb test horizons. Core D1B contains 1607 co-registered samples (each ~2 cm long) which were analyzed for stable isotopes and dust. Major ions were analyzed for 644 samples (each ~5 cm long).

207

208 <u>Reproducibility of δ^{18} O, δ D, and dust concentrations between Cores D1 and D1B</u>

209 The δ^{18} O and δ D profiles of Cores D1 and D1B show similar variations, with 210 moderate stable isotopic variability (5 to 6‰ and 45 to 50‰, respectively) (Fig. 4*A* and 211 *B*). Their δ^{18} O averages (-16.07 to -16.41‰, respectively; Table S5) are similar to the 1-

meter δ^{18} O averages from the 1972 Meren and Cartensz Glacier cores (-15.7‰) (15) 212 213 before surface melting was a major issue and arguing that the climatic records are likely 214 well preserved in both cores. Large isotopic depletions occur at depths of 8, 12, 17 and 26 215 meters. Stable isotopic ratios in both cores show gradual but significant enrichment in the top eight meters and significant smoothing in the top four meters. Both cores are 216 217 composed entirely of ice, with no observable firn and no vertically elongated, or 218 otherwise deformed, bubbles. The high-resolution insoluble dust record from Core D1 219 illustrates prominent events from 20 to 29 m depth and in the top eight meters (Fig. 4C), 220 as do the records of major ion concentrations (14). The considerable isotopic smoothing 221 in the upper part of the cores and very low aerosol concentrations below 4 meters suggest 222 that substantial post-depositional alteration has occurred in this upper section.

222

224 Post-depositional processes (e.g., evaporation, melting, and rainfall) may alter the 225 climate record in an ice core (16). However, these records indicate that the impact of 226 these processes is limited below ~4 m depth, and suggest that the climatic and 227 environmental records are at least partially preserved. This is similar to the climate record 228 from the Furtwängler Glacier on Kilimanjaro. During the drilling of the Furtwängler 229 Glacier water was encountered throughout the borehole and most soluble ions in the ice 230 column were removed, most likely by post-depositional melting (17). However, the climate record from δ^{18} O had remained intact if rather muted compared with the records 231 from the larger ice fields on Kilimanjaro. Additional support for the viability of the 232 233 record in the ENF cores comes from the lack of sharp bomb peaks in the pre-1964 tritium 234 measurements in Core D1 (Fig. 5A) which argues against post-depositional melting and 235 refreezing at lower depths (18). Core D1 is the cleanest tropical ice core yet recovered by 236 BPCRC, even more so than the Quelccaya (19) and the Kilimanjaro cores (17). This 237 likely reflects the surrounding oceans which provide few sources of dust and soluble 238 aerosols (14), very heavy annual rainfall, and possible elution of ions and dust due to 239 post-depositional processes.

240

241 <u>Timescale reconstruction – tritium (³H) analysis</u>

In the mid-1950s and early 1960s, ³H was widely dispersed during above-ground 242 atomic bomb testing. The quantity of 3 H in the atmosphere peaked in 1962 – 1963 and 243 244 has decreased thereafter. The 1962/63 peak is routinely used as a global chronological 245 marker for dating ice cores. In New Guinea Island, there were two GNIP IAEA/WMO 246 stations which recorded ³H: Jayapura (2.53°S; 140.72°E; 3 m.a.s.l), ~435 km northeast of 247 the drill sites (1957 – 1991) and Madang (5.22°S; 145.80°E; 4 m.a.s.l), ~950 km east of the drill sites (1968 - 1982) (Fig. S9A). Jayapura station recorded high ³H concentrations 248 249 (~33 TU) in precipitation in 1964 (Fig. S9B) suggesting the glaciers near Puncak Java 250 likely preserve this signal. Sixty-eight Core D1 samples (each ~0.47 m long) were 251 analyzed for ³H concentrations and reveal a peak (2.98 ± 0.42 TU) recorded at 23.4 m (Fig. 5A). Since tritium has a half-life of 12.3 years, after 47 years (from 1964 to 2010) 252 253 the tritium concentration in precipitation in 1964 would become $(33 \text{ TU}) \wedge (1/(47/12.3))$ 254 ≈ 2.49 TU which is comparable with the tritium peak recorded in Core D1 at a depth of 23.4 m (Fig. S9C). The differences in background concentrations below 26 m and above 255 256 22 m strongly suggest that the bomb tests elevated the natural ³H background level thus 257 providing an absolute time marker of 1964 at 23.4 m in Core D1.

- 259 Timescale reconstruction – δ^{18} O reference matching with NINO3 ERSST 260 The annual means of δ^{18} O of precipitation (thermal year which covers the period 261 of August of previous year to July of the current year) from the GNIP station at Javapura 262 263 (1961-1991) were compared with NINO3 ERSST (http://www.esrl.noaa.gov/psd/data/gridded/data.noaa.ersst.v4.html) (Fig. S10). The 264 precipitation δ^{18} O at Javapura is only available from 1961 to 1991. Peak δ^{18} O 265 enrichments recorded in precipitation from 1972-1973 and 1982-1984 at Jayapura are 266 267 likely associated with strong El Niño events in 1972-73 and 1982-83, respectively (14). The correlation field between the annual rainfall δ^{18} O at Jayapura and ERSST from 1961 268 269 to 1991 (31 years) is shown in Fig. S10A. By implementing the non-parametric approach 270 for data that are serially correlated (20), the annual rainfall δ^{18} O at Javapura has relatively 271 significant positive correlation with NINO3 ERSST (Area A in Fig S10A) (R = 0.31; p =272 0.09) and average SSTs over the region of 5-20°S and 120-90°W (Area B in Fig S10A) (R = 0.36; p = 0.05) in the eastern Pacific. The annual rainfall δ^{18} O at Jayapura is also 273 274 negatively correlated with average SSTs in southeast New Guinea (Area C in Fig S10A, 6-18°S; 140-160°E). This indicates that there is a link between rainfall δ^{18} O at Javapura 275 and the ENSO indices, with increased rainfall δ^{18} O during El Niño conditions, and more 276 277 depleted δ^{18} O during normal and La Niña conditions. This relationship was supported by 278 previous studies in Papua (21) and lowland Borneo (22). Thus, for simplicity, the NINO3 279 ERSST time series was used as a matching reference to construct the timescale for the Papua ice cores. The δ^{18} O data at fourteen points in Core D1 were paired with the 13-280 281 month running means of NINO3 ERSST by assigning the ³H peak at 23.4 meters depth as 282 the 1964 bomb horizon (Fig. 5A), and assuming the top layer represents the time when 283 the ice cores were collected (May 2010) (Fig. 5B). Along with these two points, the 284 twelve points consisted of seven El Niño events (2002/2003, 1997/1998, 1991/1992, 285 1987/1988, 1982/1983, 1972/1973 and 1965/1966) and five La Niña events (2007/2008, 286 1999/2000, 1995/1996, 1988/1989 and 1975/1976) (Table S6). Each time series was 287 developed using linear interpolation between matching points. Annually resolved stable 288 isotope, dust, and major ion records since 1964 were calculated by averaging values 289 within each year. The depth-age relationships for the Papua ice cores are shown in Fig. 290 S11. The timescale was independently checked by comparing the Core D1 tritium record 291 with tritium from precipitation collected at GNIP stations in Jayapura in Indonesia and 292 Madang in Papua New Guinea (Fig. 6). 293
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Figure S1 | Thinning of the East Northwall Firn from 2010 to 2016. The increasing exposure of the upper sections of an articulated PVC stake placed in a borehole in June 2010 (A) is shown in November 2015 (B), in May 2016 (C), and in November 2016 (D). The exposed two-meter PVC sections shown in (B), (C) and (D) are numbered. Note that the photographs of the stake were taken from different angles by Dave Christensen and Greg Chimura (A), Yohanes Kaize (B, C) and Donaldi Permana (D).





307 Figure S2 | Distribution of available temperature records from 1997 to 2016 (rALP -308 reconstructed ALP is included for reference; see Supplementary Text, Table S2 and Figs. 309 S6 to S7). They show no daily-averaged temperatures below freezing through this period. 310 A bootstrapped Kolmogorov-Smirnov test (1000 iterations sampling 90% of the data) 311 applied to each time series indicates that they can be modeled by a normal distribution 312 with a significance level of 0.01. The probabilities (%) of below freezing temperatures calculated from the corresponding normal distribution are 3.7×10^{-9} for GRS, 2.6×10^{-7} for 313 314 DISP, and 0.005 for ALP. 315





Figure S3 | Distribution of daily lapse rates for precipitation and temperature for

319 each pair of stations located in the study area. Positive values indicate a decrease in 320 the variable according to elevation while negative values indicate an increase.



327 Figure S4 | Hypsometric distribution of the ice covered area in Puncak Jaya

328 according to the available topographic data for the region. Each curve presents the

329 freezing line (dashed blue) as calculated from the average temperature at the ALP station

- 330 (3.36°C) and using two lapse rates. Hypsometry from ASTERGDEM (6) is the only
- source of data showing that more than 90% of the remaining glacier surface is below 122
- these freezing lines. Hypsometry derived from SRTM (23) and ALOS PALSAR (24)
 suggests that the remaining glacierized area is persistently affected by melting
- suggests that the remaining glacierized area is persistently affected by meltintemperatures.
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Figure S5 | Diurnal temperature cycle at the ALP station (4,400 masl). Applying
 different lapse rates (4.6 and 6.5°C/km) to this cycle suggests that melting temperatures

occur on most of the glacierized area while below-freezing conditions may predominate
at elevations above 4,800 m for 4 to 6 hours in the early morning, particularly from July

- to September.
- 349
- 350



Figure S6 | Linear regression models between each pair of stations at daily
 resolution. Top panels correspond to models for precipitation while bottom panels
 correspond to models for temperature.



360 Figure S7 | Comparison between observed (ALP) and modeled (rALP) variables.



361 362

363 Figure S8 | Results of the mass balance simulations using all combinations of the

364 lapse rates, RCPs, and model projections (see Supplementary Text, Tables S3 and S4).

365 The code of each panel indicates the lapse rate (e.g., LR01 is for a lapse rate of

366 0.1°C/km) and the RCP scenario utilized in the respective simulation (e.g., RCP45 is

367 RCP4.5). Results predict that across all scenarios glaciers in Puncak Jaya will disappear

- 368 by 2026 with one exception. The exception is the MPI-M-MPI-ESM-MR, which predicts
- that 60% of the ice volume will remain after 2030. Model descriptions can be found in (11)
- 370 (11).



Figure S9 | Tritium concentrations in precipitation at GNIP stations in New Guinea
Island. (A) The map of GNIP IAEA/WMO stations at Jayapura, Indonesia and Madang,
Papua New Guinea. (B) The monthly tritium concentration in precipitation was recorded
at Jayapura and Madang stations from 1957 to 1991. The tritium peak (~33 TU) was
recorded in 1964 at Jayapura station. Note that the tritium in precipitation at Madang on
September 1968 was removed due to an outliar. (C) As in (B), but reflects the decay of
tritium (half-life of 12.3 years) in June 2010 when the ice cores were recovered.



Figure S10 | Comparisons between NINO3 ERSST and rainfall δ^{18} O. (A) The 383 correlation field between annual ERSSTv4 and rainfall δ^{18} O at the GNIP station in 384 385 Jayapura (2.53°S; 140.72°E; 3 m.a.s.l, blue triangle) from 1961 to 1991 (dashed lines 386 indicate the area with p-values < 0.05). (B) The time series correlation between the rainfall δ^{18} O in Jayapura and the annual NINO3 ERSST (Area A). (C) As in (B), but with 387 388 the annual mean ERSST of Area B (5-20°S; 120-90°W). Area C (6-18°S; 140-160°E) in Panel (A) indicates a significant negative correlation between rainfall δ^{18} O and ERSST. 389 390 The thermal year covers the period of August of the previous year to July of the current 391 year.



Figure S11 | The depth-age relationships for Cores D1 and D1B. Plots are based on
the reference matching timescale in Table S6.

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420 Figure | S12 Comparison of the annual ice core δ^{18} O and temperature records. (A) 421 Annual δ^{18} O of D1 and D1B with the linear trend of D1 (dashed red line). (B) Annual 422 mean temperature anomalies of the global surface (25), the tropical (30°S - 30°N) 423 surface, and the tropical 550-mb level (25). (C) As in (B) but for the sea surface (25), the 424 surface, and the 550-mb level over the Papua glacier region (136 - 138°E; 4 - 6°S). (D) 425 The annual and seasonal mean precipitation anomalies over the Papua glacier region (26).

426 Table S1 | Characteristics of the satellite images used to determine the surface area

427 of the Papua glaciers

Date	Satellite	Source/Provider	Ground resolution	Off-nadir Angle
11 June 2002	IKONOS	Google Earth	0.8 m	6.6°
16 March 2015	WorldView-3	MapMart.com	0.3 m	15.0°
16 May 2016	Planetscope Doves	Planet.com	2.6 m	1.3°
12 August 2016	WorldView-2	MapMart.com	0.5 m	15.0°
28 March 2018	Planetscope Doves	Planet.com	3.6 m	0.3°

431 Table S2 | Characteristics of the available climatic records for the study area

		Weather		
		Grasberg-	Dispatch	Alpine
		Nursery	Tower	(ALP)
		(GRS)	(DISP)	
	Latitude	4.041 °S	4.068 °S	4.045 °S
	Longitude	137.120 °Е	137.114 °E	137.138 °E
	Elevation (m)	3945	4109	4400
	First date with data	1/1/97	1/1/99	1/1/00
	Last date with data	11/25/16	11/25/16	12/31/16
	Coverage since 1/1/1997 (%)	94.6	83.4	47.7
	Coverage since first date with	95.1	93.2	58.1
Tomporatura	data (%)			
remperature	Coverage since first date of	95.2	92.9	56.2
	the shortest time series (%)			
	Coverage since first date of	94.9	93.2	53.3
	the second shortest time			
	series (%)			
	First date with data	5/17/97	1/1/99	1/1/00
	Last date with data	11/25/16	11/25/16	12/31/16
	Coverage since 1/1/1997 (%)	93.2	83.5	63.6
	Coverage since first date with	95.5	93.3	74.9
Provinitation	data (%)			
riccipitation	Coverage since first date of	95.9	95.1	74.9
	the shortest time series (%)			
	Coverage since first date of	95.5	93.3	70.5
	the second shortest time			
	series (%)			

	P					
Period	Bias	Relative	Temperature	Precipitation	DDF for	DDF for
	(m)	bias (%)	lapse rate	lapse rate	snow (m	ice (m
			(°C/km)	(mm/km)	$^{\circ}C^{-1}d^{-1})$	$^{\circ}C^{-1}d^{-1})$
Jul 2010 –	1 22	19.2	0	-6 2[3 0]**	0.008[0.004]	0 004[0]**
Oct 2015	1.22	17.2	0	0.2[5.0]	**	0.00 [[0]
Nov 2015* – May 2016	-0.05	3.5	-0.1	-6.7[3.1] **	0.008[0.004] **	0.001[0]**
May 2016 – Nov 2016	-0.12	2.9	0.1	-6.7[3.1] **	0.008[0.004] **	0.015[0]**

433 Table S3 | Optimal parameter combinations for the mass balance model

* Due to remaining gaps in the meteorological data, the period modeled was Jan – May

435 2016.

- 436 ** Values in brackets correspond to ± 1 standard deviation.
- 437 DDF stands for Degree Day Factors (see Supplementary Text)

438

439 Table S4 | SEACLID/CORDEX-SEA model output used in this study, including

440 results from the statistical downscaling

Model Output	GCMs	Institution that	RCMs	RCP	CP Downscaling results			5
Name		developed the			Mean observation		Cramér-von Mises	
		GCMs			minus	s mean	test []	p-value]
					T(PC)	Duel	т	Du
					I (-C)	p (mm/d)	1	Рр
MPI-M-MPI-	MPI-ESM-MR	Max Planck	RegCM4	8.5	-0.14	2.72	3.94	19.22
ESM-MR		Institute for Meteorology,					[0.003]	[0]
		Germany		4.5	-0.15	0.6	4.49	2.78
							[0.002]	[0.01]
IPSL-IPSL-	IPSL-CM5A-	Institute Pierre-	RegCM4	8.5	-0.21	-1.72	2.41	20.54
CM5A-LR	LR	Simon Laplace, France					[0.015]	[0]
		Tunee		4.5	0.09	-1.57	5.66	14.86
							[0.001]	[0]
GFDL_REGCM	GFDL-ESM2M	GFDL, USA	RegCM4	8.5	-0.15	0.72	4.35	0.92
4						[0.002]	[0.066]	
				4.5	-0.13	0.85	2.54	1.29
							[0.013]	[0.046]
CSIROMK36_R	CSIRO MK3.6	CSIRO, Australia	RegCM4	8.5	-0.08	0.5	2.41	0.88
EGCM4							[0.015]	[0.069]
				4.5	-0.08	0.53	2.97	0.78
							[0]	[0.079]
CNRM5_REGC	CNRM-CM5	Centre National	RegCM4	8.5	-0.05	1.12	0.51	2.78
M4 de Recherches		de Recherches Meteorologiques					[0.1]	[0.01]
		France		4.5	-0.2	1.39	6.78	4.85
							[0]	[0.001]
1			1	1	I		I	

Table S5 | Descriptive statistics of δ^{18} O, δ D and *d* (deuterium excess) in Cores D1 and D1B

	Min	Mean	Max	Range
Core D1				
δ ¹⁸ O (‰)	-19.68	-16.07	-13.62	6.06
δD (‰)	-146.65	-116.36	-96.46	50.19
d (‰)	8.01	12.21	19.80	11.79
Core D1B				
δ ¹⁸ O (‰)	-20.32	-16.41	-14.06	6.26
δD (‰)	-147.77	-116.87	-96.56	51.21
d (‰)	10.09	14.41	19.22	9.13

Table S6 | Reference matching points of δ^{18} O in Cores D1 and D1B with NINO3 ERSST (Fig. 5B). The 1964 bomb horizon (bold) was identified at 23.4 m depth in

Core D1 and assuming the top layer represents the time when the ice cores were collected (May 2010).

Depth (
D1	D1B	Year
0.00	0.00	2010.4
3.97	4.25	2008
7.35	7.30	2003
8.25	8.15	2000
8.48	8.37	1998
9.80	9.40	1996
10.91	11.00	1992
12.02	12.06	1989
12.59	12.81	1987.5
14.92	14.23	1983
17.03	16.62	1975.5
19.54	19.34	1973
22.28	22.00	1966
23.39	23.39	1964

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