

Author's Response to Peer Review

On behalf of all of the authors I would like to thank both reviewers for their detailed and insightful comments. We feel that their advice has improved the manuscript and in particular we have:

- Improved our attribution of relevant background literature
- Edited our figures and tables so that their contents are easier to understand
- Included a new figure detailing the data coverage at each station (new figure 1)

Our response to specific comments follows, note that the reviewer comment is given in bold and our response is given in normal type. References to page numbers refer to the original manuscript.

Best wishes

Amber Leeson

Reviewer #1

page 1, line 12 (Abstract) "short period variability in time" - please quantify what time range you mean here.

Text added: '(i.e. intra-seasonal)'

p.1, l.24 Re. GrIS recent significant mass loss, please add the following two highly relevant, more recent references to Shepherd et al. 2012:

Hanna, Edward and Navarro, Francisco J. and Pattyn, Frank and Domingues, Catia M. and Fettweis, Xavier and Ivins, Erik R. and Nicholls, Robert J. and Ritz, Catherine and Smith, Ben and Tulaczyk, Slawek and Whitehouse, Pippa L. and Jay Zwally, H. (2013) Ice-sheet mass balance and climate change. Nature, 498 (7452). pp. 51-59.

van den Broeke, M. R., Enderlin, E. M., Howat, I. M., Kuipers Munneke, P., Noël, B. P. Y., van de Berg, W. J., van Meijgaard, E., and Wouters, B.: On the recent contribution of the Greenland ice sheet to sea level change, The Cryosphere, 10, 1933-1946, <https://doi.org/10.5194/tc-10-1933-2016>, 2016.

Both citations added alongside Shepherd et al., 2012 and Hanna et al., added into reference list.

p.1, l.27 Re. recent episodes of rare and extreme surface melt (2012), please add the following two highly relevant references to Ngheim et al. (2012):

Tedesco, M., Fettweis, X., Mote, T., Wahr, J., Alexander, P., Box, J. E., and Wouters, B.: Evidence and analysis of 2012 Greenland records from spaceborne observations, a regional climate model and reanalysis data, The Cryosphere, 7, 615-630, <https://doi.org/10.5194/tc-7-615-2013>, 2013.

Hanna, Edward and Fettweis, X. and Mernild, S. H. and Cappelen, J. and Ribergaard, M. H. and Shuman, C. A. and Steffen, K. and Wood, L. and Mote, T. L. (2014) Atmospheric and oceanic climate forcing of the exceptional Greenland ice sheet surface melt in summer 2012. International Journal of Climatology, 34 (4). pp. 1022-1037.

Both citations added alongside Nghiem et al., and added into reference list.

page 2, line 39: after Noel et al., 2016 reference, suggest inert new sentence:
"Alternative statistical downscaling techniques fulfill a similar purpose and give broadly comparable results (Wilton et al. 2017, Vernon et al. 2013), and add the following two relevant references. However, RCMs can also make...":

Wilton, D. J. and Jowett, A. and Hanna, E. and Bigg, G. R. and Van Den Broeke, M. R. and Fettweis, X. and Huybrechts, P. (2017) High resolution (1 km) positive degree-day modelling of Greenland ice sheet surface mass balance, 1870-2012 using reanalysis data. Journal of Glaciology, 63 (237). pp. 176-193.

Vernon, C. L., Bamber, J. L., Box, J. E., van den Broeke, M. R., Fettweis, X., Hanna, E., and Huybrechts, P.: Surface mass balance model intercomparison for the Greenland ice sheet, The Cryosphere, 7, 599-614, <https://doi.org/10.5194/tc-7-599-2013>, 2013.

Text edited as suggested including citations. Both papers added to reference list.

p.2, l.45 change to "fidelity at the regional OR SEASONAL scales does not...".

Edited as requested

p.2, l.53: GC-Net also needs Steffen et al. reference here.

Citation added.

p.3, l.69 change "max" to "maximum".

Edited as requested

p.3, ll.74/75 slightly reword to "The MAR version 3.5 used here has been EXTENSIVELY evaluated in...".

Edited as requested

p.3, l.87 "MAR-Era data ARE available...".

Edited as requested

p.4, ll.97 & 99: "data set" -> "dataset".

Edited as requested.

p.4, l.101 "below the threshold for three consecutive days"- based on daily mean temperature (and, if so, how is the latter calculated?) or what exactly? Needs a bit more detail/explanation since how this is defined can affect the results.

We use the maximum daily temperature time series to identify extreme events. We choose to use the maximum, rather than the mean, temperatures in order to capture high temperature events that may last for < 1 day.

Text edited on line 99 to read: 'type-specific threshold applied to the maximum daily temperature time series' and on line 100 to read: 'start once the maximum daily temperature'

p.5, l.127: Why is event frequency *positively* correlated with elevation in North Greenland/the dry snow zone?

We have yet to identify a satisfactory explanation for this which is why we decline to comment in the manuscript. We wonder if perhaps it is to do with increased exposure with elevation but further investigation is required to ascertain whether this is in fact the case.

p.5, l.154: change "~decadal scale" to "decadal scale".

Edited as requested

p.6, l.158: change ">1.5°C" to "≥1.5°C" since MAR_Era = 1.50°C.

Edited as requested.

p.7, ll.216-218: "extreme melt years on Greenland have been attributed to an increase in the frequency and duration of high pressure conditions...Greenland Blocking Index" -please add the two highly relevant citations and add them in the reference list:

Hanna, Edward and Cropper, Thomas E. and Hall, Richard J. and Cappelen, John (2016) Greenland Blocking Index 1851-2015: a regional climate change signal. International Journal of Climatology, 36 (15). pp. 4847-4861.

Hanna, E. and Jones, J. M. and Cappelen, J. and Mernild, S. H. and Wood, L. and Steffen, K. and Huybrechts, P. (2013) The influence of North Atlantic atmospheric and oceanic forcing effects on 1900-2010 Greenland summer climate and ice melt/runoff. International Journal of Climatology, 33 (4). pp. 862-880.

Citations added as requested, both in text and to reference list.

p.8, l.247 needs punctuation correction to "...given its assimilation of observations; however, we note that..."

Edited as requested.

p.9, l.259: "strongly controlled by geography" sounds a bit vague. Can you be more specific, e.g. say topography, elevation and ice/snow facies etc.?

ok, edited to read: '...geography (e.g. topography, elevation, latitude etc), though...'

p.18/Table 3: add units (e.g. °C/yr?) for "rate of change of mean daily temperature".

Edited as requested.

Reviewer #2

We would like to thank the reviewer for their assessment and particularly their kind words in support of our work. We are particularly pleased to hear that they found our submission 'fascinating and important'!

Our response to the reviewer comments is given here with the comment given in bold and our response in normal type. Where edits have been made to the paper on the reviewer's

recommendation, page numbers refer to the corresponding position in the **original** manuscript.

1. Using the GC net stations as representative samples of ablation, percolation and dry snow zone is problematic. Rather than being representative of a zone, they can be considered representative for a geographic region of varying sizes. The three stations from the ablation zone are in close proximity of each other on the west coast. All percolation zone stations are in the south, all dry snow zone stations are in the north. Some more rigorous analysis is needed before these stations can be assumed representative of the three regions if at all. I suggest another approach that focuses on model and station comparison rather than generalizing over the three zone. If the authors want to generalize about the three zone a more rigorous analysis of the representability of the stations for each of the zones are needed.

We agree with the reviewer that our station groupings are geographically proximal, however these groupings do also coincide with the ice sheet's melt zones; particularly the dry snow zone, which is mostly in the North of the ice sheet, and the percolation zone, the largest expanse of which is in the South. We have added contours to Figures 1 and 5 to show the coverage of these zones with respect to the distribution of stations.

We appreciate that there are only two stations in the ablation zone, albeit located in the region where a majority of ice ablation occurs (McMillan et al., 2016). We acknowledge the sparsity of the Gc-Net stations with respect to the size of the ice sheet on line 204 of the original manuscript and have added the following in line 220:

'Because we only have data for two ablation zone stations which are located in close proximity, further work is required to assess whether this is a general property of the ablation zone or restricted to this location; temperatures in general are much warmer here...'

2. Provide information about time span in addition to time series length for each of the GCnet stations and discuss implications of varying time series length and time period span on the extreme value statistics.

We have included a new figure (the new figure 1) to show the data coverage at the GC-Net stations used here, and amended Table 1 such that it shows the total amount of data (excluding gaps). We have also added text to introduce the new figure, refer explicitly to the table and discuss the significance of the missing data beginning on line 70 as follows:

'Our analysis focuses on 14 of the 18 stations; we found the remaining 4 stations to have temperature time series which were either too short or too patchy for robust statistical analysis. Figure 1 shows the data coverage at each of the 14 stations studied here and Table 1 gives the total number of years of data available, when gaps are excluded. We attribute these missing data to equipment failure and assume that it is unrelated to the occurrence of extreme high temperatures. As such we treat these data gaps as 'missing at random' and ignore them in our analysis. Since most of the missing periods cover whole years, rather than just a summer- or winter-period, this assumption is reasonable. '

It is true that varying time series length and span may have implications when comparing extreme event characteristics between sites; for example it is difficult to pick out a trend in extremes in shorter time series. In this study we did not find any evidence of a trend in extremes at any of the stations and in any case the main focus of our paper is on the comparison between GC-Net and MAR, both of which were commonly sampled at each station and subject to the same EVA procedure. We are hoping to investigate temporal trends in extremes in future work however and so we have amended line 260 to read:

'further work is needed to determine the relative contributions of potential physical drivers of extreme events at different locations and over different time periods'

3. The analysis of melt energy and extreme temperature events needs some work because extreme temperatures at several of the stations appear to not be associated with melting at all (i.e. Figure 1).

We are not sure what the reviewer is asking for here as Figure 1 does not consider melting; it presents extreme event characteristics at each of the stations.

Figure 5 shows PDDs at each station and attributes them to extreme events (blue portion of stacked bar) or normal conditions (orange portion of stacked bar). Extreme events produce PDDs at all of the GC-Net stations we consider in this study. The MAR model variants do not produce any melting at some of these stations. This is discussed in the manuscript on lines 199-201 and 250-256.

4. A clear presentation of the analysis behind the conclusions that MAR simulates duration of extreme temperature events but not frequency or magnitude of those events are not well supported with figures and tables. It seems that Figure 2 and 4 are indented for this purpose, but they are not clear (see more comments on the figure design below).

We have edited the figures in line with the reviewer's suggestions (see specific comments below) and added an additional reference to Figure 2 at line 143. This should be clearer now.

5. The analysis discussed in lines L242 to L248 belongs in the result section and needs some more elaboration to be convincing. First, a figure showing the amount of melt energy during extreme versus average conditions would be really nice to see. Second, you have to address the fact that some stations have extreme temperatures that are not above freezing and therefore no PDDs.

These points are discussed on lines 186-201 in the results section and the corresponding data is presented in Table 4. The stacked bars in Figure 5 also illustrate the PDDs during extreme (in blue) vs non-extreme (in orange) conditions. Hopefully the edits we have made to this figure will make this point clearer.

Minor comments

L14: Clarify that you are examining extreme positive temperature anomalies (as opposed to positive and negative)

Edited to read 'extreme positive temperature'

L92: Clarify what MAR grid cell elevation that lower than the AWS, e.g. the center point?

'data' changed to 'cell centre' to clarify.

L105: Explain PDD, the concept may not be widely known outside glaciologist circles.

Edited to read: 'Melting is most appropriately calculated as a function of the surface energy balance; however measurements of variables required to calculate the surface energy balance (e.g. net radiation, wind speed) are not consistently available at the Gc-Net stations. Positive Degree Days are an estimate of the magnitude and duration of above-zero temperature events and are typically well-correlated with melting (e.g. Braithwaite (1995),

Huybrechts et al. (1991)). Here we calculate positive degree-days (PDD) for both observed and modelled temperatures and take this to be a reasonable approximation for melt energy'

L108: Provide more background for equation 1. Typically PDD are a function of mean daily temperatures and a temperature to melt conversion factor.'

We choose to calculate PDDs by integrating a model of daily temperature variability because we are interested in potential melt that occurs during unusually warm conditions. It may be that the mean temperature for a day is 'normal' but the daytime has been unusually warm and the night time has been unusually cold, for example during persistent high pressure conditions in summer. As such we could 'miss' melting that occurs during the warmer part of the day if we use an equation based on mean daily temperatures. This concept is not new as a similar approach has been applied to higher resolution (i.e. sub-daily) data in studies of Antarctica (e.g. Vaughan, 2006, Barrand et al., 2013)

L135: Please write more about Figure 2. It is difficult to interest and draw conclusions from it beyond that MAR data are colder than GC-net station data.

Figure 2 is also discussed in the remainder of this paragraph, we have added an additional reference to the figure on line 143 to clarify.

L140-149: Is this data shown in figures or tables? If so clarify how, if not considering adding a figure to clearly illustrate these findings.

The text on lines 140-149 refers to Figure 2. A citation to the figure has been added on line 143, and more information has been added to the figure to clarify.

L154: Remove the gaps from the total count of data points to give an accurate count of the number of years with data (in table 1).

Done.

L170: Explain "raw Era-interim" and how it is different form "Era-interim"

We use the term raw Era-Interim in order to clearly distinguish from MAR forced by Era-Interim. Have edited to clarify as follows: '...raw Era-Interim output (i.e. not MAR forced with Era-Interim) captures...'.
Era-Interim

L180: Provide units for PDD here and elsewhere

PDDs are in °C, edited throughout to clarify

Figure 1: Show the spatial distribution of the ablation, percolation and accumulation zones

This information has been added to Figure 1 and Figure 5.

Figure 2: This figure needs work. First, a scale is needed to be able to interpret the height and width of the boxes.

Scale added.

Second, it is unclear why the bars are centered in each box.

The bars are centred in each box the better to contrast the difference with the observations.

Third, units on the colorbar are needed.

Units added

Forth, five shades of blue are displayed in the figure, extend the colorbar to capture the darkest blue.

The darkest blue is the bottom out of bounds colour. Apologies; this fell off the submitted version.

Finally, the figure captions says that the figure shows model results, but the colorbar text says it is showing deviations from observations.

Yes, it is the modelled magnitude – the observed magnitude, i.e. the model anomaly with respect to the observations. Caption text edited to be: 'of the modelled magnitude' to clarify.

Figure 3. Clarify the meaning of the black dotted line. Add units to each axis. It would be useful to also show the RMSE in the plots.

Text added to figure caption: 'Black dotted line denotes a 1:1 fit.'

RMSE added as annotation. Caption edited to read: 'Pearson's correlation co-efficient and the root mean squared error (°C) between the data is annotated'

Figure 4. Same comments as Figure 3.

We assume that you mean the same as Figure 2 and have edited accordingly.

Figure 5: Add units to x-axes

Edited as requested

Table 3: Explain the “nyrs” variable. Clarify what spatial domain the data values are calculated for.

Text added to figure caption: 'The number of years of data in total (i.e. the sum of the number of years of data at each station) is also identified (nyrs).'

The spatial domain for each melt zone is taken to be that defined in McMillan et al., 2016, this is specified in lines 112-115 in the 'methods' section.

Table 4: Provide units for PDD

Edited as requested

References

- Barrand, N. E., Vaughan, D. G., Steiner, N., Tedesco, M., Munneke, P. K., Van Den Broeke, M. R. & Hosking, J. S. 2013. Trends in Antarctic Peninsula surface melting conditions from observations and regional climate modeling. *Journal of Geophysical Research-Earth Surface*, 118, 315-330.
- Vaughan, D. G. 2006. Recent trends in melting conditions on the Antarctic Peninsula and their implications for ice-sheet mass balance and sea level. *Arctic Antarctic and Alpine Research*, 38, 147-152.

Extreme temperature events on Greenland in observations and the MAR regional climate model

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Abstract.

Melt water from the Greenland ice sheet contributed 1.7-6.12 mm to global sea level between 1993 and 2010 and is expected
10 to contribute 20–110 mm to future sea level rise by 2100. These estimates were produced by regional climate models which
are known to be robust at the ice-sheet scale, but occasionally miss regional and local scale climate variability. To date, the
fidelity of these models in the context of short period variability in time (i.e. intra-seasonal) has not been assessed, for example
their ability to simulate extreme temperature events. We use an event identification algorithm commonly used in Extreme
Value Analysis, together with observations from the GC-Net, to assess the ability of the MAR RCM to reproduce observed
15 extreme positive temperature events at 14 sites around Greenland. We find that MAR is able to accurately simulate the
frequency and duration of these events but underestimates their magnitude by more than half a degree, although this bias is
much smaller than that exhibited by coarse-scale Era-Interim reanalysis data. As a corollary, melt energy in MAR output is
underestimated by between 16% and 41% depending on global forcing applied. Further work is needed to precisely determine
the drivers of extreme temperature events, and why the model underperforms in this area, but our findings suggest that biases
20 are passed into MAR from forcing data. This is important because these forcings are common between RCMs and their range
of predictions of past and future ice sheet melting. We propose that examining extreme events should become a routine part
of global and regional climate model evaluation and addressing shortcomings in this area should be a priority for model
development.

1 Introduction

25 Since the 1990s, the Greenland Ice Sheet has shifted from a state of near mass balance, to one of significant mass loss (Shepherd
et al., 2012, Hanna et al., 2013a, van den Broeke et al., 2016), contributing approximately 10% to the measured global sea
level rise during the last two decades (Church, 2013). Since 2010, the rate of mass loss from Greenland has increased and the
ice sheet has experienced episodes of rare and extreme surface melt (Nghiem et al., 2012, Hanna et al., 2014, Tedesco et al.,
2013). For example in 2012, the summer melt extent reached 98.6% of the entire ice sheet; thought to be the greatest melt
30 extent in over a century (ibid). In addition to directly removing more of the ice sheet into the sea, melting reduces the reflectivity

of the ice sheet and can warm the perennial snow pack (through latent heat release when the melt water refreezes), both of which act as a positive feedback to further enhance melt. These processes also alter the dielectric properties of the ice sheet surface, which makes it more difficult to measure surface height change using satellite-borne radar instruments (McMillan et al., 2016). An understanding of the location, frequency, duration and magnitude of melting is therefore necessary to 1) understand the ice sheet's response to climate change, 2) interpret contemporary measurements of ice sheet volume change and 3) to constrain predictions of future ice sheet state.

Mass lost through meltwater runoff and gained through snowfall together comprise the ice sheet's surface mass balance (SMB), which is typically assessed at the ice sheet wide scale using Regional Climate Models (RCM). RCMs act as physically based interpolators of relatively coarse resolution climate reanalysis data, and produce high resolution estimates in areas where the local climate exhibits high spatial variability i.e. ice sheet margins (Noel et al., 2016). Alternative statistical downscaling techniques fulfil a similar purpose and give broadly comparable results (Wilton et al. 2017, Vernon et al. 2013). RCMs can also make high resolution predictions of future climate, when boundary forcing is applied by global climate model (GCM) output instead of reanalysis data. In the last IPCC report, the MAR, RACMO2 and MM5 RCMs reported that while SMB remains positive (net increase in mass due to surface processes) increases in melting were responsible for a sea level contribution of 0.23-0.64 mm yr⁻¹ during 2005-2010 (Church, 2013). RCMs are known to perform well when compared to integrated quantities, for example mean annual melt measured at weather stations or total mass loss from the ice sheet measured by GRACE (van den Broeke et al., 2016). However, fidelity at the regional or seasonal scales does not necessarily translate to the local scale (e.g. Medley et al., 2013). Extreme melt events, for example, tend to be localised in time (typically only lasting for a day or so). While RCM predictions of melt *extent* during extreme events have been found to be reliable (Tedesco et al., 2011), an assessment of their ability to simulate the frequency, duration and magnitude of these events, and how this might affect their projections of future ice sheet change, has yet to be performed.

In this paper, we use advanced statistical techniques for extreme event identification to compile a statistical climatology of extreme temperature events on Greenland since the 1990s using data from 14 automatic weather stations forming part of the Greenland Climate Network (GC-Net, Steffen et al., 1996). Note that these are distinct from extreme melt years as it is possible to have multiple extreme temperature events in a year. We then use these data, together with temperature estimates from the MAR regional climate model (Fettweis et al., 2017) to evaluate the model's ability to capture the frequency, duration and magnitude of these events when forced by climate reanalysis and by GCM data. Finally, we estimate melt energy available at the GC-Net stations during this time using a positive degree day sum (PDD) and assess the degree to which discrepancies between observed and modelled characteristics of extreme events affects MAR based estimates of melt energy.

2 Methods and Data

2.1 Greenland Climate Network data

The Greenland climate network (GC-Net) consists of 18 automatic weather stations (AWS) distributed around the ice sheet. We refer the reader to Steffen et al. (1996) for details but briefly summarise here. The first station (Summit) began operation in 1995, with others coming online at various times since then. The AWS measure a range of meteorological variables, of which the temperature and pressure time series are the most complete. ~~Our analysis focuses on 14 of the 18 stations (Figure 1, Table 1) we found the remaining 4 stations to have temperature time series which were either too short or too patchy for robust statistical analysis.~~ The GC-Net stations each have four temperature sensors (2 different instruments mounted at 2 different heights), here we use data from the Type-E Thermocouple mounted at position 1 at all sites except NGRIP, and Saddle during 2010-2016, for which we used data from the Type-E Thermocouple mounted at position 2. Measurements are taken hourly, we use these data to calculate daily maximum and mean values for compatibility with MAR output.

Our analysis focuses on 14 of the 18 stations; we found the remaining 4 stations to have temperature time series which were either too short or too patchy for robust statistical analysis. Figure 1 shows the data coverage at each of the 14 stations studied here and Table 1 gives the total number of years of data available, when gaps are excluded. We attribute these missing data to equipment failure and assume that it is unrelated to the occurrence of extreme high temperatures. As such we treat these data gaps as 'missing at random' and ignore them in our analysis. Since most of the missing periods cover whole years, rather than just a summer- or winter-period, this assumption is reasonable.

2.2 MAR Regional Climate Model

The MAR model is an RCM developed and extensively evaluated to study the present Greenland climate and SMB from the beginning of the last century (Fettweis et al., 2017) as well as to perform future projections of GrIS SMB for the last IPCC report till the end of this century (Fettweis et al., 2013). It is fully coupled with a snow energy balance model dealing with the energy and mass exchanges between surface, snow, ice and atmosphere. The MAR version 3.5 used here has ~~intensively been extensively been successfully~~ evaluated in Fettweis et al. (2017) with daily in situ PROMICE based AWS measurements over 2008-2010, daily satellite derived melt extents over 1979-2010 as well as SMB measurements and ice cores over 1958-2010. We chose to use MARv3.5 in this study since this is the model version which was used to make the most recent set of estimates of future ice sheet change (Fettweis et al., 2013). We refer to Fettweis (2007) and Fettweis et al. (2013, 2017) for more details about MAR and its surface scheme.

Here, we use data from MAR simulations forced with the ERA-Interim reanalysis (e.g. Fettweis et al., 2017), and with the global models CanESM2, MIROC5 and NorESM1 over 1995-2015. CanESM2, MIROC5 and NorESM1 have been found to

be the best models (in respect to ERA-Interim over 1980-1999) from the CMIP5 database over Greenland from which 6 hourly
95 outputs were available (Fettweis et al., 2013). MAR is forced every 6 hours at its lateral boundaries with temperature, humidity,
wind and surface pressure. Sea surface temperature and sea ice extent is also prescribed into the MAR integration domain from
the forcing data every 6 hours. Hereafter we refer to MAR variants with forcing by Era-Interim, NorESM1, CanESM2 and
MIROC5 as MAR-Era, MAR-Nor, MAR-Can MAR-MIR, respectively. MAR-Era data ~~is~~ is available continuously during
our study period (1995-2015 inclusive). For the GCM forced model runs, we use historical simulations until 2006 and
100 simulations performed under forcing by the RCP8.5 climate change scenario (van Vuuren et al., 2011) thereafter. This is
reasonable because observed greenhouse gas concentrations followed the RCP8.5 scenario during this period, and in any case
the differences between the RCP scenarios between 2006-2015 are very small. For comparison with the GC-Net data we pick
the MAR model grid cell closest to the AWS location in terms of latitude and longitude of the cell centre. The MAR ~~data-cell~~
centre is typically at a lower elevation than the AWS and so we apply a lapse rate based correction to MAR temperature data
105 (assuming 0.71°C per 100 m, Steffen and Box (2001)). We restrict the model time series at each station to periods where GC-
Net data are also available.

2.3 Extreme Value Analysis (EVA)

Extreme value analysis provides a toolbox of methods for the identification and statistical modelling of extreme events (Coles,
2013) i.e. events that are unusually large or small when compared to the central behaviour of a data-set. For a given site and a
110 given data type (observations, MAR-Era, MAR-MIR, MAR-Nor and MAR-Can), we identify the extreme events using a site-
and type-specific threshold applied to the maximum daily temperature time series. To enable a fair comparison, the threshold
is taken always to be the 90% quantile of the data-set in question and an extreme event is deemed to start once the maximum
daily temperature exceeds this threshold. The event ends after the temperature has been below the threshold for three
consecutive days. This method of event identification is known as the runs method (Smith and Weissman, 1994). It follows
115 that the durations, as well as both frequencies and magnitudes, of events are random. Note that here we take the magnitude of
an event to be the largest of the daily maxima within that event.

2.3 Positive Degree-Day Sum

Melting is most appropriately calculated as a function of the surface energy balance; however measurements of variables
120 required to calculate the surface energy balance (e.g. net radiation, wind speed) are not consistently available at the Gc-Net
stations. Positive Degree Days are an estimate of the magnitude and duration of above-zero temperature events and are typically
well-correlated with melting~~PDDs and melting are typically well correlated (e.g. Braithwaite (1995), Huybrechts et al. (1991)).~~
Here ~~We~~ we calculate positive degree-days (PDD) for both observed and modelled temperatures and take this to be a reasonable
approximation for melt energy; ~~PDDs and melting are typically well correlated (e.g. Braithwaite (1995), Huybrechts et al.~~
125 ~~(1991)).~~ Diurnal temperature variability is modelled using eq 1 and PDDs are calculated by integrating eq 1 where $T > 0^{\circ}\text{C}$.

$$T = A \sin(\phi t) + B \quad (1)$$

Where A is daily maximum temperature, B is daily mean temperature and ϕ is one day. Daily mean and maximum temperature are output by MAR; for GC-Net data daily mean and maximum are calculated based on hourly data.

2.3 Melt zone definitions

We use the definitions of the ablation, percolation and dry snow zones first identified in McMillan et al. (2016) using RACMO2.3 simulations of SMB and surface melt. Briefly, the area of the ice sheet lying below the equilibrium line in a majority of years between 2009 and 2014 is defined as the ablation zone. The area of ice where melt did not exceed 5 mm w.e. on any day during this period is defined as the dry snow zone, with the remainder being classed as the percolation zone. Using these definitions we find areas of 0.23, 0.61 and 0.80 million km² for the ablation, percolation and dry snow zones respectively.

3 Results

3.1 Extreme temperature events

We apply extreme value analysis to observed daily maximum temperatures from GC-Net in order to compile a statistical climatology of extreme temperature events on Greenland (Table 2). Each location is considered independently and the timing of statistically extreme events is not necessarily contemporaneous between stations. Extreme events are characterised in terms of their frequency, duration and magnitude; we use ‘duration’ and ‘magnitude’ to refer to median values accross all events observed/modelled. We assess these characteristics in the context of station geography i.e. elevation, latitude and melt zone (see methods). Each of the three characteristics is dependent on elevation, however the nature of that dependence and the role that latitude and melt zone play in the relationship is different for each (Figure 24). Extreme temperature events occur 4-8 times per year and event frequency is negatively correlated with elevation in South Greenland; events become less frequent the higher the station is on the ice sheet. Event frequency is positively correlated with elevation in North Greenland/the dry snow zone (Figure 4a2a). Events last between 5-10 days, and duration is positively correlated with elevation for all stations (Figure 4b2b). However events tend to last longer (by ~1 day) at stations in the dry snow zone/North Greenland than at stations at similar elevations in the percolation zone/South Greenland. Event magnitude is negatively correlated with elevation at all stations (Figure 4e2c), but elevation has a stronger influence on event magnitude in the dry snow and ablation zones (-4.4 °Ckm⁻¹) than in the percolation zone (-1.8 °Ckm⁻¹).

We compare the degree to which MAR is able to capture the observed climatology of extreme events at GC-Net stations by repeating the same extreme value analysis with output from each of the MAR model variants (Figure 32). In addition to considering each station independently, we also consider each model variant independently, i.e. there is no common event mask. This is because the GCM forced model variants are designed to simulate climatic variability over typically climatic periods like 20-30 yrs, which is not necessarily contemporaneous with observed variability in a given time period. We exclude

JAR2 from the remainder of this analysis due to the large discrepancy in elevation between the station and the corresponding grid cell in MAR (316 m, Table 1) and the dependency we find between elevation and extreme event characteristics (Figure 2). Whilst all of the four model variants typically simulate the duration of extreme events reasonably well (i.e. within 1 day per event), they underestimate event frequency at most of the stations (Figure 3). This is most notable for the GCM forced model variants MAR-MIR, MAR-Nor and MAR-Can which underestimate event frequency by 1.12, 0.75 and 0.65 events per year respectively. Similarly, all of the model variants underestimate the observed event magnitude by more than half a degree at most of the stations (though notably not the two remaining in the ablation zone, in fact event magnitude at Swiss Camp is overestimated). In terms of the individual model variants, the MAR-Era simulation is best able to reproduce event frequency (-0.09 events on average), the MAR-ERA and MAR-MIR simulations are both best able to reproduce event duration (+/-0.04 days on average) and the MAR-Era simulation is best able to reproduce event magnitude (-0.76 °C on average). MAR-Nor is the poorest performing model variant overall.

3.1 Mean temperature and mean summer temperature at GC-Net stations in MAR

We assess the ability of the four MAR variants to reproduce temperature observed by the GC-Net more generally by comparing the mean and trend of the entire daily mean temperature time series at each station location (Table 3). We present aggregate statistics for the entire time period in order to account for the fact that the GCM forced MAR variants are predicting climatic variability at the $\sim$ decadal scale. The number of years of data (including gaps in the time series) is given for each station in Table 1. Results are presented by melt zone, where values are an average of all stations in that zone, weighted by the number of years of data available for each station. Both MAR-Era and MAR-MIR overestimate mean daily mean temperature (i.e. the average of all of the daily means) by $\sim 1^\circ\text{C}$, although this signal is dominated by a large discrepancy in the dry snow zone where both model variants are too warm by $\geq 1.5^\circ\text{C}$. Both model variants however, show good agreement with the observations in the ablation zone (-0.24°C and -0.34°C , respectively), which is where the most melting occurs. MAR-Can and MAR-Nor both underestimate temperatures overall, and give better agreement with observations in general (-0.13°C and -0.21°C , respectively), but they exhibit a poor performance in the ablation zone (both variants $> 1^\circ\text{C}$ too cold). Considering only the summer (JJA) daily mean temperatures, with the exception of MAR-Can in the percolation zone, all MAR variants are too cold in all zones and overall. MAR-Can performs best overall in summer, with a bias of just -0.01°C . All model variants reproduce observed trends in both all and summer temperatures to within 0.1°Cyr^{-1} (most within 0.05°C per year). We evaluate the ability of MAR-Era to reproduce observed climate variability by comparing modelled vs observed mean annual and mean summer (JJA) temperatures (Figure 34). MAR-Era is well able to capture observed inter-annual variability in both. Mean annual temperatures are particularly well correlated with Pearson's correlation co-efficient (r) values in the range 0.77-1.00. Inter-annual variability in mean summer temperatures is less well captured ($r = 0.62-94$, if JAR2 is ignored). The low bias in summer temperatures described above is also evident at the inter-annual timescale in the MAR-Era simulation.

3.3. Extreme temperature events in Era-Interim data

We assess the degree to which the raw Era-Interim output (i.e. not MAR forced with Era-Interim) captures extreme temperature events at GC-Net stations (Figure 45). In comparison with the same data for MAR-Era, using the raw Era-Interim output yields a poorer match to observations at all sites except NASA-U and NGRIP. The average absolute bias in magnitude of extreme temperature events is 0.87°C in MAR-Era and 1.81°C in the raw Era-Interim data output. In general, Era-Interim underestimates temperatures during extreme events in a similar manner to MAR-Era. However Era-Interim overestimates temperatures in the Swiss Camp region and north east of the ice sheet.

3.3 Melting during extreme temperature events

We use a positive degree day sum (see methods) to approximate melt energy available during extreme and non-extreme conditions at each of the 13 stations (Figure 56). We note that the difference in abundance of melt energy between adjacent melt zones is roughly an order of magnitude, with observed total PDDs per station of 617, 96 and 5 °C in the ablation, percolation and dry snow zones respectively. All MAR variants are able to reproduce this gradient. We find that the dependence of observed melt energy on statistically extreme temperatures also scales with elevation ($r^2=0.71$, $n=13$); one-third and ~95% of all melting occurs during extreme events in the ablation and dry snow zones respectively. MAR-Era is able to reproduce this pattern ($r^2=0.71$) but the relationship is less clear in data from the GCM forced variants ($r^2 = 0.23-0.43$).

We compare differences in the total PDDs observed and predicted during the entire study period; all of the MAR variants are found to underestimate total PDDs (Table 4). During extreme events, we see a two-fold increase in the model bias for the MAR-Era, MAR-MIR and MAR-Can model variants; PDDs are underestimated by 26%, 32% and 22% during extreme events, and 12%, 18% and 10% respectively during non-extreme conditions. In the MAR-Nor simulation, PDDs are underestimated to a greater degree during non-extreme temperature conditions. The relative influence of model bias during extreme events is spatially variable. In the ablation zone, the total bias during extreme events is comparable to that during non-extreme events except that the two biases are of the opposite sign; PDDs are over-estimated in the ablation zone during extreme events and under-estimated in the ablation zone during non-extreme events. Given the relative contribution of melting during extreme events to overall melting here however (just 33%), this results in an underestimate overall of 5%, 12%, 22% and 7% for MAR-Era, MAR-MIR, MAR-Nor and MAR-Can respectively. Conversely, in the percolation zone, PDDs are underestimated during extreme events and over-estimated during non-extreme conditions. However again the signal observed during the dominant regime (i.e. 96% of PDDs occur during extremes) leads to a large underestimate overall (52%, 58%, 84% and 40% for the model variants as before). In the dry snow zone it is more difficult to partition the relative influence of extreme vs non extreme events on total PDDs because there is far less melting here; a PDD total of just 5 PDDs °C per station over the entire study period. This is particularly of note for Summit and NGRIP stations which are high up and far inland on the ice sheet; very small amounts of melting are observed here but no melting is modelled by any of the model variants.

4.1 Extreme temperature events in GC-Net Observations

Despite our relatively small sample, given the size of the ice sheet, we see clear relationships between extreme event characteristics and elevation, latitude and melt regime. It is not surprising that extreme temperature events exhibit a stronger magnitude at lower-lying locations, given the atmospheric temperature lapse rate, but it is interesting that this relationship is less strong for the five percolation zone stations than for stations in the ablation and dry snow zones. We speculate that this is a result of heat exchange at the snow surface moderating near-surface temperatures in this region; sublimation is a known energy sink in the percolation zone in the summer (Ettema et al., 2010). In South Greenland, extreme events at lower elevations tend to be more frequent and of shorter duration than those higher up on the ice sheet. Temperature anomalies can be associated with cloudiness (reflecting upwelling longwave radiation back down to the surface) and lower-lying stations are more likely to experience short-term periods of orographic cloud cover. This is particularly likely to affect West Greenland which lies in the path of the prevailing summer circulation pattern and consequently receives moisture-laden onshore flow during the summer (Ohmura and Reeh, 1991). In North Greenland however, we see that extreme events become both longer-lasting, and more frequent, as elevation increases. Longer extreme temperature events are likely associated with high pressure conditions which are relatively persistent. In fact, extreme melt *years* on Greenland have been attributed to an increase in the frequency and duration of high pressure conditions promoted by wider scale atmospheric pressure gradients such as the North Atlantic Oscillation and the Greenland Blocking Index (e.g. Nghiem et al., 2012, [Hanna et al., 2013](#), Lim et al., 2016, [Hanna et al., 2016](#)). Extreme temperature events are responsible for the vast majority of melt energy produced in the percolation and dry snow zones on the ice sheet but contribute a much smaller proportion to overall melt energy in the ablation zone. Because we only have data for two ablation zone stations which are located in close proximity, further work is required to assess whether this is a general property of the ablation zone or restricted to this location. ~~This is because t~~emperatures in general are much warmer here, and extreme events are not required to generate melting.

4.2 Extreme temperature events in MAR simulations

All of the four MAR model variants underestimate the frequency of extreme events but simulate their duration well. This suggests that MAR is able to reproduce the persistence of conditions driving extreme temperature events when they arise in the model. All MAR variants under-estimate the magnitude of extreme temperature events at most stations, in most cases by >0.5°C. This can be explained in part by a general low bias in modelled summer temperatures; although the magnitude of this bias is not sufficient to account for the magnitude of the data-model mismatch during extreme periods. For example, MAR-Era exhibits a bias of -0.35°C during summer and -0.76°C during extreme temperature events. The raw Era-Interim output also exhibits a low bias during extreme temperature events at most of the GC-Net stations, with notable exceptions being North East Greenland and the most marginal stations at which temperature during extremes is over-estimated. This suggests that the low bias we see in the MAR model during extreme periods could be an artefact of the forcing data. This is important because

Era-Interim and the GCMs examined here are commonly used to force other regional and local scale models (e.g. RACMO2); their use is not restricted to MAR. The version of MAR which is analysed here (v3.5) is known to underestimate the atmospheric liquid water content and so cloudiness (Fettweis et al., 2017) which may also contribute to the cold bias in temperature extremes. However, we repeated the analysis with the most recent version of MAR (v3.7) in which a correction for this has been incorporated and this yielded no noticeable difference in the result. All of the MAR model variants and Era-Interim over-estimate event magnitude at stations in the ablation zone, JAR and Swiss Camp. We attribute this to difference in albedo between the bare ice in the ablation zone and the snow-covered surface at higher elevations. Energy exchange in bare ice areas is generally more sensitive to sunny conditions; this likely explains why the biases are opposite in this area compared to the percolation and dry snow zones where the albedo is high enough to prevent this sensitivity.

Melt energy simulated by MAR is underestimated by 19%, 25%, 41% and 16% when forcing is provided by Era-Interim, MIROC5, NorESM1 and CanESM2 respectively. However during extreme events, model biases in terms of melt energy are double those calculated during non-extreme, positive temperature, conditions. This is important because approximately half of all melt energy is generated during extreme events. In general, the GCM forced MAR simulations perform more poorly than the Era-Interim forced simulation, with the exception of MAR-Can (bias = 16% vs 19% for MAR-Era). We would expect the reanalysis forced simulation to perform the best, given its assimilation of observations; however we note that the difference is not large.

We observe melt energy generated at the two highest/furthest inland stations in our sample; Summit and NGRIP, but none of the MAR variants simulate any melting at either of these stations during our study period. It is important to note that these are very small quantities and would not impact ice-sheet wide estimates of melting, however melting is also important because of its role in ice sheet albedo; wet snow is less reflective than dry snow. A significant melt event can be defined as achieving > 1mmWE/day (Franco et al., 2013), and with the exception of Summit in 2012 this was not achieved at either station during the study period. Nonetheless, as the climate warms melting at these locations is likely to be more abundant and properly capturing melt conditions here will become even more important.

4 Conclusions

Analysis of GC-Net temperature data shows that the frequency, magnitude and duration of extreme temperature events on Greenland are strongly controlled by geography (e.g. elevation, latitude etc), though further work is needed to determine the relative contributions of potential physical drivers of extreme events at different locations and over different time periods. The MAR regional climate model accurately predicts the duration of extreme temperature events on Greenland but underestimates their frequency by around 1 day per year and underestimates event magnitude by >0.5°C. While this is an improvement over coarse-scale reanalysis data, it nonetheless leads to an under-estimate in melt energy, which we calculate to be 16-41% during

our study period, dependant on model forcing chosen. MAR-based predictions of future melting are calculated using an energy
 285 balance method which has been shown to perform well against observations in the past (Fettweis et al., 2017). However since
 temperature plays a significant role in the energy balance equations (though melt does not linearly increase with temperature),
 it is likely that these predictions are affected by the exaggerated model bias we find during extreme events in our study. Further
 work is needed to determine why the model underperforms in this area, and if other similar models have the same limitation.
 We identify this as a model development priority to ensure that MAR based estimates of ice sheet change are both
 290 comprehensive and robust.

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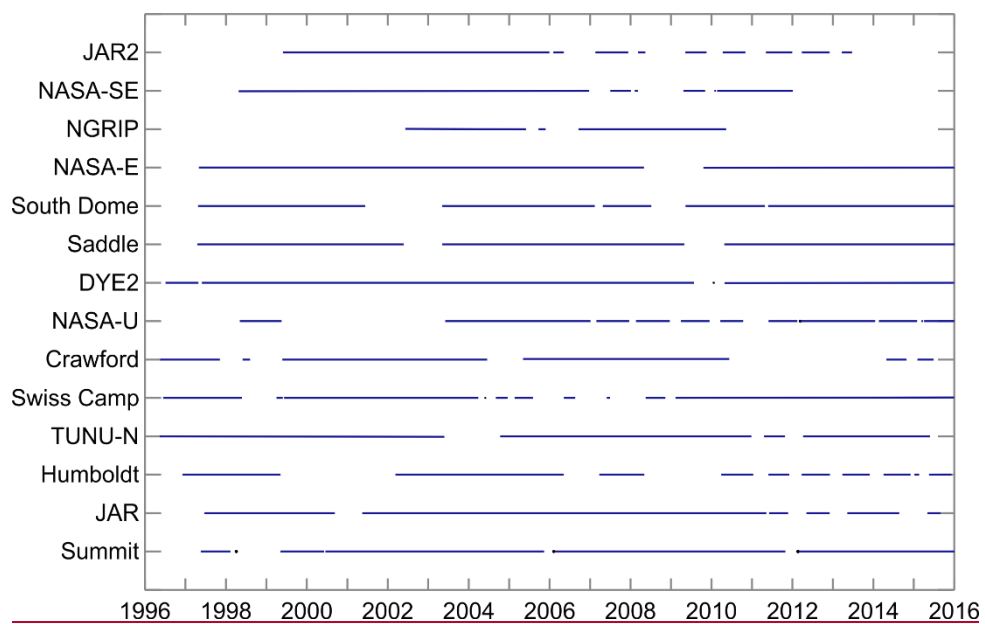


Figure 1: Data coverage in GC-Net temperature record.

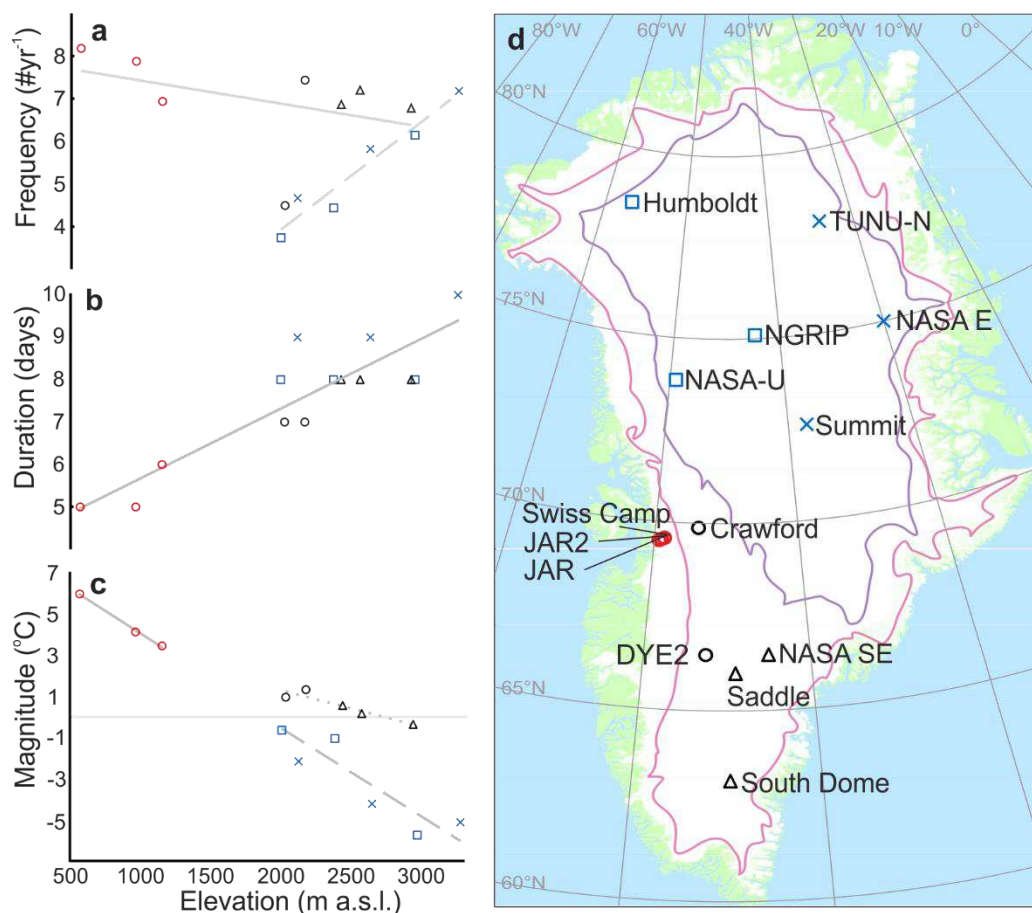
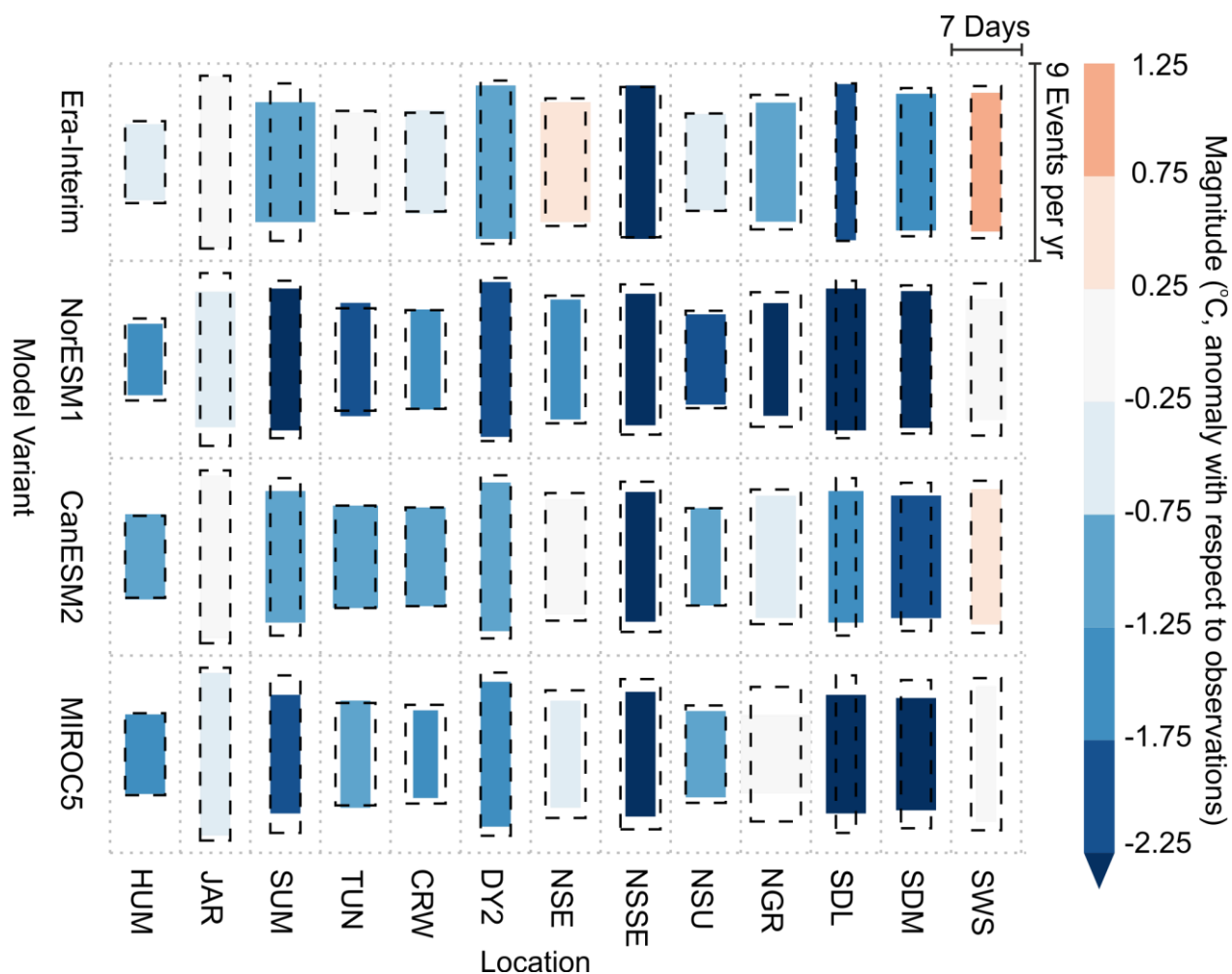


Figure 12: Frequency, duration and magnitude of extreme events observed at GC-Net automatic weather stations. Shapes indicate location: south west = circle, north west = square, north east = cross, south east = triangle. Saddle is located on the ice divide and we choose to represent it as a south east station. Colours indicate melt zone: red = ablation zone, black = percolation zone, blue = dry snow zone. In (a) solid grey line represents a linear fit to data from the ablation and percolation zone stations, dashed line in (a) represents a linear fit to dry snow zone station data. In (b), solid grey line represents a linear fit to all data. In (c) solid grey line represents a linear fit to ablation zone stations, dashed grey line represents a linear fit to dry snow zone stations and dotted grey line represents a linear fit to percolation zone stations. In (d) contours delineate lower limit of percolation (pink) and dry snow (purple) zones.



385 **Figure 32.** Median frequency, duration and magnitude of extreme temperature events as simulated by each model variant at each station. Frequency is denoted by the height of each box, duration is indicated by the width of each box, observed values are given by the dashed black boxes. Box colours indicate the departure of the modelled magnitude from the observed value, blue colours indicate an underestimate, red colours indicate an over estimate. Note that a temperature lapse rate has been applied to modelled temperatures to account for the difference in elevation between AWS and MAR elevation.

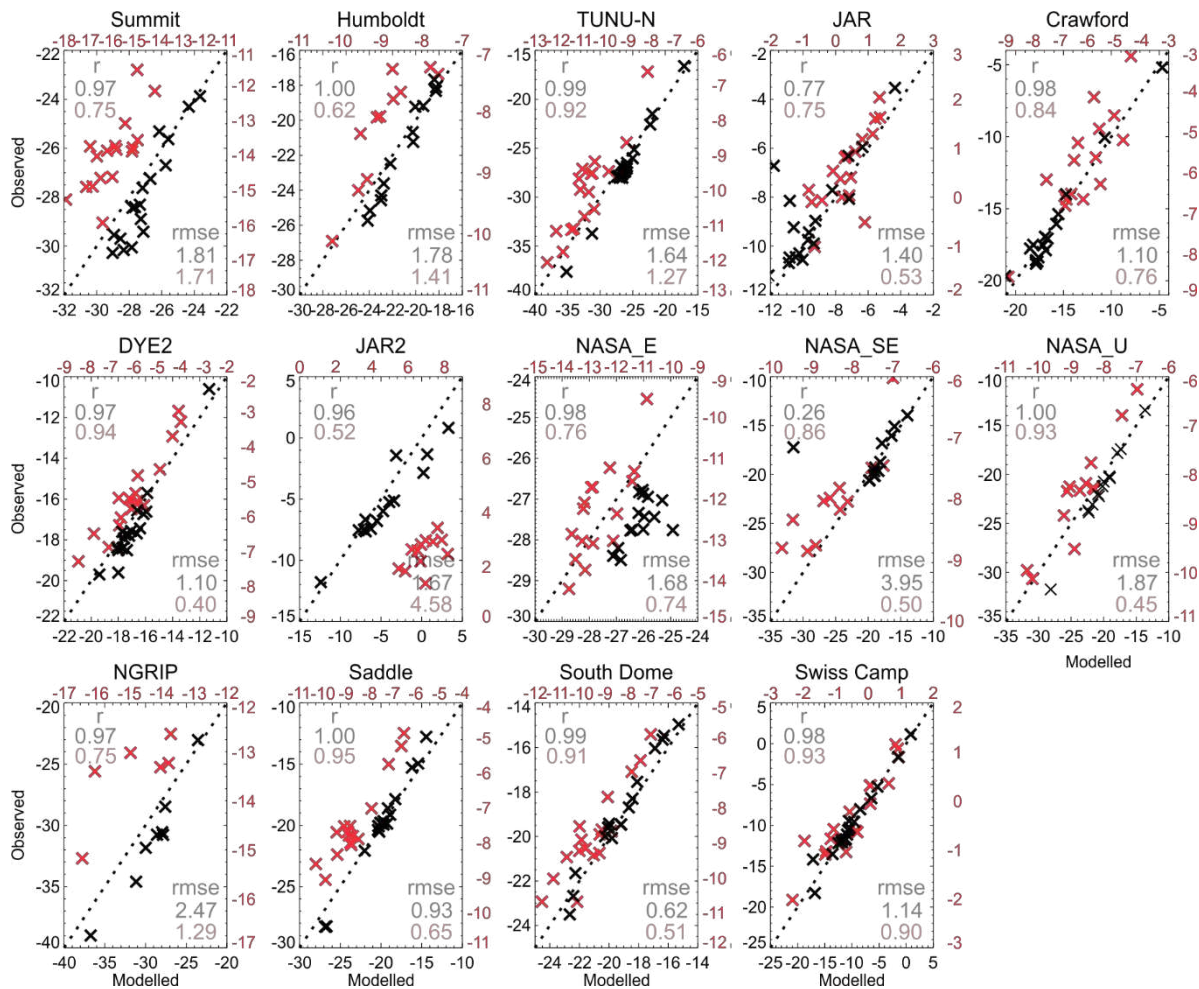


Figure 43: Modelled vs observed mean annual and mean summer temperature at each of the GC-Net stations. Modelled values are as simulated by MAR-Era. Pearson's correlation co-efficient and the root mean squared error (°C) between the data is annotated. Red symbols and text refer to summer (JJA) values, black symbols and text refer to annual values. Black dotted line denotes a 1:1 fit.

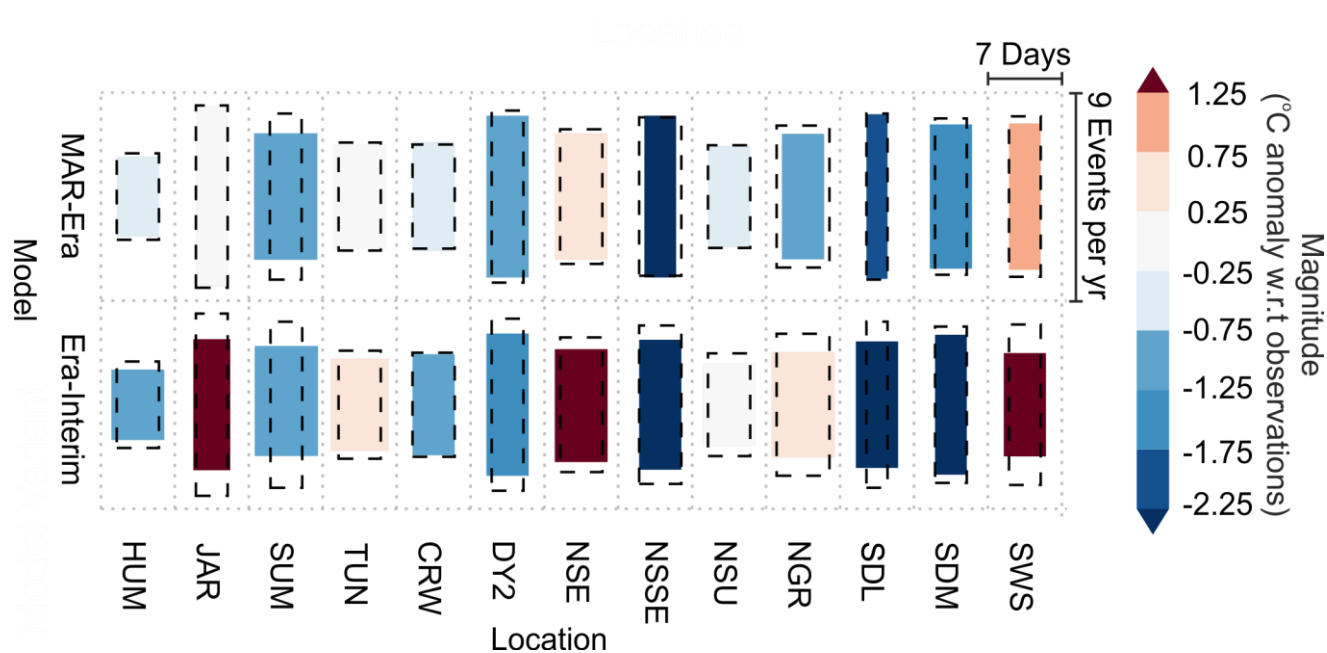


Figure 54. Median frequency, duration and magnitude of extreme temperature events as simulated by Era-Interim and MAR-Era. As before, frequency is denoted by the height of each box, duration is indicated by the width of each box and observed values are given by the dashed black boxes. Box colours indicate the departure of the modelled magnitude from the observed value, blue colours indicate an underestimate, red colours indicate an over estimate. Note that a temperature lapse rate has been applied to modelled temperatures to account for differences in elevation between AWS and Era-Interim elevation and AWS and MAR elevation.

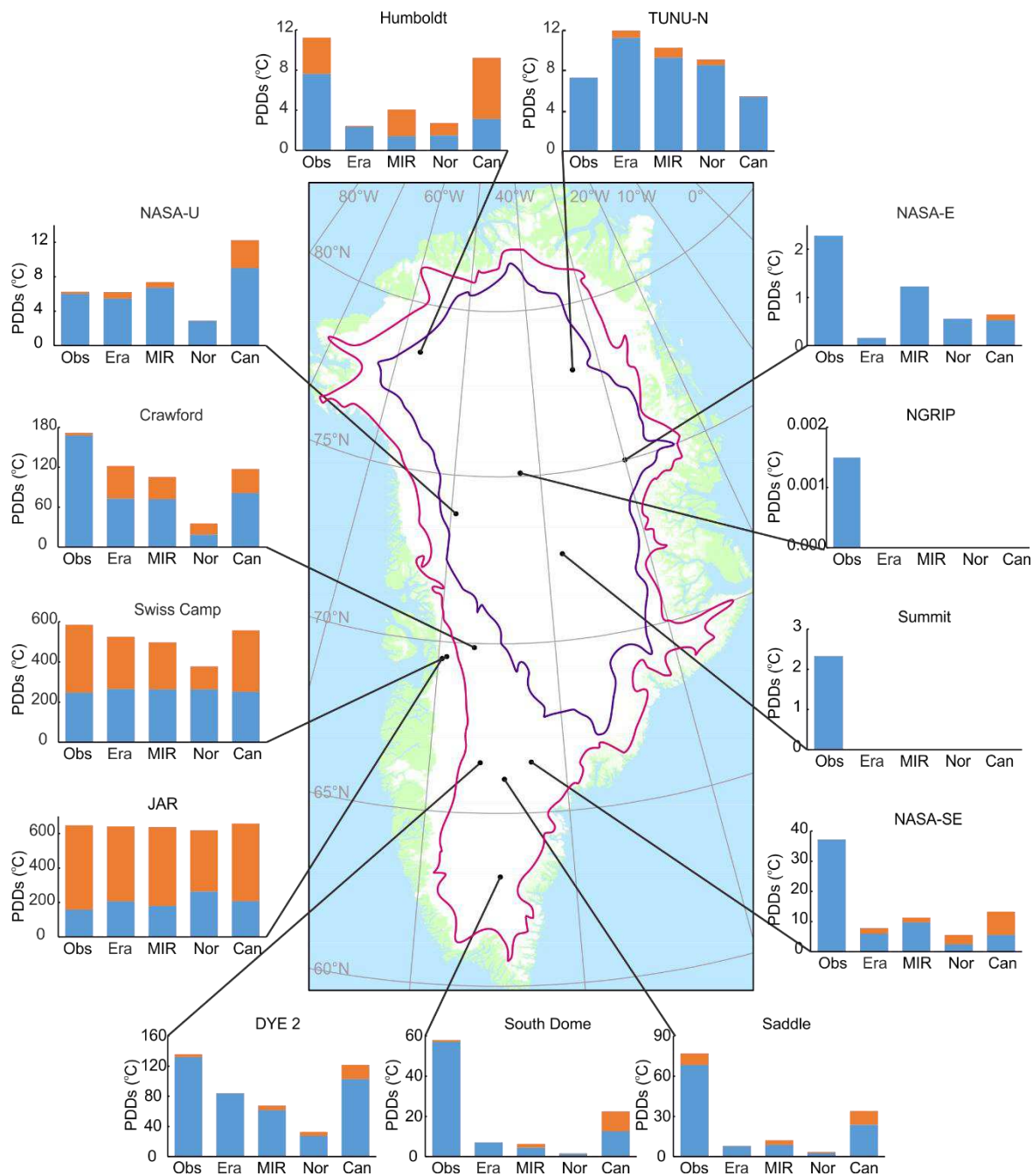


Figure 65: Total positive degree days observed and modelled at each location during the study period. Stacked bars represent total values. Blue and orange coloured portions represents contributions during extreme (non-extreme) events and non-extreme conditions respectively. Observed values are labelled Obs, model variants are labelled as Era – MAR-Era, MIR – MAR-MIR, Nor – Mar-Nor and Can – MAR-Can. Contours on map represent lower boundary of percolation zone (pink) and dry snow zone (purple). Zones are defined as in McMillan et al., (2016).

	Melt Zone	N years of data	MAR m	Gc-net m	MAR-Gc-net m	Lapse rate corr °C
Summit	Dry Snow	16.629	3201	3254	-53	-0.38
JAR	Ablation	15.939	932	962	-30	-0.21
Humboldt	Dry Snow	12.448	2020	1995	25	0.18
TUNU-N	Dry Snow	17.3820	2028	2113	-85	-0.60
Swiss Camp	Ablation	2415.47	1245	1149	96	0.68
Crawford	Percolation	13.1647	1920	2022	-102	-0.72
NASA-U	Dry Snow	4612.48	2293	2369	-76	-0.54
DYE2	Percolation	2419.58	2097	2165	-68	-0.48
Saddle	Percolation	2017.70	2475	2559	-84	-0.60
South Dome	Percolation	4915.89	2833	2922	-89	-0.63
NASA-E	Dry Snow	2018.13	2606	2631	-25	-0.18
NGRIP	Dry Snow	97.40	2921	2950	-29	-0.21
NASA-SE	Percolation	4411.55	2331	2425	-94	-0.67
JAR2	Ablation	4510.85	252	568	-316	-2.25

410 **Table 1: Discrepancy between elevation of AWS and elevation of the closest grid cell in MAR. Number of years of data (excluding gaps) is also given.** A lapse rate based correction to MAR temperature data (assuming 0.71°C per 100 m Steffen and Box (2001)) is also given; negative values occur when the grid cell in MAR is lower down in the atmosphere, i.e. is too warm.

	Humboldt	Jar	Summit	Tunu	Crawford	DYE2	JAR2	NASA-E	NASA-SE	NASA-U	NGRIP	Saddle	South Dome	Swiss Camp
Number of Events	71	134	122	84	81	134	106	99	89	71	43	122	115	125
Frequency (#/yr)	3.74	7.88	7.18	4.67	4.5	7.44	8.15	5.82	6.85	4.44	6.14	7.18	6.76	6.94
Maxima (°C)														
Minimum	-3.23	3.23	-8.16	-4.75	-0.43	-0.3	4.9	-7.38	-2.14	-3.1	-8.73	-2.23	-3.08	2.45
25% quantile	-1.64	3.55	-7.04	-3.24	0.13	0.53	5.37	-6.22	-0.86	-2.08	-7.05	-1.27	-2.04	2.94
Median	-0.64	4.1	-5.07	-2.14	0.95	1.32	5.94	-4.19	0.55	-1.04	-5.7	0.16	-0.36	3.43
75% quantile	0.02	4.85	-2.67	-0.41	2.02	1.97	6.63	-2.3	2.14	0.18	-3.87	1.41	0.88	4.36
Maximum	3.07	9.89	3.62	2.9	12.68	5.16	9.59	3.63	5.52	2.05	0.04	9.18	10.84	7.77
Duration (days)														
Minimum	1	1	1	1	1	1	1	1	1	1	1	1	1	1
25% quantile	2	1	1	2	1	1	1	1	2	2	2	1	1	1
Median	4	3	3	4	4	3	3	4	4	4	5	2	3	3
75% quantile	8	5	10	9	7	7	5	9	8	8	8	8	8	6
Maximum	42	46	31	54	67	36	22	59	29	35	29	39	37	34

Table 2: Frequency, duration and magnitude of extreme events observed at each of the 14 stations.

	ALL nyrs = 233		Ablation nyrs = 40		Percolation nyrs = 91		Dry Snow nyrs=102	
	$\bar{T}$	$\frac{dT}{dt}$	$\bar{T}$	$\frac{dT}{dt}$	$\bar{T}$	$\frac{dT}{dt}$	$\bar{T}$	$\frac{dT}{dt}$
All data								
MAR-Can	-0.13	0.03	-1.17	-0.05	-0.06	-0.01	0.22	0.11
MAR-MIR	1.19	0.02	-0.24	-0.08	1.11	-0.02	1.81	0.09
MAR-Nor	-0.21	0.03	-1.41	-0.02	-0.46	-0.01	0.49	0.09
MAR-Era	0.86	-0.03	-0.34	0.03	0.66	-0.06	1.50	-0.02
JJA only								
MAR-Can	-0.01	0.01	-0.02	-0.05	0.27	-0.01	-0.26	0.04
MAR-MIR	-0.64	-0.04	-0.56	-0.07	-0.62	-0.07	-0.68	0.00
MAR-Nor	-1.58	0.04	-1.55	0.00	-1.57	0.06	-1.60	0.04
MAR-Era	-0.35	0.01	-0.42	0.02	-0.03	0.01	-0.62	0.01

Table 3: Modelled-observed values for mean ($^{\circ}\text{C}$) and rate of change ($^{\circ}\text{Cyr}^{-1}$) of mean daily temperature during our study period. Values given in red (blue) denote an over (under) estimate of temperature in general; i.e. the model is too warm (cold). The number of years of data in total (i.e. the sum of the number of years of data at each station) is also identified (nyrs).

		All data					Extreme Events					Non Extremes				
		Observed	Era-Interim	MIROC5	Nor-ESM1	Can-ESM2	Observed	Era-Interim	MIROC5	Nor-ESM1	Can-ESM2	Observed	Era-Interim	MIROC5	Nor-ESM1	Can-ESM2
Dry snow zone (0.8mil Km ²)	PDDs (°C)	4.91	3.50	3.82	2.40	4.34	4.26	3.21	3.11	2.25	3.01	0.65	0.29	0.71	0.16	1.33
	Bias (%)		-29	-22	-51	-11		-25	-27	-47	-29		-55	10	-76	107
Percolation zone (0.61mil km ²)	PDDs (°C)	96.04	45.84	40.41	15.10	57.74	92.55	35.53	31.27	10.51	45.20	3.49	10.31	9.14	4.60	12.54
	Bias (%)		-52	-58	-84	-40		-62	-66	-89	-51		195	162	32	259
Ablation zone (0.23mil km ²)	PDDs (°C)	616.52	583.38	544.84	479.68	573.57	202.65	236.42	221.85	263.34	229.68	413.87	346.95	322.98	216.34	343.89
	Bias (%)		-5	-12	-22	-7		17	9	30	13		-16	-22	-48	-17
All Greenland (weighted)	PDDs (°C)	124.41	100.46	93.20	73.99	103.91	64.80	47.88	44.21	41.90	50.42	59.60	52.58	48.99	32.10	53.49
	Bias (%)		-19	-25	-41	-16		-26	-32	-35	-22		-12	-18	-46	-10

Table 4: Total positive degree days during the study period. Per station averages are given for the dry snow (six stations), percolation (five stations) and ablation (two stations) zones. These values are weighted by relative area to give a value for the whole of Greenland. The area of each zone is given in each row label, and is in units of million square kilometres.