Except figure 2b (blue line) shows the % of ice sheet surface was higher in 2012/2013, 2003/2004, and at the end of time series i.e 2019. Melting index was very low in 1980-1985, and during 2012-2017. It can be clearly shown the parameters (atmospheric/ice) those are exceptional in 2019 either in the form of time series graphs or maps. Other-wise, it’s very good work. The manuscript is written well.

R: We thank the colleague for the comment
This paper focuses on the atmospheric conditions in summer 2019 over Greenland that led to record or close-to-record values of SMB, runoff and snowfall. The topic is of great contemporary interest as these extreme melt events greatly impact the mass balance of Greenland and thus sea level rise. The authors are the first to present this kind of data for the year 2019.

Overall, I highly appreciate the rapid investigation of this recent event. The paper is well written but some small parts in the result section need clarification (see comments). The authors show very well the exceptional character of the 2019 melt season for different properties (runoff, SMB, melt extent, etc).

However, as the authors write themselves, that anticyclonic conditions increase SMB loss is known so a more detailed comparison to the 2012 melt season and the melt seasons in general to post 1990 would improve this paper and add to our scientific understanding. Possibly there can be drawn some more conclusions.

Main comment:

For instance (see comments below): How does Figure 10 differ in the case for 2012, JJA or averaged post 1990, JJA. This anticyclone on the west of Greenland seems to be typically for the time post 1990, (Fig. 4 Noël et al., Sci. Adv. 2019) also prominent in the year 2012 (Tedesco, 2013). Noel et al. describe similar changes in cloud anomaly for the post 1990 as the authors did for the 2019 event. How does the 2019 event compare to that. Is it “just” more persistent or does it differ in position? How to the year 2012? The authors write that persistence was a major driver in 2019, so a comparison to the persistence of the other years would be beneficial (e.g. Plot of a timeline). Can we see a general trend in persistence there? How persistent was 2012?

R: We have replied to this point below when addressing the comments. It is not clear, also, to us what the reviewer means with a “trend” of persistence. We apologize for this and would be more than glad to address this point once clarified.

Minor comments:

P3, L 92. What do you mean by that: The overall integrity of the long-term GBI time series is ensured by using homogeneity adjustments (Hanna et al. 2016). Please explain in the text

In Hanna et al. (2016), the reanalysis 20CRv2 is used in addition of the NCEP-NCARv1 reanalysis to provide an "homogenous" GBI time series over the whole 20th century by combining 20CRv2 to NCEP-NCARv1 with the help of some adjustments. Here, we discuss only the NCEP-NCARv1 based GBI index from 1948 (which is the JJA mean Z500 over 60-80°N, 20-80°W) and therefore this sentence is out of context here and we deleted it in the revised version of our manuscript.
P4, L 100. Since you refer to Fig2a, could you please either mention the day number in brackets for the dates in
the text or change the axis label to actual dates.

R: We have added the corresponding day of the year

P4, L 100-102. Maybe make 2 sentences of this one? The explanation in the brackets is too lengthy and reduces
the readability. And how do you come to the statements that the MI ranked 2nd . From Figure 2b red line it
doesn’t look like that. Or isn’t it showing the MI? Please clarify.

R: We have split the sentence and have corrected the fact that indeed 2019 ranks 3rd. thanks for catching this up.

P4, L 102-103. How do you see that? From Figure 2b I can only see the seasonal value.

R: The reference to Fig2b was wrong. Thanks for catching that up.

P4, L 108-110. I think the phrasing is a bit confusing and I needed to read the sentence several times. So are you
saying that end of July 73% of the ice surface were melted and the following two days the remaining ~ 23%?
What do you mean by the same atmospheric conditions? Is the anticyclone moving or are the feedbacks
increasing the melt area? Please clarify.

R: we rewrote the sentence as follows:

Indeed, the persistency of the atmospheric conditions at the end of July that were responsible for promoting melting over 73%
of the ice sheet in a single day (July 31, 2019) was extending melting during the next few days over regions that were not
originally involved in the melting on July 31 with cumulative melt extent for the 3-day period (July 31 – August 2) reaching
up to ~ 97% of the ice sheet surface.

P4, L 111 -113. Please make a reference here.

R: Done, thanks !

P4, L113-114. You write the air masses in 2019 came from the east. For completeness, can you please add, where
did the air masses in 2012 come from?

R: We mention this in the preceding sentence.

P4, L115-118. Interesting, but, -since it is in the result section- do you have the data or Figures showing that?

Bellow, we can see the absolute values of the temperature at 700hPa (T700), 500hPa (T500) and mean specific humidity
over 700-500hPa from NCEP-NCARv1 reanalysis on the 12th of July 2012 and on the 31st of July 2019. While the
temperature anomalies were higher on 31-Jul-2019 than on 12-Jul-2019 with respect to the climatology of Mid-July or of the
end of July, the absolute values were higher in 2012 than in 2019. In addition, we can see that the humidity content was also
higher in 2012 than in 2019 over the ice sheet, showing the important role of liquid clouds in the 2012 extreme melt event.
(Bennartz et al., 2013). On this figure, we can finally see that the hot air bubble was centred over the ice sheet in 2012 but was rather along the north-east coast in 2019. These differences in temperature and humidity pattern explain why the 2012 highest melt event was more extreme than the 2019 one.


You say that persistence was a major driver in 2019. Was is more than in 2012? I like your definition of persistence showing a time line for 1948-2019 would be interesting. Can we see a trend?

R: we apologize but this question is not clear to us. The persistency of the 5 top 2019 atmospheric patterns is shown in Fig. 14. We are not sure the trend of which variable the reviewer is asking. Apologies again but this is not clear to us.

Also how does Figure 10 differ in the case for 2012 or post 1990 JJA. This anticyclone on the west of Greenland seems to be typically for the time post 1990, (Fig, 4 Noël et al., Sci. Adv. 2019) also prominent in the year 2012 (Tedesco, 2013). It would be nice to see how the position of the anticyclone differs in respect to post 1990 or 2012.

R: We added a new figure (Fig. 10, shown above) to highlight the differences between the 2012 and 2019 years and added the following text:
As a reference, Fig. 11 shows the absolute values of the temperature at 700hPa (T700), 500hPa (T500) and mean specific humidity over 700-500hPa from NCEP-NCARv1 reanalysis on the July 12th 2012 and on the July 31st 2019. While the temperature anomalies were higher in 2019 with respect to the climatology of Mid-July or of the end of July, the absolute values were higher in 2012 than in 2019. In addition, the humidity content was also higher in 2012 than in 2019 over the ice sheet, showing the important role of liquid clouds in the 2012 extreme melt event (Bennartz et al., 2013). These differences in temperature and humidity pattern explain why the 2012 highest melt event was more extreme than the 2019 one.

Is it possible to check, whether the occurrence of any of the 28 classes correlated with the GBI anomaly or the SMB anomaly? Could you identify any significant trend?

R: This is a very interesting point and we did, indeed, look at this aspect. We found a modest correlation between the GBI and the frequency of SOMs nodes characterized by strong anticyclonic conditions. This is expected in view of the definition of GBI. The strength of the correlation can change year by year. We think that this (together with the “modest” correlation) is due to the fact that the GBI is an integrated measure of GPH anomalies over a specific geographic region where the atmospheric patterns identified with the SOMs extend well beyond that region. When we looked at the correlation between GBI / SOM frequency and melting we found similar correlations (summer integrated values). We are not sure if the reviewer is asking for trends in the GBI or other quantities.

Figures:

Fig 3a: Maybe put “≤ 5” on the color bar, otherwise the reader might get the impression there are only 5 melting days.

R: Done.

Fig 14a: Could you please add a grid, otherwise the reader gets lost when searching for a specific class in a specific year.

R: we tried to add a grid but the figure became hard to read. As an alternative, we have increased the resolution of the x-axis to improve readability.

Typos:

P3, L 176. “.” after “(”

R: Done.

P3, L 177. check “); .”

R: Done.

P6, L167. Something went wrong in the sentence order.

R: thanks for that. We fixed the sentence,
References:

Noël et al., Rapid ablation zone expansion amplifies north Greenland mass loss, Science Advances, 5, doi: 10.1126/sciadv.aaw0123

Anonymous Referee #2

Received and published: 22 January 2020

The authors show that summer 2019 was an exceptional melt season for Greenland, with record or near-record values in runoff, snowfall, and SMB. They find that summer 2019 was characterized by persistent anticyclonic conditions and melting was enhanced by melt-albedo feedback and warm air advection. Comparing 2019 to the extreme melt season of 2012, they show that although the two years have similar values of runoff and SMB, these exceptional conditions were driven by different atmospheric circulation patterns.

This study provides valuable insights into this latest Greenland melt season within the context of long-term trends. Overall, the analysis is thorough and well done, and the manuscript is well written.

— Main Comments —

In the SOM analysis, I’m concerned that the persistency of atmospheric patterns is defined somewhat arbitrarily and the exceptional persistency of 2019 may be overstated. In lines 246-249, 257-259, and Fig. 14c, the cumulative number of days for the top 5 most frequent classes of 2019 is compared to the cumulative number of days of these same classes in other years and is found to be much higher. But this seems to be an inevitable result by construction, since the top 5 classes vary from year to year, so we’re comparing 2019’s top 5 classes to lesser ranked classes in other years. If these 5 classes were all adjacent on the SOM map, and represented some broader category of circulation pattern, then I could see how grouping them together is physically meaningful for comparison across years. But in this case, the grouping seems artificial and perhaps not a very robust approach to comparing across years.

For example, in Fig. 14c, the cumulative number of days for the 5 classes appears to be approximately 25 in 2012, much lower than the 55 days for these classes in 2019. However, we can see from Fig. 14b that 2012 had high persistency in atmospheric patterns as well, just with different classes than those in 2019. From Fig. 14b, I estimate the cumulative number of days for 2012’s top 5 classes is approximately 47, which is pretty close in magnitude to the 55 days for 2019’s top 5 classes. Thus, I would conclude that both 2012 and 2019 probably had high persistency, contrary to the analysis presented here.

If you were to repeat the analysis but instead compute the cumulative number of days for each year based on the top 5 classes in that specific year, then how do the results change? Is the persistency of atmospheric conditions in 2019 as exceptional as stated? It would be interesting to see if both 2012 and 2019 stand out as exceptional in this approach.

R: thanks for the comment. We have added the following text at the end of the Results to highlight the similarities between the 2012 and 2019 summers:

[Additional text added in response to the referee's comments]

6
Nevertheless, we observe from Fig. 14b that 2012 also had high persistency in atmospheric patterns though such patterns belong to different classes than those in 2019. For example, the cumulative number of days for 2012’s top 5 classes is 47, being similar in magnitude to the 55 days for 2019’s top 5 classes. Moreover, similarly to 2019, the top five classes in 2012 were all characterized by high GPH anomalies and strong anticyclonic conditions (though different in terms of spatial distribution of the GPH anomalies). In this regard both 2012 and 2019 can be assumed to be exceptional from an atmospheric point of view.

Concerning the other years: we performed the analysis suggested by the reviewer of looking at the top 5 (or even 4 or 6) patterns (in terms of frequency) and found that 2012 and 2019 were the years with the highest occurrence of the top 5 patterns. Moreover, we looked at the top 5 nodes (e.g., patterns) for the remaining years and used the mean GBI (as well as the cumulative sum) to quantify the anticyclonic conditions and found that, also in this case, the top 5 patterns were not characterized by anticyclonic conditions as strong as the ones in 2012 and 2019. Said this, we do acknowledge that there is extra work that needs or can be carried out to identify “extreme” atmospheric patterns or their persistency and we plan to carry out a deeper analysis in our future work and are thankful to the reviewer for his/her suggestions.

— Minor Comments —

L21-22 “the total number of days with the most frequent atmospheric pattern that characterized the summer of 2019 was 5 standard deviations above the 1981 – 2010 mean”: This seems misleading, since the number of days referred to here is the cumulative total for the top 5 most frequent patterns of 2019, not the single most frequent pattern.

R: We changed the sentence as follows: The analysis of the frequency of daily 500 hPa geopotential heights obtained from artificial neural networks shows that the total number of days with the five most frequent atmospheric patterns that characterized the summer of 2019 was 5 standard deviations above the 1981 – 2010 mean, confirming the exceptional nature of the 2019 season over Greenland.

L102-103 “When looking at the different summer months separately, the MI values in 2019 ranked 5th in June, 7th in July and 9th in August (Fig. 2b)”: Fig. 2b only shows summer averages, not monthly values, so perhaps the reference to it should be removed?

R: this was removed. Thanks.

L109-110 “was also responsible for the cumulative 3-day melt event”: Perhaps change “the cumulative...” to “a cumulative...”, because this 3-day melt event hasn’t been previously introduced. It would also be helpful to mention the specific dates of this event.

R: Done, thanks.

L112-118: Is this analysis of air mass trajectories all from the current study and details are not shown here? In particular regarding the 2012 summer, it’s not clear if the discussion is summarizing earlier studies which should be cited here, or if it’s referring to current analysis.
R: We have added a reference to the first sentence to clarify the fact that it refers to a published study. Thanks for the suggestion.

L167-168: The switch to positive albedo anomalies is confusing here, since the rest of the discussion centers around negative albedo anomalies. Also, looks like a typo here “In June, only 23%... was 23%.”

R: Corrected, thanks.

L188-189: The text refers to “geopotential height anomalies”, but the values listed (5567 m, etc.) seem to be actual geopotential heights, rather than anomalies.

R: thanks for catching this up. We have corrected the error. It should be simply “geopotential heights”.

L197 “high pressure system centered near Summit over the whole 2019 summer (Fig. 5a)”: I think Fig. 10b should be referenced here, rather than Fig. 5a.

R: That is correct. We have added the reference to the right figure. Thanks.

L198: I think Fig. 5e should be referenced here, rather than Fig. 5d.

R: That is correct. Thanks.

L205-206: Should the area integrated anomalies be reported in units of W rather than W/m²?

R: We prefer to list our anomalies in W/m² as using W only is very dependent of ice sheet mask/resolution used, which we want to avoid according to the recommendations of Fettweis et al. (2020):


L218 “We classify the daily 500 hPa GPH”: Should specify that it’s the GPH anomalies that are being classified.

R: we added “anomalies” here and in the successive sentence.
L220-221: This description of the training phase with “existing SOM nodes” seems a bit off. The SOM nodes are defined in an iterative process during training – they don’t exist prior to training.

R: we added more explanation on the fact that the nodes are initially random structure in which the data is allocated.

L218-221: Were the input data fields weighted to account for grid cell area variation at high latitudes (for example, as in Mioduszewski et al. 2016)?

R: Yes.

L229-230: Can you explain in more detail how the 4x7 SOM shape was selected? Were any sensitivity tests performed to determine the impact of SOM size / aspect ratio and analyze error metrics?

R: As we mention in the text, the selection of the number of nodes and the architecture of the SOM does not have specific rules but mostly directions. A first aspect to be considered is the computational time, as the network will have to train a large number of points (e.g., daily data for the past 60 summers gridded over the Arctic). One aspect also to account for is the fact that the number of elements to be classified are roughly the same within each node. We based the selection of our architecture on these two considerations by testing several configurations and building on our experience published in Mioduszewski et al, 2016. Once the architecture was selected, we trained 50 different SOMs and compare their outputs to assess the effect of the random initialization on the 50 different runs. We found that the differences in classification for the multiple SOMs was less than 0.1 %. We also tested our configuration using ad-hoc prepared training datasets with known shapes and patterns and tested the selected configuration. Also in this case, the error in terms of classification was smaller than 0.1 %.

L263 “frequency and occurrence of the atmosphere”: What does this mean?

R: Thanks for pointing that out. We rewrote the sentence as follows:

In the future, we plan to analyse how the frequency and occurrence of GPH anomalies has been changing at higher levels (e.g., 300 hPa, 100 hPa) to eventually quantify potential missing links between the stratosphere and the troposphere that might be responsible for the exceptional conditions.

— Figures —

Figure 1b: The colour scheme is inconsistent with Fig. 1a, Fig. 3a, and Fig. 5a, which all use red for more melting days and blue for fewer melting days. I recommend reversing the colour scheme in Fig. 1b to be consistent with the others.

R: we have reversed the colorbar.

Figure 2b: The caption describes the blue line as “Summer-averaged melt extent”, but this is not a summer-averaged quantity, is it? It looks like the blue line shows, for each summer, the overall area subject to at least one day of melting.

R: we have reversed the colorbar.
Figure 5: Captions for subplots (b)-(e) are mixed up (i.e., (b) is snowfall anomaly but caption says 2m temperature, etc.). It would also be helpful to add a bit more horizontal space between subplots, so that there is some space between the colour bar labels and the y-axis of the right-adjacent subplot.

R: we corrected the caption and added the space.

Figure 9: The annotation reads “Melt extent reaches ~97%”, whereas the main text reports this value as 96%.

R: we corrected the main text for the right value of 97%.

Figure 10a: In the caption “Anomaly of the JJA 2019 averaged geopotential height anomalies”, extra “anomaly” should be removed.

R: Done, thanks

Figure 11: Typo in the y-axis label: “yStandardized”.

R: Adjusted, thanks

Figure 13: The subplots are very tiny. Can these be enlarged? Also, 3-4 decimal places in the average geopotential heights seems excessive - they could be rounded to 0 or 1 decimal place in these annotations.

R: We have removed the decimal place and enlarged the maps.

Figure 14c: Caption and y-axis label describe this data as anomalies in the cumulative number of melting days, but the values shown aren’t anomalies. Also, are they the cumulative number of melting days, or just cumulative number of days (melting or not)?

R: thanks for this. We changed the caption accordingly.

— Typographic Corrections — Punctuation / spacing typos:
- L45 “i.e., Kohonen”
- L76 “eg. Fettweis”
- L77 “2011); .”
- L90 “2015-2018;”
L115 “relative cold” ==> “relatively cold”
L259-260 “Despite similar in terms of runoff and SMB” ==> “Despite being similar . . .”
- R: Done for all, thanks!

R: that is correct. We have rewritten the caption.
Unprecedented atmospheric conditions (1948 – 2019) drive the 2019 exceptional melting season over the Greenland ice sheet

Marco Tedesco\textsuperscript{1,2} and Xavier Fettweis\textsuperscript{3}

\textsuperscript{1}Lamont-Doherty Earth Observatory, Columbia University, Palisades, NY 10964, USA.
\textsuperscript{2}NASA Goddard Institute for Space Studies, New York, NY 10025, USA.
\textsuperscript{3}Department of Geography, University of Liège, Liège 4000, BELGIUM

Correspondence to: Marco Tedesco (mtedesco@ldeo.columbia.edu)

Abstract.

Understanding the role of atmospheric circulation anomalies on the surface mass balance of the Greenland ice sheet (GrIS) is fundamental for improving estimates of its current and future contributions to sea level rise. Here, we show, using a combination of remote sensing observations, regional climate model outputs, reanalysis data and artificial neural networks, that unprecedented atmospheric conditions (1948 – 2019) occurring in the summer of 2019 over Greenland promoted new records or close-to-record values of SMB, runoff and snowfall. Specifically, runoff in 2019 ranked second within the 1948 – 2019 period (after 2012) and first in terms of surface mass balance negative anomaly for the hydrological year September 1, 2018 – August 31, 2019. Summer of 2019 was characterized by an exceptional persistence of anticyclonic conditions that, in conjunction with low albedo associated with reduced snowfall in summer, enhanced the melt-albedo feedback by promoting the absorption of solar radiation and favored advection of warm, moist air along the western portion of the ice sheet towards the North, where the surface melt has been the highest since 1948. The analysis of the frequency of daily 500 hPa geopotential heights obtained from artificial neural networks shows that the total number of days with the five most frequent atmospheric patterns that characterized the summer of 2019 was 5 standard deviations above the 1981 – 2010 mean, confirming the exceptional nature of the 2019 season over Greenland.

1 Introduction

Understanding the role of atmospheric circulation changes on the surface mass balance (SMB) of the Greenland ice sheet (GrIS) is crucial for improving estimates of its current and future contribution to sea level changes and for studying recent mass loss trends in the context of multi-decadal timescales. Atmospheric patterns modulate the GrIS mass balance through snowfall and runoff (e.g., Hanna et al., 2008, 2013, 2016; Tedesco et al., 2011, 2016a, 2016b) as well as radiative forcing and surface turbulent heat fluxes (e.g., clouds, longwave and shortwave radiation). Recent studies (e.g., Hanna et al.,
2014; Mattingly et al., 2016; McLeod and Mote, 2016) have focused on linking the observed variability of climate indices such as the North Atlantic Oscillation (NAO, i.e., Hanna et al., 2015) or the Greenland Blocking Index (GBI, Hanna et al., 2018) to the recent changes in runoff and accumulation over Greenland. Other studies (i.e., Tedesco et al., 2016b) have recently pointed out to the increased frequency of persistent anticyclonic conditions favoring atmospheric blocking and explaining most of the recent surface melt increase (Fettweis et al., 2013).

In this paper, we report the results of an analysis of SMB and surface energy balance (SEB) components obtained from satellite data and model outputs for the summer of 2019, their linkages to anomalies in the atmospheric circulation and analyze them within the long-term context (1948 – 2019). Specifically, we use spaceborne passive microwave data collected between 1979 and 2019 at 19.35 GHz, horizontal polarization, for detecting melting following the approach reported in Tedesco, 2007; Tedesco et al., 2007 and Tedesco, 2009). We also use estimates of broadband albedo derived from data collected by the Moderate Resolution Imaging Spectroradiometer (MODIS) for the period 2000 – 2019 (https://terra.nasa.gov/about/terra-instruments/modis). We complement satellite data with the outputs of the Modèle Atmosphérique Régionale (MAR) regional climate model (RCM, Gallée and Schayes, 1994; Gallée, 1997; Lefebre et al., 2003) forced by National Centers for Environmental Prediction/National Center for Atmospheric Research (NCEP-NCARv1, Kalnay et al., 1996) reanalysis dataset over the period 1948 – 2019. We lastly make use of self-organizing maps (SOMs, i.e., Kohonen 2001) to classify pan-Arctic summer 500 hPa geopotential height (GPH) anomalies (1981 – 2010 baseline period) also obtained from the NCEP/NCAR reanalysis dataset (Kalnay et al., 1996) between 1948 and 2019. The pan-arctic region is here defined as the portion of the northern hemisphere poleward of 60º N. We focus on the 500 hPa GPH values because of their strong correlation with SMB quantities and for consistency with other studies using them to compute climate indices, such as the GBI (e.g. Hanna et al., 2016). Moreover, 500 hPa is also a standard height for gauging the effects of jet stream blocking on synoptic weather patterns (e.g. McIlveen, 2010).

2 Methods and data

2.1 Satellite data

Passive microwave (PMW) brightness temperatures (Tbs) are a crucial tool for studying the evolution of melting over the Greenland and Antarctica ice sheets (e.g. Abdalati and Steffen, 1995; Tedesco, 2007; Tedesco et al. 2009; Tedesco 2009; Fettweis et al., 2011). The capability of passive microwave sensors to collect useful data during both day- and nighttime and in all-weather conditions provides data at a high temporal resolution (at least daily over most of the Earth), with high latitudes being covered several times during a single day. Since the launch of the Scanning Multichannel Microwave Radiometer (SMMR) in October 1978, Tb data is available in multiple bands every other day (in the case of SMMR) and daily starting in 1987, with the launch of the Special Sensor Microwave Imager (SSMI). PMW brightness temperature records are the longest available time series and an irreplaceable tool in climatological and hydrological studies, especially for those regions, such as the ice sheets, where in-situ observations lack and fieldwork is logistically difficult, if not
impossible. Specifically, we make use of data distributed by the National Snow and Ice Data Center (NSIDC, https://nsidc.org/https://catalog.data.gov/dataset/near-real-time-dmsp-sm5-smmi-pathfinder-daily-ease-grid-brightness-temperatures-version) at a spatial resolution of 25 km at K band (≈ 19 GHz), horizontal polarization. Melting is detected following the procedure described in Tedesco (2007) and Tedesco (2009).

We complement PMW data with the MODIS daily surface reflectance product (MOD09GA Version 6) and daily snow cover product (MOD10A1 Version 6, https://nsidc.org/sites/nsidc.org/files/files/MODIS-snow-user-guide-C6.pdf). The MOD10A1 data include broadband albedo estimated based on the MOD09GA product. We used the Version 6 data in view of its improvement in sensor calibration, cloud detection, and aerosol retrieval and correction relative to version 5 (e.g., Casey et al., 2017). Version 6 data are optimal for assessing temporal variability of surface albedo as they are corrected for sensor degradation issues impacting earlier versions (Casey et al., 2017). The spatial resolution of the MODIS datasets is 500 m. We use the cloud mask in the MOD10A1 data to exclude clouds.

2.2 The MAR regional climate model

The MAR regional climate model (Fettweis et al., 2017) combines atmospheric modeling (Gallée and Schayes, 1994) with the Soil Ice Snow Vegetation Atmosphere Transfer Scheme (De Ridder and Gallée, 1998) and has been extensively evaluated and used to simulate surface energy balance and mass balance processes over GrIS (e.g., Fettweis, 2007; Fettweis et al., 2011). In this study, we use the version 3.10 of MAR, at a horizontal spatial resolution of 20 km as in Fettweis et al. (2017) and 6 hour temporal resolution forced with the NCEP-NCAR v1 reanalysis (Kalnay et al., 1996). Outputs generated at sub-daily temporal resolution are then, averaged to obtain daily values. We refer to Fettweis et al. (2017) for the evaluation of this NCEP-NCAR v1 forced simulation and to Delhis et al. (2019) for the list of improvements made since MARv3.5 used in Fettweis et al. (2017).

2.3 NCEP/NCAR reanalysis data and the Greenland Blocking Index (GBI)

We use geopotential heights at 500 hPa obtained from the NCEP/NCAR reanalysis dataset, consisting of globally gridded data that incorporate observations and outputs from a numerical weather prediction model from 1948 to present (Kalnay et al., 1996). We also use the so-called Greenland Blocking Index (GBI), defined as the mean 500 hPa geopotential height over the area bounded by the following coordinates 60°-80°N, 20°-80°W (e.g., Hanna et al., 2015, 2018). Positive GBI conditions are generally associated with surface high pressure ‘blocking’ anomalies over the Greenland region (Hanna et al., 2016). There is also a strong and significant anti-correlation between Greenland blocking and the North Atlantic Oscillation (NAO, the first mode of atmospheric surface pressure variation over the North Atlantic), with Greenland blocking typically linked to a southward deflection of the jet stream (Hanna et al., 2015, 2018; Tedesco et al., 2016b). Here, we use a recent reconstruction of GBI from 1851-2019 (Hanna et al., 2018) that combines data from the 20CRV2c Reanalysis (Compo et al., 2011) with newer (1948-2015) data from the NCEP/NCAR reanalysis (Kalnay et al. 1996).
3 Results

Melt duration in 2019 (Fig. 1a) estimated from PMW data exceeded the long-term (1981 - 2010) mean by up to 40 days along the west portion of the ice sheet where dark, bare ice is exposed (Fig. 1b). Over the rest of the ice sheet, the anomaly of the number of melting days during the summer of 2019 from PMW data was around 20 days. Negative anomalies were rare and geographically concentrated over a small area in the southern portion of the ice sheet. Surface melting in 2019 started relatively early, around mid April (Fig. 2a, day of the year, DOY 105), and exceeded the 1981 – 2010 mean for ~ 82 % of the days during the period June 1 – August 31, 2019 (DOY 152 – 244). A measure that is commonly used for quantifying melting from passive microwave observations is the so-called melting index (MI), defined as the number of melting days times the area undergoing melting and being a measure of the intensity of surface melting (Tedesco, 2007). In 2019, the MI ranked third, after 2012 and 2010. When looking at the different summer months separately, the MI values in 2019 ranked 5th in June, 7th in July and 9th in August. The 2019 updated trends for MI and melt extent (here defined as the area subject to at least one day of melting) are, respectively, 78.836 Km²*day decade⁻¹ (p<<0.01, MI) and 7.66 % decade⁻¹ (p<<0.01, trend is here expressed as a percentage of the total area of the ice sheet). The maximum daily melt extent was reached on July 31, 2019 covering ~ 73 % of the ice sheet surface. In comparison, the average daily maximum extent from PMW data for the same day for the 1981–2010 period is 39.8%. Notably, the total area that at anytime underwent melting was 95.8 % of the total ice sheet in 2019 (Fig. 2b), against the 1981-2010 averaged value of 64.3 %.

Indeed, the persistency of the atmospheric conditions at the end of July that were responsible for promoting melting over 73 % of the ice sheet in a single day (July 31, 2019) was extending melting during the next few days over regions that were not originally involved in the melting on July 31 with cumulative melt extent for the 3-day period (July 31 – August 2) reaching up to ~ 97% of the ice sheet surface. We note that a similar value for the maximum melt extent was reached in 2012, though in this case it did happen in one day. As in 2019, the exceptional melt in 2012 was associated with the advection of a very warm and wet air masses coming from the south and promoting the presence of liquid water clouds promoting surface melt in the dry snow zone (e.g., Tedesco et al., 2016b). However, in 2019, the air mass came from the east after promoting an exceptional heat wave in Europe, being warmer and drier than the air mass in 2012. Moreover, by crossing the relatively cold Atlantic Ocean from Scandinavia, in 2019 the lower atmospheric layers cooled down increasing the stability of the air mass and then limiting the formation of liquid water clouds compared to July 2012, explaining why the melt extent was lower during this 2019 big melt event than in July 2012 while the temperature anomaly was higher in the free atmosphere in 2019 than in 2012.

We investigated the possibility that the sporadic melting detected at high elevations could have been due to a malfunctioning of the sensor or other issues related to data quality. Fig. 3a shows a map of the number of melting days constrained to values ranging between 1 and 4 days to highlight those areas where melting occurred for a few days at high elevations. In the figure, we also show the time series of brightness temperatures for those pixels where melting occurred for only one day (Fig. 3b) or for 2 days (Fig. 3c). The sharp, sudden increase of brightness temperatures is not associated with...
data quality issues but rather with the insurgence of melting in both cases. Melting at high elevations is also confirmed from the analysis of in-situ data. For example, Fig. 4a shows air temperature (2m) recorded at the EGP PROMICE station (75.6247°N, 35.9748°W, 2660 m a.s.l., https://www.promice.dk/WeatherStations.html) together with time series of spaceborne Tb's at 19.35 GHz, horizontal polarization, recorded over the pixel containing the location of the EGP station (blue line). Air (2m) pressure (hPa) recorded at the same station is also reported as a red line in the bottom plot. The figure shows that air temperature exceeded the value of 0°C when Tb values sharply increased from ~ 170 k to ~ 220 k.

Concurrently, surface air pressure reached at EGP peak values of ~ 749 hPa, likely as a consequence of the persistent anticyclonic conditions occurring during that period. We also note that air temperature exceeded the melting point at least twice in 2019 at the EGP station beside July 31, according to the in-situ data. The first time on day 163 (June 12) and the second time on day 201 (July 19). In both cases, however, the passive microwave data did not detect the presence of liquid water. This might be a consequence of the fact that air temperature can be exceeding the melting point when snow temperature is not and that the second event when air temperatures exceed the melting point was characterized by relatively low pressure, hence suggesting that the radiative forcing associated with the incoming solar radiation might have not been as strong as in the case of the end of July.

The spatial distribution of the anomaly of the number of melting days obtained from PMW observations is consistent with the one obtained from the MAR regional model, as shown in Fig. 5a. Here, we consider those cases when the integrated liquid water content in the top meter of the snowpack reaches or exceeds 1 mmWE, following Fettweis et al. (2007). Meltwater runoff in JJA 2019 simulated by MAR and integrated over the whole ice sheet ranked second (consistently with the MI values obtained from the PMW data), reaching a total of 560 Gt in 2019 against an average value of 300±85 Gt yr⁻¹ for the 1981-2010 period. As a reference, the value of runoff simulated by MAR for the JJA 2012 period (when the record was established) was 610 Gt. Despite ranking second in terms of surface runoff, the September 2018 - August 2019 (used to define the mass balance “year”) ranks first in terms of integrated SMB negative anomaly simulated by MAR, with a total surface mass loss anomaly of ~ 320 Gt yr⁻¹ with respect to the 1981-2010 SMB average, breaking the previous record established in 2011-2012 of ~ 310 Gt yr⁻¹ (Fig. 6, blue bars), though by only 10 Gt yr⁻¹. It is however important to note that such difference is below the uncertainty of the MAR model estimated to be 10% of the mean SMB.

The SMB negative anomaly in 2018-2019 is larger than 2011-2012 mainly because the 2018-2019 snowfall negative anomaly (~ -50 Gt) is larger in magnitude than the one that occurred during the 2011-2012 SMB year (~ -20 Gt), with large negative summer snowfall anomalies in 2019 occurring along the southern and western portions of the ice sheet (Fig. 5b). The early melt onset and the negative snowfall anomaly promoted the exposure of bare ice prematurely, hence further enhancing melting and runoff through the melt-albedo positive feedback mechanism (i.e., Tedesco et al., 2016b). This is evident from the analysis of summer broadband albedo simulated by MAR (Fig. 5c), showing negative anomalies down to -0.2 along the western portion of the ice sheet. These results are also confirmed by albedo estimates obtained from MODIS (Fig. 7a), indicating a large, negative albedo anomaly occurring along the west coast where bare ice is exposed. Specifically, summer MODIS albedo ranked 4th (Fig. 8) within the 2000 – 2019 MODIS period, being -1.45 standard
deviations (σ) below the mean (2000 – 2010 baseline period). The summer of 2019 precedes the ones of 2010 (-1.79σ), 2016 (-1.95σ) and 2012 (-3.33σ) in terms of MODIS albedo. When considering the summer months separately, June and July 2019 ranked, respectively, 10th (June) and 7th (July). A new record was, nevertheless, established in August 2019, with the absolute value averaged over the whole ice sheet reaching 77.51 % (-2.39σ) in 2019, followed by 2012 (77.86 %, -2.05σ) and 2016 (78.1 %, -1.81σ). The updated trends over 2000-2019 for summer broadband albedo is -0.4% decade¹, though it is not statistically significant (R² = 0.04). Similarly, the trends for June (-0.1 %), July (-0.6 %) and August (-0.7 %) are also not statistically significant.

The analysis of the maps of the monthly averaged albedo (Fig. 7b through d) indicates that, as mentioned above, negative albedo anomalies occurred along the western portion of the ice sheet in June and July but, during the same period, albedo was within the average over most of the rest of the ice sheet. In June, only 23 % of the ice sheet surface was showing positive albedo anomalies. The value for July was 25 %, to be reduced to only 6 % in August. During this month, the negative albedo anomalies in the south are confined along a relatively small portion of the west margin of the ice sheet, but they extend further inland, reaching high elevations in the northern regions (Fig. 7e). The presence of negative albedo anomalies in August at higher elevations is consistent with the sporadic melting that occurred over the same region at the end of July and beginning of August 2019 (Fig. 3). The impact of such event is, indeed, well observable in the albedo changes of the pixel that underwent melting for two days at the end of July (Fig. 9, being the same as the one whose Tb values are shown in Fig. 3b), showing a reduction from 87.4 % to 77.8 % due to the increase in grain size associated with the melting and refreezing cycle.

4 Discussion

A major driver of the exceptional melting season in 2019 was the persistency of high pressure systems over the GrIS that promoted an increase in the absorbed solar radiation as well as the flow of warm, moist air along the western portion of the ice sheet towards the north of the ice sheet. The anticyclonic conditions were responsible also for reduced cloudiness in the south and consequent below-average summer snowfall and albedo in this area. Similarly to 2012, anticyclonic conditions dominated summertime (Fig. 10a). The anomaly also occurred at the surface (Fig. 10b), suggesting that the pressure anomaly in the mid-troposphere was driven by atmospheric circulation rather than by the warming of the free atmosphere below 500 hPa levels. The anticyclonic conditions also promoted the advection of warm air that reached the northern portion of the ice sheet explaining why the highest temperature anomaly at 700 hPa occurs in this area (Fig.11c). As a reference, Fig. 11 shows the absolute values of the temperature at 700 hPa (T700), 200 hPa (T200) and mean specific humidity over 700-500hPa from NCEP-NCARv1 reanalysis on the July 12th, 2012 and on the July 31th, 2019. While the temperature anomalies were higher in 2019 with respect to the climatology of Mid-July or of the end of July, the absolute values were higher in 2012 than in 2019. In addition, the humidity content was also higher in 2012 than in 2019 over the ice sheet, showing the important role of liquid clouds in the 2012 extreme melt event (Bennartz et al., 2013). These differences in
temperature and humidity pattern explain why the 2012 highest melt event was more extreme than the 2019 one. Over the center of the ice sheet, surface temperature was close to the 1981–2010 average, suggesting a larger role of the radiative forcing than the thermal one. The mean summer sea level pressure (SLP) averaged over the 60-80°N, 20-80°W region (i.e., the same area used to compute GBI, Hanna et al., 2016), reached a breaking record value of 1016 hPa vs. a 1981–2010 summer average of 1010+/−2 hPa. Also the summer averaged 500 hPa geopotential height integrated over the same area, set a new record of 5567 m, against a 1981-2010 average of 5497+/−25 m (Fig. 1a). We computed the persistency of anticyclonic conditions, defined here as the number of days when the daily mean SLP averaged over the Greenland ice sheet exceeds 1013 hPa (the common value of the standard pressure), and found that during the summer of 2019 such conditions existed for 63 of the 92 summer days (68% of the summer). In perspective, the average number of days with the same conditions during the period 1981–2010 was 28+/−12 days.

The anticyclonic conditions that characterized the summer of 2019 promoted negative cloudiness anomalies over the southern portion of the ice sheet and positive ones over the northern region (Fig. 5d), pointing to the important role of clouds in enhancing melting in this area (i.e., Hofer et al. 2017). In the North, the exceptional persistence of a high pressure system centered near Summit over the whole 2019 summer (Fig. 11b) favored advection of warm and wet air along the west side of Greenland towards the North, promoting higher than average surface temperatures (Fig. 5e) and positive anomalies of long wave downwelling radiation (Fig. 5f). In the southwest, dry and sunny conditions dominated. This promoted positive anomalies of the incoming shortwave radiation (Fig. 5g) which, in turn, when combined with the relatively low albedo (due to reduced summer snowfall) promoted positive anomalies of the absorbed shortwave radiation (Fig. 5h) higher than 30 W m−2. Such drier conditions also allowed temperatures to reduce during nighttime, explaining why the temperature anomaly was not playing a larger role over these regions. Integrated over the whole ice sheet, the anomalies of shortwave and long wave downwelling radiation were not significant but, as a result of a quasi permanence of exposure of low albedo zones, the anomaly of absorbed shortwave was the highest since 1948, with an anomaly integrated over the whole ice sheet of 7.9 W m−2, being four times the 1981-2010 standard deviation (inter-annual variability) of 1.9 W m−2. The strong relationship between runoff and atmospheric conditions is also apparent in Fig. 13b, where scatter plots of runoff with 500 hPa GPH summer mean anomalies (Fig. 13a) and with 700 hPa temperature (Fig. 13b) are shown, together with the coefficients of the linear regression between runoff and the two atmospheric quantities. Reinforcing the idea that radiative forcing played a large role with respect to thermal forcing, the summer of 2019 (marked in the two panels with a large, orange circle) is beyond two standard deviations from the mean in the case of the 700 hPa temperature where it falls closely to the regression line in the case of the 500 hPa GPH.

To further understand the role of the atmosphere on the 2019 SMB record and the linkages between atmospheric circulation and SMB, we classified summer (JJA) daily 500 hPa GPH anomalies between 1948 and 2019 into a set number of classes to study how the frequency of such classes has changed over the past decades and how the 2019 summer positioned itself within the 1948–2019 record. We focus on the 500 hPa GPH because of its strong correlation with the surface melt (Fettweis et al., 2011b) and because it is a standard height for gauging the effects of jet stream blocking on
We classify the daily 500 hPa GPH anomalies by means of Self Organizing Maps (SOMs), being artificial neural network algorithms that use unsupervised classification to perform nonlinear mapping of high-dimensional datasets (Kohonen 2001). Initially, a set number of nodes is created (set by the user) and the nodes are randomly filled with the GPH anomalies daily fields. During the training phase of the SOMs, each of the daily 500 hPa GPH fields is allocated to one of the classes depending on the Euclidean distance of the new element from the existing SOM nodes. Once trained, the SOM network is interrogated by providing the daily 500 hPa GPH anomaly fields (1981 – 2010 baseline) and obtaining the corresponding class to which that particular atmospheric field belongs. From here, it is possible to calculate the frequency of occurrence of the classes of the atmospheric circulation patterns to provide insight into possible temporal changes associated with the identified classes and their relationship with SEB and SMB quantities. The number of nodes, which also corresponds to the number of classes in which the atmospheric patterns are classified (Kohonen, 2001), is defined by the user: using fewer nodes allows the user to include a broader range of circulation patterns within the same class but it decreases the amount of variability captured by the SOMs, while increasing the number of nodes results in classes that are less frequent and more closely resemble each other. Based on previous work (e.g., Mioduszewski et al., 2016) and following Kohonen (2011), we selected a total number of 28 classes. Fig. 14 shows the 28 nodes identified through the SOM analysis ordered according to the mean GPH values computed over the same area where GBI is calculated. For each node, the position of each class in the original grid is reported (shown as class #) together with the mean 500 hPa GPH value. The maps in Fig. 14 are obtained by averaging the 500 hPa GPH values over those days when the specific node was occurring according to the SOM classification. For reader’s convenience, in Fig. 15a we show the anomaly of the summer frequency of occurrence of each class (y-axis) for the years 1949 through 2019 (x-axis) with respect to the 1981 – 2010 period. Further, in Fig. 15b we show the number of days occurring in 2019 (blue bars) and 2012 (red line) for the different classes (x-axis). We selected 2012 and 2019 because of the enhanced surface melting that characterized both summers. We note that the atmospheric patterns characterizing the two summers show differences and similarities. Both summers, indeed, had a high number of days when class # 20 (highlighted with a rectangle with dashed contour in Fig. 14) was occurring (up to ~ 10 days in 2019). This class is characterized by large positive 500 hPa GPH anomalies (above 80 m) over Greenland and the Canadian archipelago, negative anomalies over Scandinavia and large positive anomalies over Siberia. Differently from 2012, however, classes # 11, 12, 13 and 28 were persistently present in 2019 (highlighted in Fig. 14 with rectangle with a continuous line). Classes # 12 and #13 show relatively low 500 hPa GPH anomalies over Greenland but strong positive anomalies over the Arctic ocean (class # 13) and the Canadian archipelago, eastern Siberia and Scandinavia (Class # 12). Classes # 11 and 28 show large positive anomalies over Greenland reaching both the Canadian archipelago and northern Europe and relatively high positive 500 hPa GPH anomalies over Siberia and Alaska. Notably, the cumulative number of days for classes # 11, 12, 13, 20 and 28 above identified exceeded 55 days in 2019 (Fig. 15c) being 5.1 standard deviations above the 1981 – 2010 mean of 14.2 days and pointing out, again, to the exceptional nature of the atmospheric conditions over Greenland during the summer of 2019. Nevertheless, we observe from Fig. 14b that 2012 also had high persistency in atmospheric patterns though such patterns belong to different classes than those in...
2019. For example, the cumulative number of days for 2012’s top 5 classes is 47, being similar in magnitude to the 55 days for 2019’s top 5 classes. Moreover, similarly to 2019, the top five classes in 2012 were all characterized by high GPH anomalies and strong anticyclonic conditions (though different in terms of spatial distribution of the GPH anomalies). In this regard both 2012 and 2019 can be assumed to be exceptional from an atmospheric point of view.

5 Conclusions

Using a combination of remote sensing observations, regional climate model outputs and reanalysis datasets as well as self organizing maps (SOMs), we have shown that exceptional anticyclonic conditions occurred in the summer of 2019 that promoted new records or close-to-record values of SMB, runoff and snowfall. Runoff in 2019 was the second highest after 2012 and SMB was the lowest on the record according to MAR forced by NCEP-NCARv1. The exceptional nature of the mass balance components in 2019 was strongly driven by albedo reduction associated with reduced summer snowfall, enhanced absorption of solar radiation and the flow of warm, moist air along the western portion of the ice sheet. The analysis of the frequency of daily 500 hPa GPH obtained from SOMs shows that the persistency of the atmospheric patterns (i.e., frequency expressed as number of days) characterizing most of the 2019 summer was unprecedented, being 5 standard deviations above the 1981–2010 mean, confirming the exceptional nature of the 2019 season over Greenland. Despite being similar in terms of runoff and SMB, the 2012 and 2019 exceptional melting seasons differ in terms of atmospheric patterns that drove those exceptional conditions, highlighting the importance of studying the spatio-temporal evolution of the atmospheric quantities, rather than only looking at integrated indices such as NAO ad GBI. In the future, we plan to analyse how the frequency and occurrence of GPH anomalies has been changing at higher levels (e.g., 300 hPa, 100 hPa) to quantify potential missing links between the stratosphere and the troposphere that might be responsible for the exceptional conditions. We plan to look at these potential linkages during the fall and winter months, when the coupling between the stratosphere and the troposphere is stronger than in summer and will explore the potential influence of winter and spring conditions on the summer atmosphere. As mentioned in the Introduction, understanding the role of atmospheric circulation changes on the surface mass balance of the Greenland ice sheet is a crucial step for improving estimates of its current and future contributions to sea level changes. This assumes even more importance when considering that such exceptional conditions are not captured by the Climate Model Intercomparison Project datasets (CMIP5, Hanna et al., 2018b), and they can increase the projected surface mass loss by a factor 2 according to Delhasse et al. (2018).

6 References


7 Acknowledgements

M. Tedesco would like to acknowledge financial support by the National Science Foundation (PLR-1603331, PLR-1713072, OPP 19-01603), NASA (NNX17AH04G, 80NSSC17K0351) and the Heising-Simons foundation. Computational resources for running MAR have been provided by the Consortium des Équipements de Calcul Intensif (CÉCI), funded by the Fonds de la Recherche Scientifique de Belgique (F.R.S.FNRS) under grant no. 2.5020.11 and the Tier-1 supercomputer (Zenobe) of the Fédération Wallonie-Bruxelles infrastructure funded by the Wallonia region under the grant agreement no. 1117545.

Data availability

Competing interests.
The authors declare they have no conflict of interest.

Authors contribution.
MT and XF conceived the study. MT collected and analyzed the remote sensing data and performed the atmospheric classification using the SOMs. XF generated the MAR outputs. Both authors contributed to the analysis and to the final version of the manuscript.
Figure 1 a) Number of days when melting occurred during the 2019 summer (June, July, August, JJA) according to spaceborne passive microwave observations (e.g., Tedesco et al., 2007). b) Anomaly of the number of melting days with respect to the 1981–2010 baseline period obtained from spaceborne passive microwave data shown in a).
Figure 2 a) Daily time series of melt extent (expressed as a percentage of the total ice sheet area) during 2019 (black line). Red line indicates the average values for the baseline period 1981 – 2010. Vertical gray bars indicate the standard deviation of melt extent for the 1981 – 2010 baseline period. b) Summer maximum melt extent (as a percentage of the ice sheet surface, blue line, left axis) and melting index (e.g., number of melting days times the area undergoing melting, Km$^2$/day, orange line, right axis) obtained from spaceborne passive microwave observations for the period 1979 – 2019.
Figure 3  

a) Number of melting days in 2019 obtained from spaceborne passive microwave observations. The data is the same as Figure 1 but the range has been reduced between 1 and 4 days to highlight melting occurring in the interior at high elevations. 

b) and c) Daily time series of spaceborne microwave brightness temperatures for the two selected points indicated by the tail of the arrow.
Figure 4 Time series of daily air temperature (top red line) recorded at the EGP PROMICE station (75.6247°N, 35.9748°W, 2660 m a.s.l.) together with time series of spaceborne brightness temperatures at 19.35 GHz recorded over the pixel containing the location of the EGP station (blue line) together with air pressure [hPa] recorded at the same station (bottom red line). Dashed orange line represents the 273.15 K values where the blue, dashed line represents the threshold on Tb above which melting is considered to occur.
Figure 5 Spatial distribution of the anomaly of the a) number of melting days b) snowfall c) albedo d) cloudiness e) 2m temperature f) longwave downwelling g) shortwave downwelling and h) shortwave absorbed obtained from the MAR model (1981–2010 baseline) forced by the reanalysis NCEP-NCARv1.
Figure 6 Time series of 1949 – 2019 annual (Sept. 1 2018 – August 31st, 2019) SMB (dark blue), snowfall (red) and runoff (yellow) values simulated by MAR over the whole Greenland ice sheet.
Figure 7 MODIS 2019 broadband albedo values anomalies (2000 – 2010 baseline) for a) summer b) June c) July d) August.
Figure 8 Time series of MODIS 2019 broadband albedo values for summer (dark, dotted line), June (medium blue line with triangles) July (dark blue line with disks) and August (light blue line with squares).
Figure 9 Time series of values of MODIS broadband albedo for the pixel whose brightness temperature is shown in Fig. 3b.
Figure 10 a) JJA 2019 averaged geopotential height anomalies at 500hPa (Z500 in m, baseline period 1981-2010) from the NCEP-NCARv1 reanalysis. Anomalies below two times the interannual variability (i.e., not statistically significant) are hatched. b) Same as a) but for the sea level Pressure (hPa) and for c) JJA temperature at 700hPa (°C). In each panel, arrows represent the anomaly of JJA winds at a) 500 hPa, b) 10 m and c) 700hPa.
Figure 11 Absolute values of the temperature at (a,d) 700hPa (T700), (b,e) 500hPa (T500) and (c,f) mean specific humidity over 700-500hPa from NCEP-NCARv1 reanalysis on the 12th of July 2012 (a,b,c) and on the 31st of July 2019 (d,e,f).
Figure 1: Standardized (1981–2010) a) summer (JJA) averaged GBI values for the period 1948–2019.
Figure 1. Scatterplot between runoff (in mmWE) and a) 500 hPa GPH (m) and b) 700 hPa Temperature (in °C). Red disks show 2019 values in both plots. The coefficient of a linear regression analysis are reported within each plot, together with the coefficient of determination R.
5391-Class #1
5405-Class #9
5432-Class #17
5442-Class #16

5460-Class #24
5461-Class #2
5461-Class #15
5462-Class #10

5471-Class #23
5477-Class #27
5479-Class #19
5485-Class #18

5489-Class #21
5490-Class #8
5498-Class #6
5502-Class #4

850
Figure 1. Mean 500 hPa GPH anomalies for the 28 classes identified by the SOM using the NCEP/NCAR reanalysis data for the period 1948–2019 ordered from the lowest to the highest mean GBI values computed using those days when the classes were occurring.
Figure 1. a) Anomaly of the number of days (1981 – 2010 baseline) of the occurrence of each of the 28 classes identified through the SOM analysis for the years 1948 – 2019. B) Number of days when the 28 identified classes (x-axis) occur during the summers of 2012 (red line) and 2019 (blue bars). C) Anomaly of the number of days for classes #11, 12, 13, 20, and 28 for the period 1948 – 2019.