Response to Review 2:

We thank the reviewer for their helpful comments and their insight on this subject. The authors acknowledge the constructive criticism and comments from the reviewer and propose the following revisions. We appreciate the comments by the reviewer which have resulted in a significantly strengthened manuscript.

The original comments from the reviewer are in black and in blue are the author's responses, with blue italics to show the in-text changes. The authors want to point out that due to many useful suggestions, the major revisions have been implemented resulting in significant changes to the manuscript, as can be seen in the document attached below.

Overview

This paper investigates the relationship between changes in snow specific surface area (SSA) and its isotopic composition, focused on d-excess, at EastGRIP. The Authors focus on precipitation events, after which rapid SSA decays are observed, coupled to a decrease in d-excess. The Authors propose an exponential rate law for SSA decay, which is temperature independent between 0 and -25°C. The Authors then discuss the interplay between snow metamorphism and d-excess, and the possible impact of their findings on the interpretation of the ice core isotopic record.

General comments

The idea underlying this research is very nice: snow metamorphism results in sublimation-condensation cycles which should lead to isotopic fractionation. SSA decay is taken as a proxy for the intensity of metamorphism, and the expected correlation between SSA decay and isotopic fractionation is found, and is readily visible in d-excess. Such a study is clearly relevant to the interpretation of the ice core isotopic record and the data presented therefore deserves attention.

However, my opinion is that the experimental protocol is partly flawed, and this unfortunately casts doubt on the validity of the data obtained and on the conclusions derived. The first point is that SSA is measured on a 1 cm thick layer while isotopes are measured on a 2.5 cm thick layer. Furthermore, no detailed observations of surface snow are mentioned to ensure that the thicker 2.5 cm sample was the same snow layer as the top 1 cm snow. In many cases, the authors may then be measuring 2 little-related snow samples, which would in fact completely invalidate their study.

The reviewer addressed a major concern related to the sampling protocol for SSA and isotopes. The authors primarily want to clarify that the isotopic composition was directly measured from each SSA sample. Thus, each SSA sample has a corresponding isotopic composition. The offset we refer to comes from the measurement resolution of SSA due to the e-folding depth of 1310 nm radiation in high density snow. We apologise that this was not made clear enough in the original manuscript, and we hope that this clarification gives the reviewer increased confidence in the sampling protocol. We add the following text to the manuscript:
“Individual SSA samples were put in separate bags and subsequently sampled for water isotopic composition. Thus, each day the 10 SSA samples have a corresponding isotopic composition.”

Many processes can affect the very surface snow layer. These include fog deposition, the formation of surface hoar or sublimation crystals, and wind drifting. All this is hardly mentioned, so that I am not even sure that adequate observations were systematically made. These are absolutely necessary for any careful snow physics investigation. If a 0.5 cm-thick fog deposit or surface hoar formation takes place, then clearly the SSA value will mostly reflect this deposit while the isotopic measurement will mostly characterise the underlying snow layer. Relating both measurements will then be totally meaningless. It is clear to me that the authors should have sampled only the top layer for isotopic measurements. If not enough material was present in their ICE CUBE sample holder, then they should simply collect more surface sample nearby.

The reviewer here addresses important comments related to other relevant processes for surface snow. Daily observations were recorded for snowfall, snowdrift, and ground fog, although there was no consistent documentation of surface hoar/sublimation crystal surface features. There is no doubt that fog deposition and surface hoar etc are processes that are important for SSA studies as documented by Domine et al., 2009; Gallet et al., 2014; Fergyresy et al., 2018. However, the observed SSA value for surface hoar is ~54 m² kg⁻¹ (Domine et al., 2009), is similar to the values on the initial day of our events. We would therefore most often expect an increase in SSA in the instance of surface hoar and snow drift (Kuhn et al., 1977; Grenfell et al., 1994; Domine et al., 2009; Libois et al., 2014).

We add the following text to highlight the importance of addressing potential of such surface features. In the methods, to clarify the potential that SSA increase is the result of precipitation, snowdrift or surface hoar:

“We here use the term deposition events to describe rapid increases in SSA, expected to be from precipitation, drifted snow or hoar formation. Previous studies have indicated that surface hoar and sublimation crystal-like grain growth features at the surface have an SSA value around 54 m² kg⁻¹, based on the SSA of hoar frost (Domine et al., 2009).”

In the discussion:

“However, we consider potential increases in SSA in the absence of precipitation under the following conditions: 1) surface hoar formation on an aged snow surface (SSA < 50 m² kg⁻¹), 2) the effective sieving of small, fragmented grains into the pore space via wind, and 3) from sublimation and subsequent fragmentation of snow grains while suspended by the wind (Domine et al., 2009). Selecting only rapid decreases in SSA reduces the probability of capturing these processes in our analysis.”

Regarding the depth of the sample, we add that each sample had 2.5 cm of snow. However, we can only say for certain that the top 1 cm of each sample was measured given the e-folding depth (now edited in manuscript from light penetration depth) for 200 kg m⁻³ is 1 cm, which is lower than the mean density for EastGRIP surface snow. We state this as a limitation.
Wind drifting is another important process, which is not detailed. The threshold of 6 m/s for the mean daily wind speed is simply not adequate. Hourly values must be considered, and in fact ideally maximum, not average values, are most useful to evaluate wind speed effect on drifting. But the best data on this aspect is observations. Wind drifting can easily be detected by observations. I appreciate that such observations cannot be done 24 hours a day, but the consequences of wind drifting are easily observable by looking at changes in the snow scene.

We agree with hindsight that this threshold is insufficient to reduce the likelihood of surface perturbation, and to address this we now use the 10-minute data from PROMICE. It is important to note here that 209 out of the total 237 sampling days have daily maximum wind speed exceeding 5 m s$^{-1}$ and no events had wind-speed consistently below 5 m s$^{-1}$ (two had 5.1 m s$^{-1}$). In addition, snowdrift events were documented in the EastGRIP field diary and correspond to wind-speeds above 7 m s$^{-1}$. Several events have maximum wind-speed between 6- 7 m s$^{-1}$, and no snowdrift documented. Based on this analysis and observations from the literature, we define two wind categories, as briefly suggested by the reviewer in a later comment, we have added a secondary wind-speed category for comparison of SSA decay when wind-speed is <6 m s$^{-1}$ (low-wind events), and when maximum wind-speed is between 6- 7 m s$^{-1}$ (moderate-wind events). The following text is added to the document:

“A set of criteria are required to reduce the potential of analysing events with wind-perturbed surfaces, resulting in the removal of surface snow. In Antarctica, unconsolidated surface snow has been observed to drift at wind speeds as low as 5 m s$^{-1}$ measured at 2 m height (Birnbaum et al., 2010). However, a study from Greenland documented snowdrift starting at 6 m s$^{-1}$ (Christiansen, 2001), likely due to warmer temperatures allowing for the surface snow to become more bonded (Li and Pomeroy, 1997). At EastGRIP, calm conditions correspond to wind speeds from 0 –5.2 m s$^{-1}$ according to field diary observations. The mean daily maximum wind speed for the three sampling seasons was 6.8 m s$^{-1}$, while blowing snow was documented only when wind speeds exceeded 7 m s$^{-1}$. Based on this assessment, we define two wind-speed categories for comparison of the effects of wind-speed on SSA decrease. The first includes events with wind-speed consistently below 5.2 m s$^{-1}$, hereafter referred to as low-wind events, to ensure no surface perturbation. Secondly, we consider events where the maximum wind-speed is between 6 –7 m s$^{-1}$, hereafter referred to as the moderate-wind events. The inclusion moderate-wind events allow an assessment of the influence of wind-speed on SSA decrease.”

Out of the 21 initially defined events, only 2 are below the wind-speed threshold with maximum values of 5.1 m s$^{-1}$ in both events. We expect negligible snowdrift for these two events allowing us to confidently argue that the surface is unperturbed and isotopic change is the result of snow metamorphism. The likelihood of drifting snow during moderate-wind events is considered using the equation defined from Li and Pomeroy (1998), where the threshold wind-speed for snowdrift is defined as a function of temperature.

Following the same structure as in the original manuscript, we construct the SSA decay model with parameter values set for the two wind-regimes. We add the revised figure to this response. Intuitively, the SSA decay rate is higher for moderate-wind events (-0.53 m$^2$ kg$^{-1}$ day$^{-1}$) compared to low-wind events (-0.41 m$^2$ kg$^{-1}$ day$^{-1}$). As the reviewer will see later in this response, we add the results from the comparison of our data and SSA decay model to
existing models from Flanner and Zender (2006) and Taillandier et al. (2007).

Drifting can remove newly precipitated snow or accumulate it some places. This must be recorded when sampling. It is fairly easy to recognize snow layers from careful observations. All these mandatory observations do not appear to have been done.

I very strongly recommend that the authors detail whatever observations were done and clearly say what has not been done. In their analysis, they should only keep data for which they are certain that SSA and isotopic measurements were on the same layer. All data with surface hoar, fog or sublimation crystals should be eliminated. Drifting events resulting in non-homogeneous layers that were sampled must likewise be eliminated. If there are not sufficient observations to sort the data, then I fear the study may be invalid.

We refer back to our previous response regarding the documentation of snowfall, snowdrift, and ground fog. We remove the events with snowfall and wind drifted snow and Table A with event overview is kept in the Appendix. In addition, we add the following text to the Methods section “Defining SSA decay events”.

“We here use the term deposition events to describe rapid increases in SSA, expected to be from precipitation, drifted snow or hoar formation. Previous studies have indicated that surface hoar and sublimation crystal-like grain growth features at the surface have an SSA value around 54 m² kg⁻¹ day⁻¹, based on the SSA of hoar frost (Domine et al., 2009). If snowfall/snowdrift/ground fog was documented during the SSA decay, this event is removed from analysis due to perturbation of the surface layer.”

We wish to highlight that the one of the low-wind events was preceded by ground fog, not snowfall. We see value in including these events given that we have ensured negligible wind-perturbation during the event. It is interesting to compare the isotopic change during these two events. We now explicitly include this in the results section “3.2 SSA decay events”:

“Both E10 and E11 had consistent clear sky conditions. We note here that E11 was preceded by significant ground fog, not snowfall, indicating that the peak value of 46 m² kg⁻¹ was likely the result of surface hoar, and thus, rapid SSA decay follows an SSA peak not caused by precipitation.”

To further accommodate this comment, we present the latent heat flux and temperature gradient data from the two low-wind events, and extend the discussion of isotopic change with regard to the near surface fluxes. "4.3 Rapid SSA decay and isotopic composition”. Here we state that a lack of consistent observation of surface hoar in the SSA samples as a limitation to the study, but we take every precaution to ensure we are analysing unperturbed surface snow.

The organization of the paper must also be modified. Data appear in the discussion. All results should be reported in the results section and extra figures showing wind speed and snow surface conditions must be drafted.

The structure of the paper has been modified to address this comment. Most restructuring is applied to the results, and the discussion then follows suit. Meteorological conditions are
presented in the first section of the results, highlighting the inter-annual variability in temperature, accumulation, and latent heat flux between the sampling years. A description of SSA and isotopic composition is then presented alongside the EOF analysis, before focussing on the SSA decay events. Having outlined the modified event criteria in the methods, the suitable events are defined, and the decay model is presented with parameter values best fit to the two wind-regimes. Comparison to physical based models from the literature is included in the model evaluation. The final results section on isotopic change primarily considers events from both the low- and moderate-wind regimes, before focussing in detail on the two low-wind events. The latent heat flux and temperature gradients are assessed to infer processes driving isotopic change. The revised results facilitate a more concise discussion.

Regarding the SSA decay rate law, I am not sure this is the best formula. Since sublimation is thermally activated, the absence of a temperature effect is strange. Perhaps when data is sorted, such an effect will appear. The authors quote (Cabanes et al., 2003) to support their choice of analytical expression, but those Authors had a temperature-dependent rate law. Furthermore, subsequent studies on SSA decay rate laws proposed other analytical expressions, and their exploration should be discussed when the rate law is investigated, not line 309 in the discussion.

We appreciate that the reviewer has pointed this out. The temperature-dependence is now stated from Cabanes et al. (2003) in the introduction, along with the subsequent models proposed by Legagneux et al. (2004), Flanner and Zender (2006) and Taillandier et al. (2007). We have added the following text to the introduction:

"Previous studies have proposed SSA decay models using a combination of field measurements and controlled laboratory experiments (Cabanes et al., 2002, 2003; Legagneux et al., 2003, 2004; Flanner and Zender, 2006; Taillandier et al., 2007). Exponential models to describe SSA decay are documented to be the best fitting to in-situ data from Arctic Canada (Cabanes et al., 2003). However, the lack of physical basis led Legagneux et al. (2003) to construct a new equation based on laboratory experiments to describe a temperature dependent SSA decay."

The rarity of consistent low-wind conditions limits all in-situ studies regarding the duration of SSA decay events. However, we feel the documentation of SSA decay at the surface is valid and useful for planning of future campaigns, where more detailed observations would be beneficial, and for remote sensing studies.

In summary, this potentially interesting study may be partially of totally invalidated by an inadequate experimental protocol, at least based on the information supplied in the paper. If the authors have made observations not reported in this version, they should report all relevant information in a revised version. I then recommend sorting the data and removing all data where there is a reasonable suspicion that SSA and isotopic measurements were not on the same snow layer. I also strongly recommend a more logical organization of the paper. The discussion is often unfounded speculation and must be considerably shortened. I propose below numerous specific comments that I hope will be useful to the Authors in preparing an extensively revised version, for which I recommend a second round of review. These
comments were written before the general evaluation, so there is some repetition. And finally, I kindly request that all Authors involved in this work make a careful reading of the revised version. This does not seem to have been done for the version I read, which is not very respectful for the reviewers.

We are grateful for the time and effort taken by the reviewer to comment on this manuscript. The edited manuscript follows a more logical format and the edits made based on the reviewer's comments have improved the quality of the study. We apologise for mistakes in the original manuscript, we will ensure the revised document is carefully checked for errors.

**Specific revisions required:**

Line 35. Spell out SSA=specific surface area, which is the surface area of the ice-air interface per unit mass of snow, expressed in m2 kg-1. It is not assumed to be linked to the optical grain size dopt, as mentioned by the Authors, it is rigorously and simply linked by a geometric relationship SSA=6/ρice*dopt, as shown in equation (1) of (Gallet et al., 2009), which is probably a more relevant reference than Linow 2012. In fact this relationship was already implicitly mentioned by (Grenfell and Warren, 1999), although they did not use the term specific surface area.

We apologise for missing this, we have changed this to:

"The snow-air interface can be described by the widely used parameter snow specific surface area (SSA), where the SSA of a snow sample is dependent on optical grain radius and density of ice (SSA = 6 / ρice *dopt) (Gallet et al., 2009), and can be utilised as a measure for snow metamorphism (Cabanes et al., 2002, 2003; Legagneux et al., 2002)."

Lines 41-43. The reasons for SSA decrease (of dry snow) are not explained well and even erroneously. Wind fragmentation in fact increases SSA since smaller crystals are formed (Domine et al., 2009). Sublimation does not necessarily lead to SSA decrease as it reduces crystal size; and likewise vapor diffusion does not necessarily lead to SSA decrease. What actually leads to SSA decrease is the disappearance of small structures, often by sublimation, and the growth of larger crystals, often but not only by vapor diffusion in the pore space.

We appreciate the reviewer's insight here and have made changes to the text to correct this mistake.

Introduction: "Freshly deposited snow has a high SSA which decreases with time under both isothermal (<10 °C m−1) and temperature gradient (>10 °C m−1) conditions (Cabanes et al., 2002; Legagneux et al., 2004; Domine et al., 2007; Genthon et al., 2017). Decrease in SSA is predominantly the result of Ostwald Ripening, where large grains grow at the cost of smaller grains (Lifshitz and Slyozov,1961; Legegneux et al., 2004), vapour diffusion in the pore space driven by sublimation and deposition (Flin and Brzoska, 2008; Sokratov and Golubev, 2009; Pinzer et al., 2012), and wind effects (Picard et al., 2019). Under natural conditions SSA decrease is driven by a combination of these processes depending on surface conditions (Cabanes et al., 2003; Pinzer and Schneebeli, 2009a), each potentially modifying the isotopic composition of the snow (Ebner et al., 2017)."
An additional sentence or two are proposed for the discussion to explain the influence of wind, specifically relating to the results from EOF analysis to mention the potential for SSA increase due to sieving of fragmented grains (Domine et al., 2009), and wind-pumping potentially reducing SSA via sublimation (Town et al., 2008). This is particularly of interest when we observe the covariance between the SSA and isotopic parameters, given that some increases in SSA could be due to this effect, and the corresponding isotopic change would be the result of fractionation and not from precipitation or wind-blown snow. To account for the ambiguity, we focus on decreases in SSA where grain growth is likely happening and refer to latent heat fluxes and temperature gradients when assessing isotopic change.

Line 47. It is erroneous to state that “While current versions of the so-called decay models exist, these are mostly based on lab-experiments and non-polar snow observations”. The works of Cabanes and Taillandier are mostly based on Arctic and subarctic observations. Granted, none of these studies used data obtained on ice sheets, and this could be mentioned, if there are reasons to believe that ice sheet processes involved in SSA decrease are in general different from those on seasonal Arctic snowpacks. By the way, (Carmagnola et al., 2014) tested various SSA decay models against data from Summit, Greenland, and this may me relevant to the authors’ topic.

We thank the reviewer for pointing this out and have corrected this error. The paper Carmagnola et al. (2014) is a useful reference for the comparison to models. Like Linow et al. (2012), they look at the snow properties over a vertical profile as opposed to looking at the temporal evolution of the exposed surface snow. We therefore maintain that our continuous SSA data from EastGRIP is a valid approach to quantify the in-situ SSA decay under natural conditions. Even in the case of elevated wind-speed, we believe it is useful to document how the surface SSA is influenced with regard to remote sensing, as the reviewer also pointed out. The analysis for remote sensing was outside the scope for this paper unfortunately.

The following edit is proposed to acknowledge the previous SSA studies for polar snow, and highlight that we are referring here to SSA studies in the accumulation area of ice sheets:

“While continuous surface SSA measurements exist from Antarctica (Gallet et al., 2011; Gallet et al., 2014; Picard et al., 2014), those from Greenland focus on the depth evolution of SSA (Linow et al., 2012; Carmagnola et al., 2013). A continuous dataset of daily SSA and corresponding isotopic composition measurements from the accumulation zone of the Greenland Ice Sheet can contribute to understanding the relevance of snow metamorphism for surface energy budget and for ice core studies.”

Line 78. What is meant by surface temperature? Is this the skin temperature measured by IR emission? Or is it the air temperature near the surface? Mentioning a reference is not sufficient. A paper must be self-standing and must not require looking up references for understanding, especially for such a central variable. If this is skin temperature, all relevant details must be given here, including the instrument used, the wavelength range and the emissivity value used. Furthermore, validation of the skin temperature measurements would be desirable. IR sensors require very careful calibration to be accurate.
We apologise for the oversight here and have added instrument specifics to Table 1. The surface temperature is calculated from upwards and downwards longwave radiation with long wave emissivity set to 0.97 and is added to the text.

"Surface temperature from PROMICE is calculated from upwards and downwards long-wave radiation (measured using Kipp & Zonen CNR4 radiometer) with long-wave emissivity set to 0.97."

Line 85 ff. Sampling procedure. It is essential to note when there is a change in the snow layer sampled, i.e. when there was wind drift or precipitation. I guess precipitation events were readily identified, but what about wind drift? Did the authors note when the layer being sampled changed because of wind erosion of wind accumulation? This is critical for data interpretation.

Wind drift was documented in the field diary as well as snowfall and ground fog. However, detailed observations of surface features were not measured consistently over the 3 sampling years. High spatial variability in SSA and accumulation gives us an indication of a heterogeneous surface. Moreover, we consider each sample site individually to avoid attenuation of signals by using the mean. The field observation protocol is added in the methods, and a description of the surface conditions has been added in the results.

Line 100. "Light penetration depth in snow of 200 kg m\(^{-3}\) is approximately 1 cm". Light penetration does not just depend on density, but also on SSA. Thus for 200 kg m\(^{-3}\), a penetration depth of 1 cm corresponds to a precise SSA value. Furthermore, penetration depth is not very meaningful. Do the authors mean e-folding depth? Note that if the e-folding depth is 1 cm, still 27% of the reflected light intensity will be due to depths >1 cm. Also did the authors make detailed observations of detailed surface processes such as surface hoar, sublimation crystals or rime events (these are frequent at Summit, perhaps also at EastGrip)? This is important because these thin surface deposits will greatly impact measured SSA, while they will be diluted in isotopic measurements. To evaluate penetration depth and the impact of surface deposits on SSA measurements, the Authors can use the TARTES model. https://snow.univ-grenoble- alpes.fr/snowtartes/ . This will allow them to make valid quantitative statements, and to explore the impact of surface deposits on measured SSA.

We appreciate the reviewer's insight here and clarify that we are referring to e-folding depth which has been corrected in the manuscript. As previously mentioned, significant fog, snow drift and snowfall were documented in the field diary. However, no consistent detailed observations of surface features such as surface hoar/rime/sublimation crystals were made. We propose to use the eddy-covariance LE measurements to identify the potential of these deposits during the low-wind events used to observe concurrent isotopic change. The reviewer mentions the TARTES model which is a valuable tool. Nonetheless we are constrained by the nature of our Greenland surface observations and unfortunately this limits us from getting accurate additional information. For future work, TARTES is surely very useful. The following text has been updated in the methods:

"The e-folding depth of 1310 nm radiation in snow of 200 kg m\(^{-3}\) is approximately 1 cm (Gallet et al., 2009). At EastGRIP, the mean snow density from 2017, 2018 and 2019 is 293 kg m\(^{-3}\)
resulting in each measurement being heavily weighted to the top <1 cm of the 2.5 cm sample (307 ± 40 kg m\(^{-3}\), 278 ± 47 kg m\(^{-3}\), 294 ± 50 kg m\(^{-3}\) for 2017, 2018 and 2019 respectively).”

Line 117-118. It is strange the Authors did not sample the top 1 cm for isotopic measurements, to ensure better correspondence with the SSA measurements.

In our responses above we have clarified the sampling procedure in greater detail than provided in the original manuscript. To ensure that the isotopes correspond to the SSA samples directly, this procedure was preferentially used, instead of taking two separate snow samples for SSA and isotopes. We hope these answers fulfil the reviewers request on these matters.

Table 1. Usually Table captions are concise and explanation are in footnotes. Most of the caption is in fact unnecessary and can be deleted.

All the Table captions have been edited to be more concise.

Lines 132-133. Eq. (1) was indeed proposed by Cabanes et al. as the most empirically accurate, but this was just to fit their limited data set. Legagneux (2005) proposed a theoretically correct equation (his Req. 2). That equation was also used by (Flanner and Zender, 2006). Taillandier et al. (2007) used an approximation of that equation to fit experimental data and their equation has a log form. I believe the expression of Taillandier is more suitable. From the discussion, the Authors tested it, but this should be detailed here, not in the discussion.

We acknowledge the usefulness in presenting the results of our inter-model comparison and have added the results in Section 3.3 Model Evaluation. Prior to this, the following text has been added to the methods section ‘2.2.1 Modelling SSA decay” to accommodate this suggestion:

"The first empirical SSA decay model was proposed by Cabanes et al. (2003) who described a temperature-dependent exponential decay based on snow samples collected from the Alps (Cabanes et al., 2002) and Arctic Canada (Cabanes et al., 2003). A following logarithmic equation (Eq. log) fit controlled to laboratory experiments was proposed by Legegneux et al. (2004), where parameters A and B are arbitrarily related to the decay rate and initial SSA of each sample and are linearly correlated at -15°C. To improve the physical basis of the model, the theory of Ostwald Ripening, describing grain growth driven by a physical need to reduce surface energy, was implemented into the model (Legagneux et al., 2005). The equation (Eq. 4) has two parameters \( \tau \) and \( n \); \( T \) is the decay rate and \( n \) relates to the grain growth. The physical model was developed by Flanner and Zender (2006) to incorporate more specific physical quantification to the parameters to include information about temperature, temperature gradient and density. Based on these three conditions, they created a look-up table for \( \tau \) and \( n \).

Taillandier et al. (2007) proposed two equations based on the logarithmic model first proposed by Legagneux et al. (2004) to define the decay rate under isothermal and temperature gradient conditions where they were able to directly incorporate a surface temperature
An empirical decay model is constructed upon previous studies (Cabanes et al., 2002, 2003; Flanner and Zender, 2006; Legagneux et al., 2002, 2003; Taillandier et al., 2007). This model uses continuous daily SSA measurements from EastGRIP to describe the behaviour of surface snow SSA in polar summer conditions. All samples of defined SSA decay events are used to quantify surface snow metamorphism.”

Lines 162-164. Ground temperatures are not very relevant to the explanation of crystal shapes, as these form in clouds at a different temperature. And by the way Domine et al. (2008) is not the most suitable reference for this. I recommend (Kuroda and Lacmann, 1982) and references therein.

This is a valid point which we overlooked; however, this explanation has ultimately been removed from the revised manuscript given that the SSA decay during the single ‘cold’ event is likely to have been influenced by snowdrift. The reference to Kuroda and Lacmann (1982) is appreciated for general understanding, and we apologise for the inaccurate referencing here.

Line 165. The upper threshold for wind speed used here is a daily mean value of 6 m s$^{-1}$. When the daily mean value is 6 m s$^{-1}$, it is very likely that gust speeds were much higher and that wind drifting took place, with major modifications in SSA. Perhaps transport even brought other layers. I think combining events with and without snow drift is not adequate to derive SSA decay rate laws. At the minimum, events with and without drifting should be treated separately to investigate wind effects. Regarding isotopes, the sampling of blowing snow would have been interesting. Was that performed?

To address this comment, we refer back to our response to an early comment in the ‘General comments’ section. As we previously noted, the SSA decay rate for moderate wind events (max. wind-speed 6- 7 m s$^{-1}$) is substantially higher than for low wind events (< 6 m s$^{-1}$). Here, we make use of the physical based model from Flanner and Zender (2006) and Taillandier et al. (2007) by comparing their predictions to those of our data and empirical model. These comparisons are presented in the results section “Model evaluation”, and then discussed in the section “SSA decay at EastGRIP”. Unfortunately, there was no sampling of blowing snow, but we mention this, as well as sampling of surface hoar, as a suggestion for future studies.

We add an additional figure (Figure A2) to show the results of the model comparison for the two low-wind events (E10 and E11), and for examples of moderate-wind events (E2 from 2017 and E18 from 2019). The two moderate-wind events have maximum 3 m wind-speeds of 6.26 m s$^{-1}$ and 6.28 m s$^{-1}$. Based on the drift threshold defined in Li and Pomeroy (1998), E2 has potential influence from snowdrift, but not E18 (U(10) = 7.09 m s$^{-1}$ and 8.17 m s$^{-1}$ for E2 and E18 respectively), which agrees with an underestimation of decrease from FZ06 compared to observation during E2. Interestingly, we get the lowest RMSE values for FZ06 and the moderate wind events. Possible explanations include the initial snow conditions and event duration, which are included in the discussion.

Line 178. What is the RMSE? This is mentioned line 194 but would be better mentioned here
Yes of course, this has now been added earlier in the text for all models used. Based on the revised analysis, the RMSE based on low-wind events is 3.64 m² kg⁻¹ for the exponential model from this study, 3.45 m² kg⁻¹ for FZ06 and 6.34 m² kg⁻¹ for T07 based on the individual sample sites. For the moderate-wind events the RMSE is actually smaller, the values are 2.48 m² kg⁻¹, 1.28 m² kg⁻¹ and 5.63 m² kg⁻¹ for this study, FZ06 and T07 respectively.

Line 190. The authors indicate intermittent snowfall during day 2 of E14. Why did they not remove this presumably thin layer to avoid this artefact? The thin layer greatly affected the SSA measurement but probably had little impact on the 2.5 cm-thick isotope sample.

Events with intermittent snowfall/snowdrift/ground fog are now removed from further analysis. Removing surface artefacts would likely result in a degree of compaction in the sampling holder, and therefore to avoid any disturbance to the samples, they were handled as little as possible.

Line 197. Why is not an equation proposed and tested for the lower temperatures?

During our sampling period, there was only one event with mean temperatures below -30°C. As previously mentioned, the wind-speed during this event is higher than the threshold. During the initial analysis, we grouped the events by temperature ranges, however, we did not observe a clear temperature dependence of the decay rate. After the removal of events likely to have surface perturbations, we observe a single event in the moderate-wind category which is poorly predicted by the equation for the wind-speed category. As the additional text below explains, this event had the lowest mean air temperatures and thus we do observe the expected temperature dependence.

"Event 9 in 2018 is poorly represented by the moderate-wind SSA decay model from this study. The mean air temperature for this event was -20.8°C, 5°C less than the next coldest (E11 at -15.3°C). Fitting the model for E9 alone gives a decay rate of 0.44 m² kg⁻¹ day⁻¹, similar to that of the low-wind events. We therefore observe a temperature dependence of SSA decay like Cabanes et al. (2003). Based on the limited number of events used here, we document low-winds having a similar effect to air temperatures below -20°C on the SSA decay rate."

Line 208. Are the units correct here?

Apologies, these have been changed.

Lines 204-205. No influence of basic environmental variables. How about cloudiness? A very important variable for SSA decay is the temperature gradient in the snowpack. Near the surface, this is going to be greatly affected by cloudiness. In the absence of clouds, there will be a much stronger temperature gradient near the surface than under cloudy conditions. This probably deserves a bit of exploration. Various proxies for cloudiness can be tested, in particular the longwave budget.
We had explored this in the original manuscript and found that there is no significant relationship between the SSA decay rate and cloudiness based on linear regression analysis. However, to clarify, we are not suggesting that these variables do not affect the SSA and the decay rate, but that based on our data alone, we do not observe a significant relationship. We do observe an interesting relationship between the principal components of SSA, d-excess and $\delta^{18}$O, and cloudiness/longwave radiation over the entire sampling period. The purpose of this analysis was to identify any systematic influence of the decay rate for the defined events, and therefore, we decide to focus on the dominant influences on the events we are analysing.

We observe an interesting relationship between the principal components of SSA, d-excess and $\delta^{18}$O, and cloudiness/longwave radiation over the entire sampling period. The purpose of this analysis was to identify any systematic influence of the decay rate for the defined events, and therefore, we decide to focus on the dominant influences on the events we are analysing.

We evaluate cloudiness when assessing the isotopic change during low-wind events. However, both events correspond to near constant clear skies.

Lines 242-243. Shaded regions in Fig 4 are said to indicate largely homogeneous snow cover. But The caption to Figure 4 says "Grey shaded regions indicate periods of high spatial variability in isotopic composition." I am confused.

Apologies for the mistake. To fix this inconsistency and to improve the coherence of the manuscript, we move the EOF analysis prior to the SSA decay model results. The principal components of each variable (SSA, $\delta^{18}$O and d-excess) are assessed for statistical significance, and we find that there are opposing regimes between the years. In 2019, $\delta^{18}$O and d-excess covary in the spatial and temporal dimensions, contrasted with the strong significant relationship between the principal components of SSA and d-excess in 2019. It is apparent that the two years differed significantly in overall temperature conditions, which is clear is the mean $\delta^{18}$O values, which is potentially related to the opposing NAO phase in 2017/2018 and 2019. Even in the SSA decay events the behaviour is different. The specific SSA decay shape, which is clearly identifiable in 2017 and 2019 is less obvious in 2018. Furthermore, this is relevant for the discussion of processes driving isotopic change in the low-wind events.

Lines 241-249. This discusses the correlation between SSA and d-excess. The coherence is better when the snow layer is homogeneous. Could that just be due to wind effects? When the wind speed is low and there is no wind drifting, the snow remains unperturbed and a priori homogeneous. On the contrary, under greater wind speeds, drifting takes place, heterogeneity is generated and SSA and d-excess become decorrelated. Furthermore, since SSA measurements probe about the top 1 cm while isotopic measurements probe the top 2.5 cm, it is clear that when wind drifting takes place, both measurements may measure highly different layers, explaining the decorrelation. How about limiting data analysis to those events without wind speed?

This is a useful insight from the reviewer, and we acknowledge that this could be the case. The correlation in 2019 is continuous throughout the season, which suggests that increases in PC1 of SSA, closely linked to precipitation, and decreases, closely linked to post depositional processes, are similarly influencing d-excess. The following text is added to the discussion:

"PC1 of SSA is interpreted as depositional events causing increase in SSA in the positive mode (Domine et al., 2009), and snow metamorphism or wind erosion in the negative (Cabanès et al., 2002, 2003; Legagneux et al., 2003, 2004; Taillandier et al., 2007a; Flanner and Zender,
2006). However, we consider potential increases in SSA without precipitation in the instance of 1) surface hoar formation on an aged snow surface (SSA < 50 m² kg⁻¹), 2) the effective sieving of small, fragmented grains into the pore space via wind, and 3) from sublimation and subsequent fragmentation of snow grains while suspended by the wind (Domine et al., 2009).

For the revised manuscript, we look in detail at the low-wind events only to ensure the same surface layer persists. By reducing the number of events, we can assess temperature gradients and latent heat flux for individual events, allowing for a more concise discussion.

An issue with limiting EOF analysis to the low-wind events alone is that the deposition/precipitation input is then removed, which is a key component of the relationship between SSA and d-excess while at the surface. A later comment from the reviewer observes that large increases in SSA (possibly precipitation, or another form of deposition) corresponds almost always to an increase in d-excess. We argue that this observation supports the argument that there is an overall decrease in d-excess during snow metamorphism.

Lines 256-257. Here the authors mention fog and negative LHF, i.e. likely surface hoar formation. Thus the authors may have observed snow conditions. All these observations must be mentioned when results are first presented. Data analysis must consider which processes were involved for each event. By the way, the standard abbreviation for latent heat fluxes is LE, not LHF.

We hope the previous responses have clarified the observations that were made. Observations are in Table A1 in the appendix, as well as a new plot with these observations indicated on the timeseries. LHF has been changed to LE throughout the text. Isotopic analysis of low-wind events now includes the LE and temperature gradient measurements to infer the vapour fluxes in the surface snow.

Line 268. The authors invoked re-exposed old snow to explain some d-excess values. Careful observations during sampling can answer this question. If there was 1 cm of recent snow over old snow, the SSA measurement will have measured recent snow while isotopic measurements will have measured predominantly old snow. This will affect the quality of the SSA-d-excess correlation analysis. Again, inadequate samples must be removed from the analysis.

Unfortunately, there is no precise documentation of layering of the snow used for samples. Instead, we refer to the accumulation data to identify changes in snow surface height during the analysis of isotopic change for the low-wind events.

Line 287-288. Changes in snow physical properties observed are probably not due to precipitation and metamorphism *sensu stricto* (i.e. involving only water vapor transport within the snow layer). Processes involved also include wind drift, fog deposition, surface hoar deposition, and also possibly sublimation crystal formation. This last process is due to vapor transport within the snow, but since the growth of completely new crystals is involved, I suspect their isotopic composition would be very different from that of the snow layer they originate from. Sublimation crystals are in fact very frequent on cold snow under intense sunlight, even though reports are few (Weller, 1969; Gallet et al., 2014).
These are really useful points from the reviewer. Looking at surface crystal growth through the perspective of isotopes to determine sublimation crystals from deposition of hoar crystals would be interesting, and a great contribution to the quantification of sublimation driven isotopic fractionation. The sampling strategy used here favoured a broad study looking at the macroscale relationships between snow metamorphism and isotopic composition, and the large decrease threshold was used to extract changes in SSA over the transect after high initial SSA values had been recorded.

Given that surface hoar/sublimation crystals were not documented, we use LE measurements to determine whether there was significant surface hoar formation during analysed events. Determining sublimation crystals is more ambiguous here but we look at temperature gradients throughout the events to explore the possibility. We refer here to the recent paper by Casado et al. (2021) where the snow isotopic composition and modelled precipitation isotopes were used to infer the relative influence of precipitation and snow metamorphism on the isotopic signal. To accommodate this comment and significantly strengthen our study, the revised discussion presents the expected fractionation effects of processes driving snow metamorphism and infers the mechanisms of isotopic change based on previous studies (Hughes et al., 2021; Wahl et al., 2021; Casado et al., 2021). The following text is added:

"Three key mechanisms are expected to drive the rapid SSA decays; 1) large grains growing at the expense of small grains (Legagneux et al., 2004; Flanner and Zender, 2006), 2) diffusion of interstitial water vapour (Ebner et al., 2017; Touzeau et al., 2018; Colbeck, 1983), 3) sublimation due to the wind ventilating the saturated pore air, known as ‘wind-pumping’ (Neumann and Waddington, 2004; Town et al., 2008). The dominant mechanisms can theoretically be identified by a combination of the change in isotopic composition - indicating the fractionation effect - and the LE and temperature gradient data.

In theory, mechanism 1) causes minimal change in the bulk isotopic composition of a snow layer under isothermal conditions (Ebner et al., 2017). Therefore, observations of SSA decay corresponding to negligible isotopic composition change could be explained by this mechanism. We observe no events with consistent isotopic composition throughout. In the instance of 2) interstitial diffusion, light isotopes are preferentially diffused, while the heavy isotopes will be preferentially deposited onto the cold snow grains (Ebner et al., 2017; Touzeau et al., 2018; Colbeck, 1983). Thus, diffusion of water vapour in the pore space causes a decrease in d-excess and slight increases in δ18O due to kinetic fractionation (Casado et al., 2021). 3) Sublimation has been widely documented to cause an increase in δ18O of the remaining snow mass due to equilibrium fractionation, and a significant decrease in d-excess due to kinetic fractionation (Ritter et al., 2016; Madsen et al., 2019; Hughes et al., 2021; Wahl et al., 2021; Casado et al., 2021).

An overall increase in δ18O and decrease in d-excess during E10 can be attributed to a combination of 2) and 3) based on observation of net-sublimation and high amplitude diurnal temperature gradient variability indicating vapour transport within the pore space. The period between 9th June at 15:18 UTC and 10th June 10:40 UTC recorded net deposition corresponding to an overall decrease in δ18O during the first day and minimal decrease in d-excess, potentially due deposition of atmospheric water vapour (Stenni et al., 2016; Feher et al., 2021; Casado et al., 2021).
A 30% decrease in d-excess corresponds to negligible change in δ18O during E11. Net-sublimation, double that of E10 is measured, but with reduced amplitude in both TGs. Moreover, the largest decrease in d-excess occurs after the first day when the surface-subsurface TG is consistently negative. This indicates that vapour diffusion is controlling the isotopic composition, and the effect of equilibrium fractionation during sublimation from the surface only weakly influences the bulk isotopic composition (Casado et al., 2021).”

Line 291-292. For older snow also, sublimation and vapor diffusion are not the only processes involved. In particular, wind drifting is probably important.

This has been included in the text.

Line 297. The correct reference is Cabanes 2003, not 2002

We apologise for this mistake and have changed this in the text.

Line 309-310. The comparison with the equation of Taillandier should be indicated in results. In fact, the choice of Cabanes’ equation should be justified earlier on. Its interest as well. By the way, (Cabanes et al., 2003) used a temperature-dependent exponential coefficient.

In addition to a more extensive introduction to the models in the methods “modelling surface snow metamorphism” we have added a brief inter-model comparison to the results, using the temperature-gradient model from Taillandier et al. (2007), and the model from Flanner and Zender (2006), with tau and n determined by their look-up table based on the event conditions.

Lines 311-318. This paragraph is not physically very sound and is not based by any quantitative analysis. Since the temperature gradient near the snow surface is not evaluated, there is no basis to say that isothermal metamorphism is dominant after precipitation. Then, since the Authors do not find any significant effect of temperature, they assume their observations are explained by the temperature gradient, implicitly implying that the temperature gradient show little variations between events. This paragraph should just be removed. All the statements are unsubstantiated. Furthermore, what is important in TG metamorphism is not the magnitude of the temperature gradient, but the magnitude of the water vapor flux, which is temperature- dependent. Lastly, it can be affected by wind speed through wind pumping and also by convection (Trabant and Benson, 1972; Benson and Trabant, 1973; Johnson et al., 1987; Sturm and Johnson, 1991). All these aspects would need to be discussed and quantified to engage in the discussion proposed in this paragraph.

We apologise for not stating that there are snow temperature measurements from PROMICE from 2017 and 2018, and from a separate campaign from 2019. The inclusion of all events in the original manuscript did not facilitate in-depth analysis of individual events. However, thanks to the reviewers’ suggestions, the revised manuscript now includes the temperature gradient and latent heat flux for the low-wind events. Ultimately, this paragraph has been removed, but the influence of temperature gradients on the low-wind events has been discussed in the following sections, where aspects of both paragraphs have been merged.
Lines 319-322. Here again, the authors make unfounded statements. How do they know the temperature gradient is negligible during polar night? Under clear sky conditions, radiative cooling will on the contrary induce strong temperature gradients near the surface of the snow. The authors may just conclude that since their model is empirical it only applies under the conditions where data were obtained. In fact, it may not even be valid at this site in summer during other years.

The paragraph has been corrected and instead of suggesting that temperature gradients are minimal, we have discussed this in terms of absolute temperatures being lower, and thus the SSA decay would be slower (Flanner and Zender, 2006), as evidenced by E7 with temperatures < -30°C.

“Snow metamorphism is thermally activated given the dominant influence of sublimation and deposition (Cabanes et al., 2002, 2003; Legegneux et al., 2004). During winter, the temperatures are very low (<-30°C) and minimal insolation reduces the diurnal near-surface snow temperature gradients, resulting in isothermal metamorphism being dominant which reduces the rate of snow metamorphism, or SSA decay, compared to temperature gradient snow metamorphism (Dadic et al., 2008).”

Furthermore, we appreciate that the empirical model construction in this study is limited by synoptic weather variability being consistent wind and potential for surface perturbation. With consideration to this limitation, we believe it is still useful to document with the decay model, given the relationship between SSA and surface energy budget.

Lines 324-331. Could not the authors compare their model to data obtained using the algorithms developed in (Kokhanovsky et al., 2019)? It seems possible to determine precipitation events using Sentinel data, as indicated by high-SSA periods, and then investigate the decay to test whether the model developed here indeed applied to the accumulation zone of the GIS. This paragraph lacks convincing arguments and sound a bit like just wishful thinking, while tests are possible.

Yes, we agree that this would be a useful comparison. However, for this paper we decided to focus on the relationship with isotopes. This paragraph is removed, and the satellite potential is mentioned briefly in the previous section instead related to the usefulness of defining the SSA decay rate as a function of different wind-regimes.

Lines 336-337. Why would this correlation between SSA and d-excess be observed in only 72% of cases? I think it would be interesting to explore which events actually monitored a constant layer, rather than a layer perturbed by wind drift, the formation of surface hoar or sublimation crystals, or fog deposition.

This is a useful point from the reviewer, and we have added a section in the results that looks only at the isotopic change during the minimally perturbed low-wind events. We have included LE and temperature gradient data to identify the dominant direction of vapour flux during the events. Isotopic change is now documented in the context of sublimation and deposition between the surface and atmosphere, while the 10 cm snow temperature data gives an indication of the direction of vapour flux within the snow. The discussion is edited in parallel
with a more concise comparison to expectations from previous studies such as Casado et al. (2021). The same analysis has now been applied to events with minimal perturbation from ground fog, snowdrift, and snowfall.

Lines 339-351. This discussion of snow metamorphism could be significantly improved. I am not sure surface curvature effects played a detectable role. In any case, the authors need to substantiate this with quantitative calculations, they cannot just make such statements without a demonstration. I would think water vapor fluxes caused by temperature gradients and wind pumping, and perhaps thermal convection, can explain most observations.

We acknowledge the reviewer’s suggestions and have modified the discussion to explain the increased decay rate under moderate-wind conditions. The following text is added to the discussion:

"The expected temperature dependence on the SSA decay rate is apparent during E9, where the mean air temperature in less than -20°C, which agrees with the accepted knowledge that snow metamorphism is slower in colder conditions due to sublimation and deposition being thermally activated processes (Cabanes et al., 2003). In addition, we focus on the influence of wind-speed of the SSA decay rate and observe a more rapid SSA decay with increased wind-speed, potentially due to increased ventilation of saturated pore air acting as a catalyst for snow metamorphism (Cabanes et al., 2003; Flanner and Zender, 2006; Neumann and Waddington, 2004). Wind erosion cannot be definitively ruled out due to dis-continuous documentation of surface conditions. However, high wind-speeds are documented to increase SSA via fragmentation and sublimation of suspended snow crystals, which are then re-deposited and effectively sieved into the pore spaces of the surface snow layer (Domine et al., 2009)."

Lines 360-366. This paragraph discusses the relationship between SSA increases and concomitant d-excess increases. However, this seems very misleading to me. This paper is focused on SSA decrease of a given snow layer over time. Here, the approach is different. The authors consider changes in the SSA of surface snow, regardless of whether these changes involve the same layer. In fact, their SSA increases seems to always involve a change in layer, e.g. due to precipitation. Therefore, plotting data obtained by the evolution of a given identified layer together with data involving a change of layer seems meaningless to me. What I understand from this paragraph is that new layers with high SSA have a higher d-excess value than older (and different) layers with low SSA. This may be interesting, but is different from the main topic of this paper, and should therefore not presented as the same topic.

We acknowledge that there was a lack of clarity here and appreciate the reviewer’s comments to allow us to clarify and strengthen our findings. By measuring the isotopic composition of the SSA sample, we remove the uncertainty from spatial variability. Analysing isotopic change over 2-days ensures that new/re-deposited snow will have more time to equilibrate with the sub-surface snow. If repeated, precipitation and surface hoar isotopes would be measured to determine the influence of the surface depositions on the 2.5 cm isotope measurements. We add this as a limitation of our study.

Regarding the observations of snow with high SSA having a higher d-excess value than old
snow, we acknowledge the reviewer’s comment that this is not the same topic but propose the this feature as supporting evidence. Given that we observe no seasonal trend in d-excess, the consistently increased d-excess values with high SSA cannot be attributed to increasing d-excess throughout the season. Moreover, the documentation of d-excess decrease during low-wind events ensures negligible removal of snow. Therefore, with the support of LE and temperature data, we argue that this feature is the result of decrease in d-excess during snow metamorphism due to the combined influence of grain growth via vapour diffusion, and sublimation into the atmosphere (Ebner et al., 2017; Hughes et al., 2021; Wahl et al., 2021; Casado et al., 2021).

Lines 368-373. It is surprising to see data presented in the discussion. This should be in the results section. So in fact there seems to have been observations of snow surface conditions and changes. Wind drifting, a key process for data interpretation, may have been observed after all. We need to see those data. Fig. A1 needs to also show mean hourly wind speed, and ideally maximum hourly wind speed if available, as well as observations of drifting. In fact, all surface snow observations, including fog deposition, the formation of surface hoar or sublimation crystals, and any other relevant information, must be shown in a Figure.

This figure has been incorporated into the results section, where we present the daily and 2-day change in one figure and only for events with minimal surface perturbation. The additional figures and description of conditions have been added at the start of the results.

Lines 393-399. The speculation between insolation, temperature gradient and d-excess may be potentially interesting, but lacks a clear basis. Since the authors did not measure T gradients and did not adequately discuss their role on d-excess, I think this paragraph is not very useful. Please substantiate or remove.

This paragraph has ultimately been removed. We agree with the reviewer that this is an interesting discussion point, but based on our results alone, we feel we cannot adequately substantiate the arguments.

The section on ice core implications could perhaps be strengthened a bit by treating specific examples. For examples, how is the d-excess signal affected by more frequent precipitation that metamorphose without wind perturbation, in comparison to precipitation events that rapidly form a wind slab with time-stable SSA? How does that relate to climate scenarios (e.g. glacial vs. interglacial). This is just a suggestion. I am sure the Authors can present other interesting cases. This is where I expected more in-depth discussions.

We agree that there is a lot of potential discussion points relating to implications for ice core studies. Specific examples are addressed in the section “Isotopic change during SSA decay events”, where we compare our observations to the fractionation effects expected from the different processes driving snow metamorphism. In addition, we discuss the inter-play between precipitation intermittency and temperature conditions as an explanation for the different regimes between 2018 and 2019. We appreciate that there are numerous interesting discussion points which could be added, and we thank the reviewer for the suggestions made here.
Exploring the role of snow metamorphism on the isotopic composition of the surface snow at EastGRIP

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Abstract.

Stable water isotopes from polar ice cores are invaluable high-resolution climate proxy records. Recent studies have aimed to improve knowledge of how the climate signal is stored in the water isotope record by addressing the influence of post-depositional processes on the surface snow isotopic composition. In this study, the relationship between changes in surface snow microstructure after precipitation/deposition events, surface snow metamorphism and water isotopes during precipitation-free periods is explored using measurements of snow specific surface area (SSA). Continuous daily SSA measurements from the East Greenland Ice Core Project site (EastGRIP) situated in the accumulation zone of the Greenland Ice Sheet during the summer seasons of 2017, 2018 and 2019 are used to develop an empirical decay model to describe events of rapid decrease in SSA, driven predominantly by vapour diffusion in the pore space and atmospheric vapour exchange. Linking snow metamorphism, wind-speed. The relationship between surface snow SSA and snow isotopic composition is primarily explored using Empirical Orthogonal Function (EOF) analysis revealed a coherence between the dominant mode of variance of SSA and d-excess during periods of low spatial variability of surface snow over the sampling transect. A coherence between SSA and d-excess is apparent during 2019, characterised by above-average temperatures and increased sublimation rates, suggesting that processes driving change in SSA also influence d-excess. Moreover, we observed changes in isotopic composition consistent with fractionation effects associated with sublimation and vapour diffusion during periods of rapid decrease in SSA. Our findings highlight the need for future studies to decouple the processes driving surface snow metamorphism in order to quantify the fractionation effect of individual processes on the snow isotopic composition.
1 Introduction

The traditional interpretation of stable water isotopes in ice cores is based on the linear relationship between local temperature and first-order parameters $\delta^{18}O$ and $\delta D$ of surface snow on ice sheets (Dansgaard, 1964). The second order parameter $d$-excess ($d$-excess = $\delta D - 8 \cdot \delta^{18}O$) is a result of kinetic fractionation caused by different molecular diffusivities of oxygen and hydrogen and has traditionally been interpreted in ice core records as reflecting moisture source conditions (Merlivat and Jouzel, 1979). Many factors must be accounted for when reconstructing temperature in ice cores. Accurate reconstruction requires consideration of including precipitation intermittency (Casado et al., 2020; Laepple et al., 2018), past variations in ice-sheet elevation (Vinther et al., 2009), sea ice extent (Faber et al., 2017; Sime et al., 2013), and firn diffusion (Johnsen et al., 2000; Landais et al., 2006; Holme et al., 2018). In addition, recent studies have documented isotopic composition change in the surface snow during precipitation-free periods (Steen-Larsen et al., 2014; Ritter et al., 2016; Casado et al., 2018; Hughes et al., 2021), linked to synoptic variations in atmospheric water vapour composition and subsequent snow-vapour exchange (Steen-Larsen et al., 2014). Current research aims to quantify the influence of post-depositional processes on isotopic change of the surface snow (Steen-Larsen et al., 2014; Ritter et al., 2016; Madsen et al., 2019). Post-depositional processes at the surface involve additional kinetic effects adding complexity to the interpretation of $d$-excess (Hughes et al., 2021; Casado et al., 2021).

Surface snow undergoes structural changes as grains form bonds, grow. This process is called snow metamorphism, which is active at the surface and at greater depths, depending on temperature (gradient) conditions (Colbeck, 1983; Pinzer and Schneebele, 2009b). Major change the snow is undergoing, is the reduction of the ice-air interface to reduce energy (Legagneux and Domine, 2005). We here explicitly refer to snow that is lying at the surface for an unknown amount of time and thus does not directly represent freshly precipitated snow. Surface snow metamorphism initially drives a reduction in the snow-air interface to reach thermodynamic stability (Colbeck, 1980; Legagneux and Domine, 2005). The snow-air interface can be described by the widely used parameter SSA.

It is assumed to be linked to the optical grain size equivalent (Linow et al., 2012) and can be utilized as a measure for snow metamorphism (Cabanes et al., 2002, 2003; Legagneux et al., 2002). In this study, we use SSA to describe the (rapid) change of surface snow as one measure for snow metamorphism.

This manuscript focuses on surface snow property changes after precipitation. We here explicitly refer to snow which is lying at the surface for an unknown amount of time and thus does not directly represent freshly precipitated snow. Fresh snow crystals have a high value of SSA. After deposition of the crystals on the surface, the SSA rapidly decreases from its...
initial value due to crystal growth (Cabanes et al., 2002; Legagneux et al., 2004; Domine et al., 2007). The reasons for the SSA decrease are wind-driven fragmentation (Comola et al., 2017; Neumann et al., 2009), interstitial Freshly deposited snow has a high SSA which decreases with time under both isothermal (<10°C m−1) and temperature gradient (>10°C m−1) conditions (Cabanes et al., 2002; Legagneux et al., 2004; Domine et al., 2007; Genthon et al., 2017). Decrease in SSA is predominantly the result of Ostwald Ripening, where large grains grow at the cost of smaller grains (Lifshitz and Slyozov, 1961; Legagneux et al., 2004), vapour diffusion in the pore space between snow crystals (Pinzer et al., 2012; Flin and Brzoska, 2008) and sublimation (Sokratov and Golubev, 2009) driven by sublimation and deposition (Flin and Brzoska, 2008; Sokratov and Golubev, 2009; Pinzer et al., 2012), and wind effects (Picard et al., 2019). Under natural conditions, SSA decrease is driven by a combination of these processes depending on surface conditions (Cabanes et al., 2003; Pinzer and Schneebeli, 2009a), each potentially modifying the isotopic composition of the snow (Ebner et al., 2017).

Models can provide a quantitative description of the rapid SSA decrease after precipitation deposition. Previous studies have proposed SSA decay models using a combination of field measurements and controlled laboratory experiments (Cabanes et al., 2002, 2003; Legagneux et al., 2003, 2004; Flanner and Zender, 2006; Taillardier et al., 2007). While current versions of the so-called decay models exist, these are mostly based on lab experiments and non-polar snow observations. Exponential models to describe SSA decay are documented to be the best fitting to in-situ data (Cabanes et al., 2003). However, the lack of a physical basis led Legagneux et al. (2003) to construct a theoretical equation to describe SSA decay based on grain growth theory, which was then developed by Flanner and Zender (2006) who defined parameters based on surface temperature, temperature gradient and snow density.

Existing SSA decay models have not yet been extensively applied to polar ice sheet surface snow. Conditions for surface snow on polar ice sheets such as Greenland are however not necessarily comparable to other alpine regions. The dry-accumulation zone of the Greenland ice sheet has only small amounts of intermittent precipitation. Furthermore, the and Arctic regions regarding negligible melt and the high-latitude radiation budget is different than in other alpine regions. Only few continuous datasets of daily- Moreover, while continuous surface SSA measurements exist from the remote regions of the polar ice sheets (Libois et al., 2014; Picard et al., 2014). While SSA observations from Greenland exist (Carmagnola et al., 2013; Linow, diurnal datasets covering multiple months and years provide a better foundation for Antarctica (Gallet et al., 2011, 2014; Picard et al., 2014), those from Greenland focus on the depth evolution of SSA (Carmagnola et al., 2013; Linow et al., 2012). Continuous datasets of daily SSA and corresponding isotopic composition measurements from the accumulation zone of the Greenland Ice Sheet can contribute to understanding the relevance of snow metamorphism for ice core studies. In particular, studies of SSA and snow metamorphism from Greenland are relevant for isotope surface energy budget and for ice core studies. This is because snow metamorphism is expected to influence the snow isotopic composition as The latter is of particular interest owing to observations of isotopic fractionation during snow metamorphism documented in laboratory studies (Ebner et al., 2017) and field experiments (Hughes et al., 2021). Nonetheless, few studies have focused on the direct relationship between physical snow properties, such as SSA, and post-depositional changes in isotopic composition.

An SSA decay model optimized for Greenland conditions would provide a better quantitative foundation for a process-based understanding of surface snow metamorphism on Greenland. Furthermore, a quantitative description of Greenland SSA decay
would provide a basis to explore how snow metamorphism at the surface plays a role for the alteration of isotopic composition of Greenland snow after deposition.

In this manuscript, the aim is to explore the behaviour of surface snow metamorphism on polar ice sheets using daily SSA measurements and compare from Northeast Greenland during summer and compare the change in physical properties to the isotopic composition measurements. The primary focus is to document events where changes in SSA occur rapidly over a duration of a few over a number of days. We first identify events of rapid SSA decreases (decays) and explore how the isotopic composition of the snow changes during these events. Periods of rapid decrease in SSA are used as a proxy for snow metamorphism. Using daily field observations of snow properties from Northeast Greenland during summer, events of Events of rapid SSA decrease (SSA decay events) are used to 1) quantify and model surface snow metamorphism in polar snow and, 2) assess isotopic change during surface snow metamorphism. The data presented here has the potential to contribute to the understanding of the influence of post-depositional processes on physical and isotopic changes in the polar ice sheet surface snow. This allows for better understanding of snow properties at remote regions of polar ice sheets and contributes to the interpretation of water isotopes in polar ice cores.

2 Study site and methods

2.1 EastGRIP site overview and meteorological data

All data used in this paper were collected as part of the Surface Program corresponding to the international deep ice core drilling project at the East Greenland Ice Core Project site (EastGRIP 75.65°N, 35.99°W; 2,700m.a.s.l) during summer field seasons (May-August) of 2017, 2018 and 2019. The accumulation rate is approximately 14 cm w.eq yr⁻¹ (Schaller et al., 2017).

Meteorological data used for this study are from the Program for Monitoring of the Greenland Ice Sheet (PROMICE) Automatic Weather Station set up by the Geological Survey of Denmark and Greenland (GEUS) at EastGRIP in 2016 (Fausto et al., 2021). The data are 10-minute mean values for a multitude of variables. Snow temperature was measured using a thermistor string at 0.1 m intervals during 2017 and 2018 but was modified to 1 m intervals in 2019. An additional thermistor string was thus installed in May of 2019, from which we use the 0.1 m measurements. Instrument specifics can be found in Fausto et al. (2021). Mean weather conditions vary between sampling years, as outlined in Table 1. Instrument specifics can be found in Fausto et al. (2021). Mean summer surface temperatures for 2019 were 10.6±5°C, 5°C higher than 2017 and 2018. Westerly winds prevail, with mean wind speed of 4.5 ms⁻¹ (Madsen et al., 2019) during 2017 and 2018 with a wind direction of 227° N, while 2019 had a prevailing south-westerly wind (239° N), corresponding to opposing phases of the North Atlantic Oscillation (NAO).

An Eddy Covariance tower eddy-covariance (EC) measurement tower was set up at EastGRIP in 2016. The relevant variable measured from this system is 2016 to measure wind and humidity fluxes (Madsen et al., 2019; Wahl et al., 2021). Here we use the 30-minute latent heat flux (LHF) which is directly determined by LE measurements which are calculated from the measurement of humidity fluxes between the surface and atmosphere. Positive LHF-LE indicates upwards energy flux in the
Figure 1. SSA Sampling Procedure

a) A map of Greenland with a black star indicating the EastGRIP site (Source: Eric Gaba—Wikimedia Commons user: StingVisitGreenland).
b) A photograph of the clean snow area at the field site (Credit: Bruce Vaughn), with black lines indicating the SSA sampling transect with 10 m spacing shown as dashed lines. c) A photograph of SSA sampling cups (Credit: Sonja Wahl), and d) an illustration of the sampling device from Klein (2014).

All field seasons had net sublimation, with the highest magnitude observed in 2019 (See Data Availability Section A).

Significant weather conditions such as ground fog, drifting snow and snowfall, were documented each day in the EastGRIP field diary.

2.2 Snow sampling procedure

Each summer season of 2017, 2018 and 2019 snow samples were taken once a day from May to August at 10 sampling sites, each marked by a stick, along a 90m transect with 10m spacing upwind of the EastGRIP camp to ensure clean snow (Fig. 1b). The specific dates for each season are given in Table 1. The precise location of each sample was marked by a small stick to ensure the adjacent snow is sampled the next day and to avoid sampling snow from different depths. A 6cm diameter sampling device collected the top 2.5 cm of surface snow (Fig. 1c). Snow density is determined using the weight of each snow.
Table 1. Weather statistics - 2017, 2018 and 2019

<table>
<thead>
<tr>
<th>Mean Mean Mean</th>
<th>Instrument</th>
<th>2017</th>
<th>2018</th>
<th>2019</th>
</tr>
</thead>
<tbody>
<tr>
<td>Height Surface Temperature (°C)</td>
<td>(Kipp and Zonen CNR1/CNR4 radiometer)</td>
<td>-14.5 ± 6.2</td>
<td>-15.76 ± 7.6</td>
<td>-10.6 ± 5.4</td>
</tr>
<tr>
<td>Relative Humidity (with respect to ice) (%%)</td>
<td>(Calculated)</td>
<td>96.95 ± 15</td>
<td>96.95 ± 16</td>
<td>94.93 ± 14</td>
</tr>
<tr>
<td>Wind Speed (ms⁻¹)</td>
<td>(R.M. Young 05103-5 ±0.3ms⁻¹)</td>
<td>4.9 ± 2.0</td>
<td>4.2 ± 1.9</td>
<td>4.5 ± 1.6</td>
</tr>
<tr>
<td>Latent Heat Flux (W m⁻² W m⁻²)</td>
<td>(IRGASON Campbell Scientific)</td>
<td>1.28 ± 4.2</td>
<td>1.13 ± 4.3</td>
<td>2.6 ± 5.9</td>
</tr>
</tbody>
</table>

Mean and standard deviation for weather variables, surface temperature (calculated from upwards and downwards long-wave radiation with long-wave emissivity set to 0.97), relative humidity, wind speed and latent heat flux during the three sampling seasons. Surface temperature, relative humidity and wind speed were measured by the PROMICE weather station based on 10-minute measurements. Latent heat flux an upwards flux from the eddy-covariance tower.

Each snow sample is placed into the Ice Cube sampling container below an Infra-Red (IR) laser diode (1310nm), where the SSA is calculated based on IR hemispherical reflectance, explained in Gallet et al. (2009), while information on the Ice Cube device can be found in Zuanon (2013). Light penetration depth. The e-folding depth of 1310 nm radiation in snow of 200kgm⁻³ is approximately 1 cm (Gallet et al., 2011), resulting in a measurement of the top <1 cm of each sample (Mean snow density at EastGRIP (Gallet et al., 2009). The mean snow density from 2017, 2018 and 2019 is 293 kg m⁻³ (307 ± 40 kg m⁻³, 278 ± 47 kg m⁻³, 294 ± 50 kg m⁻³) for 2017, 2018 and 2019 respectively), resulting in each measurement being heavily weighted to the top <1 cm of the 2.5 cm sample. The light reflected from the snow samples is converted into inter-hemispheric IR reflectance using a calibration curve based on methane absorption methods (Gallet et al., 2009). A radiative-transfer model is used to
retrieve SSA from inter-hemispherical IR reflectance. To avoid influence from solar radiation, SSA was measured inside a ventilated white tent kept at temperatures between -5°C and -10°C. SSA measurements have an uncertainty of 10% for values between $5.1305 \pm 1.130 \text{ m}^2 \text{kg}^{-1}$ (Gallet et al., 2009).

### 2.4 Surface snow isotopes

Samples collected following the sampling procedure outlined in Section 2.2 were also used for isotopic composition measurements, resulting in Individual SSA samples were put in separate bags and subsequently sampled for water isotopic composition. Thus, every day the 10 daily isotope measurements taking SSA samples have a corresponding isotopic composition. The resultant isotope value is the average composition over the top 252.5 mm cm of snow. Each sample was sealed in polyethylene bags to avoid any air to equilibrate with the snow and affect the isotopic composition. All samples were kept frozen during transportation and storage.

After melting, each bag was shaken to ensure the isotopic composition of the sample is representative. 1.25 µl of each sample was then pipetted into a vial ready for isotopic analysis. The snow samples were then analysed at Alfred Wegener Institute in Bremerhaven using a cavity ring-down spectroscopy instrument model Picarro L-2120-i and L-2140-i following the protocol of Van Geldern and Barth (2012). This technique is used to obtain measurements of $\delta^{18}O$ and $\deltaD$ with an uncertainty of 0.15‰ and 0.8‰ respectively. $d$ excess is calculated by the equation $d_{\text{excess}} = \deltaD - 8 - \delta^{18}O$ with a resultant The calculated values for $d$-excess have an uncertainty of 1‰. Observing relationships between our SSA and isotope data requires consideration for the depth offset between the SSA measurements and the isotopic composition measurement which measures the entire 2.5 cm snow layer.

### 2.5 Data analysis

### 2.6 Data analyses

#### 2.5.1 Defining SSA decay events

This study focuses on the events where the SSA measurements decay rapidly over a duration of a few days. SSA decays are here defined as the events where the 2-day change of daily mean values are higher than a given threshold. This threshold is the same value for all years and is calculated based on the To systematically identify rapid decreases in SSA, which we use as a proxy for events of snow metamorphism after deposition (identified based on the high mean SSA values), a threshold is set using the bottom 10th percentile of the decays and set at-percentile of SSA decreases over a two-day period (-13 m² kg⁻¹ 2-day⁻¹). If the daily mean changes over a 2-day period is higher than the threshold, then this period is selected as a rapid SSA decay event. The duration of the event is set to start at the rapid decay and end on 2-day⁻¹. This was found to result in the most equal number of events from each sampling year compared to 1- and 3-day changes. SSA decay events are defined as by the initial peak, identified by the threshold, through to the next increase in SSA (rather than decrease).
We here use the term deposition events to describe rapid increases in SSA, expected to be from precipitation, drifted snow or hoar formation. Previous studies have indicated that surface hoar and sublimation crystal-like grain growth features at the surface have an SSA value around 54 m²•kg⁻¹, based on the day when the mean SSA measurements increase (rather than decrease) again. SSA of hoar frost (Domine et al., 2009). Accumulation data and field observations are used to identify the initial conditions.

A set of criteria are required to reduce the potential of analysing events with wind-perturbed surfaces, resulting in the removal of surface snow. In Antarctica, unconsolidated surface snow has been observed to drift at wind speeds as low as 5 ms⁻¹ measured at 2 m height (Birnbaum et al., 2010). However, a study from Greenland documented snowdrift starting at 6 ms⁻¹ (Christiansen, 2001), likely due to warmer temperatures allowing for the surface snow to become more bonded (Li and Pomeroy, 1997). At EastGRIP, calm conditions correspond to wind speeds from 0–5.2 ms⁻¹ according to field diary observations. The mean daily maximum wind speed for the three sampling seasons was 6.8 ms⁻¹, while blowing snow was documented only when wind speeds exceeded 7 ms⁻¹.

Based on this assessment, we define two wind-speed categories for comparison of the effects of wind-speed on SSA decrease. The first includes events with wind-speed consistently below 5.2 ms⁻¹, hereafter referred to as low-wind events, to ensure no surface perturbation. Secondly, we consider events where the maximum wind-speed is between 6–7 ms⁻¹, hereafter referred to as the moderate-wind events. The inclusion moderate-wind events allows an assessment of the influence of wind-speed on SSA decrease.

### 2.5.2 Modelling surface snow metamorphism

The first empirical SSA decay model was proposed by (Cabanes et al., 2003) who described a temperature-dependent exponential decay based on snow samples collected from the Alps (Cabanes et al., 2002) and Arctic Canada (Cabanes et al., 2003). A following logarithmic equation (Eq. (2)) fit controlled to laboratory experiments was proposed by (Legagneux et al., 2003), where parameters A and B were found to be arbitrarily related to the decay rate and initial SSA of each sample, and are linearly correlated at -15°C.

\[ SSA(t) = SSA_0 \cdot e^{-\alpha t} \]  
\[ SSA(t) = B - A \cdot \ln(t + \Delta t) \]  

To improve the physical basis of the model, the theory of Ostwald Ripening, describing grain growth driven by a physical need to reduce surface energy, was implemented into the model (Legagneux et al., 2004). The equation (Eq. (3)) has two parameters \( \tau \) and \( n \); \( \tau \) is the decay rate and \( n \) relates to theoretical grain growth. The physical model was developed by Flanner and Zender (2006) to incorporate more specific physical quantification to the parameters to include information about
temperature, temperature gradient, and density. Based on these three conditions, they created a look-up table for $\tau$ and $n$.

$$SSA(t) = SSA_0 \left( \frac{\tau}{t + \tau} \right)^{1/n}$$

Taillandier et al. (2007) proposed two equations based on the logarithmic model, defined by Legagneux et al. (2004), to define the decay rate under isothermal and temperature gradient conditions where they were able to directly incorporate a surface temperature parameter.

An empirical decay model is constructed building upon previous studies (Cabanes et al., 2002, 2003; Flanner and Zender, 2006; Legagneux et al., 2002, 2003; Taillandier et al., 2007). This model uses continuous daily SSA measurements from EastGRIP to describe the behaviour of surface snow SSA in polar summer conditions. The post-precipitation decreases in SSA are hereafter referred to as decays. All samples of defined SSA decay events are used to quantify surface snow metamorphism.

$$SSA(t) = SSA_0 e^{-\alpha t}$$

3 Results

Eq. (1) is proposed by Cabanes et al. (2003) as the most accurate description of SSA decay, where $SSA_0$ is the initial SSA value, $\alpha$ the decay rate. To best describe grain coarsening and the processes of sublimation and deposition driving mass redistribution of a new snow layer, days with mean wind speeds above 6

3.1 EastGRIP conditions

Meteorological variables over the three sampling seasons vary substantially. Figure 2 shows the 10-minute mean values of air temperature, wind-speed, relative humidity and latent heat flux (LE). The accumulation in Fig. 2d are daily mean values (see Section 2.2). Air temperatures were below 30°C between May 5th and May 8th, such low temperatures were not recorded for 2017 and 2019. However, when comparing the period from May 27th (start of 2019 season) to August 5th of each year, 2018 air temperatures (-13.3°C) were still 0.5°C lower than 2017 and 3.2°C lower than 2019. Two days during 2019 recorded air temperature above 0°C.

The 2017 season was characterised by high wind intrusions of $>13$ m s$^{-1}$ are removed to reduce the influence of wind redistribution. Individual sample analysis is preferentially used to avoid daily mean values possibly attenuating any signals due to spatial variability in surface snow age. Aged snow patches are expected to respond differently to surface processes than new snow patches due to different original crystal structures at the start of events, at approximately 20-day intervals. Considering all three sampling years, 2017, 2018 and 2019, the average daily maximum wind speed is 7 m s$^{-1}$, with 209 out of the total 237 sampling days having maximum wind speed above 5 m s$^{-1}$. The distributions of daily maximum wind-speed compared to 10-minute mean values are found in the Supplemental Fig. A1. Relative humidity is consistent throughout the years with mean values around 95% and similar variability of $\sim$ 7%.
Figure 2. Meteorological data from 2017, 2018 and 2019

Data is presented for the specific sampling periods for each year. The 10-minute mean data from PROMICE is shown for air temperature (a), wind-speed (b) and relative humidity (c). The bold lines indicate the mean values, based on the snow sampling time interval, for air temperature and relative humidity, and the maximum value for wind-speed. The Relative humidity is determined from vapour pressure and saturation vapour pressure. Latent heat flux (d) is 10-minute averages from the eddy-covariance tower, with the bold grey line showing the daily sum. Accumulation is presented in panel d).

4 Results

3.1 SSA decay events

There was a total of 5 cm accumulated snow over the 89-day season of 2017, half the amount of 2018 and 2019. The field season for 2018 started on the 5th of May, 9-days earlier than 2017 (14th May), and 22 days earlier than 2019 (27th May). Substantially more sublimation was recorded in 2019, where the daily sum was approximately double that of 2018.

3.0.1 Spatial and temporal surface variability
SSA data collected at EastGRIP indicate continuous changes in the physical structure of the snow crystals during all sampling seasons, with both temporal and spatial variability. The temporal SSA variability shows changes in physical snow structure with peak values closely associated with precipitation and decreases. A recent study at EastGRIP has shown the significant in-homogeneity in surface snow due to post-depositional reworking of the snow. Summer-seasonal SSA evolution is presented in Fig. ?? for 2017, 2018 and 2019 with each faded line representing individual samples (10 per day), and the bold line showing the daily mean. Spatial variability between sites is most prevalent when there are high SSA values, indicating fresh snow.

A total of 21 rapid SSA decay events are identified, with 6 (Zuhr et al., 2021). To avoid attenuation of isotopic signal, each sample is treated independently. Using a confidence interval of 95% (p<0.05), the relationship between SSA and isotopic

![Figure 3. SSA Timeseries 2017, 2018 and 2019](image)

Time-series of SSA time-series between May and August for (a) 2017, (b) 2018 and (c) 2019. Faded lines represent the 10 individual samples from principal components (PC1) of each variable (d). For each plot, the 90 m-markers indicate the individual sampling transect, while sites and the bold line-link shows the daily mean values. Gaps. The secondary y-axis in panel a) shows the timeseries represent missing data accumulation. Grey-The grey bars highlight the periods of decrease in SSA defined by the threshold algorithm for each year. Six decrease decay events are observed in 2017 and 2019, while nine are observed in 2018. Decrease events are interpreted as rapid grain growth due to snow metamorphism, and stars indicating days with precipitation.
composition is tested using Empirical Orthogonal Function (EOF) analysis. The purpose of EOF analysis is to identify the dominant modes of variance in both the temporal and spatial dimensions for each parameter - SSA, δ18O and d-excess - which are all measured from the same sample.

All parameters continuously change throughout the field seasons of 2017, 2018 and 2019 respectively. Grey bars in Fig. 3 highlight events defined by the decrease threshold. (Fig. 3), with large spatial variability in isotopic composition. SSA is characterised by peaks, often corresponding to large spatial variability, followed by gradual decreases over a number of days, a feature which is most prominent during 2017 and corresponds to negligible accumulation. The amplitude of SSA variability is largest in 2019. The start of the 2018 season has very high SSA values (daily mean -88 m² kg⁻¹) corresponding to low and homogeneous δ18O. Maximum SSA values for individual samples for 2017, 2018 and 2019 are 9285.3 m² kg⁻¹ and 82.953 m² kg⁻¹ respectively, while during 2017 there are only two instances of daily mean SSA being above 60⁻¹ and 86.7 m² kg⁻¹ respectively.

A visual inspection of the decay events - Inter-annual variability is observed in δ18O, with seasonal mean values of -31.6‰, -32.7‰ and -27.3‰ for 2017, 2018 and 2019 respectively (Fig. 3a). Note that the 2019 field season started approximately 15 days later than 2017 and 2018, resulting in a bias towards mid-summer conditions. Throughout the season δ18O follows a gradual increasing trend from May to August. Some cases of abrupt decreases (-10‰) are observed in the late summer, for example, on July 12th in 2018 and July 25th in 2019. No clear seasonal trend is observed in d-excess (Fig. 3b) but with periods of gradual decreases. Total daily spread in δ18O and d-excess is approximately 15‰.

The spatial and temporal principal components of each variable are presented in Fig. 3. Indicates a relationship between initial SSA and subsequent magnitude of decrease. To test whether the mechanisms of decay are consistent throughout events, observed SSA decays are analysed to construct an empirical model. During 2017, 2018 and 2019 all variables have one dominant mode of variance, or principle component (PC1). PC1 of SSA (PC1SSA) explains 61%, 77% and 72% of variance for the respective years, PC1 of δ18O (PC1δ18O) explains 69%, 83% and 75% of the total variance respectively, while PC1 of d-excess (PC1d-excess) explains 47%, 51% and 60%.

Distinct differences are observed between the sampling years, most prevalent is the opposing regime from 2018 to 2019. During 2018 PC1δ18O and PC1d-excess exhibit a significant relationship, with a strong negative correlation for the spatial component of PC1δ18O and PC1d-excess. A significant relationship is also observed for the temporal component of PC1SSA and PC1d-excess. In contrast, data from 2019 are characterised by significant relationships between PC1SSA and PC1d-excess in both the spatial (r=0.75) and temporal dimensions. No relationship is observed between PC1δ18O and PC1d-excess during 2019. For 2017, significant relationships (p<0.05, 95% confidence) are observed between the temporal component of PC1SSA and PC1d-excess and the temporal and spatial component of PC1δ18O and PC1d-excess. A shift is observed after July 15th where PC1d-excess changes from co-varying with PC1δ18O to PC1SSA.
3.1 SSA decay events

3.2 EastGRIP SSA decay model

Continuous SSA measurements allow for the construction of an empirical model to describe SSA decay at EastGRIP through
time while exposed to surface processes. All samples of defined SSA decrease events defined in Section 3.1 are used to quantify
surface snow metamorphism. For all events with mean temperature above –25°C visual inspection of the SSA decay events
highlighted in Fig. 3a indicates a relationship between initial SSA and subsequent magnitude of decrease. Prior to analysis,
we assess the meteorological conditions and field observations to remove SSA decay events with potentially perturbed surface
snow. This includes all events coinciding with observations of ground fog, snowdrift, and snowfall (indicated in Fig. 3), and
events where the wind-speed exceeds the thresholds defined in Section 2.5.1.

From the years 2017, 2018 and 2019 a total of 21 events are identified that fulfil the SSA decay criteria (as defined in
Section 2.5.1). These events are named E1, E2 etc (see Table A for more information on the individual events). Exploring
weather conditions for these events reveals that 12 out of the 21 events are influenced by either snowfall, snowdrift, or
ground fog according to field diary observations. Of the remaining 9 events, two are in the low-wind category (E10 and E11, the
mean SSA of the final day is around 30 = 5.1 m s\(^{-1}\)), and 7 in the moderate-wind category. Both E10 and E11 had consistent
clear sky conditions. We note here that E11 was preceded by significant ground fog, not snowfall, indicating that the peak
value of 46 m\(^{2}\) kg\(^{-1}\) (referred to as the background decay state). A relationship is observed between the was likely the result of
surface hoar, and thus, rapid SSA decay follows an SSA peak not caused by precipitation.

SSA samples are treated individually to quantify SSA decay rate for the different categories. The rate of SSA decay is closely
linked to the SSA value at the start of each event (initial SSA vs. magnitude of decrease during the decay period \(r^2 = 0.4\)) (Fig.
4), suggesting the rate of change is proportional to the absolute value, as described by exponential decay law (\(r= -0.71\) and
\(r= -0.84\) for low- and moderate-wind events respectively) (Cabanes et al., 2003).

The mean air temperature for all SSA decay events was between -20.8°C and -7°C. The first day of each event is characterised
by the largest change in SSA, followed by a decrease in magnitude over the subsequent days. This feature is most apparent for
the longer events (E1, E2 and E4), where SSA has minimal change below 25 m\(^{2}\) kg\(^{-1}\).

3.2 EastGRIP SSA decay model

SSA decay rate is quantified by plotting the rate of change in SSA per day against the absolute SSA value for all 10 sampling
sites for all low- and moderate-wind events (Fig. 4a). We observe a linear relationship between the rate of change in SSA
per day (ΔSSA) and SSA. Outliers are measurements from days with mean air temperature below –25°C as highlighted in
Fig. 4a. This observation is in agreement with theoretical understanding of snow crystal formation transitioning from dendrites
to columns at approximately –22°C (Domine et al., 2008). We therefore define the SSA decay model for a temperature range
between –25°C and 0°C and daily mean wind speeds below 6 m s\(^{-1}\) based on hourly averaged values. An overview of event
conditions using field observations are presented in Table A.
Figure 4. Decay Model Construction and Predictions
Linear regressions for change in SSA against the SSA for the low-wind (blue) and moderate-wind (purple) SSA decay events (a). Filled markers indicate the daily mean values and transparent markers show the individual samples sites. The observed SSA decays are show for the moderate-wind events (b), and the low wind events (c), followed by the modelled SSA decays for the respective events in d) and e). The legend in d) and e) indicates the SSA decay event number, presented in Table A.

Constructing the SSA decay model for EastGRIP is based on constructing using the differential equation for the linear relationship between ΔSSA and absolute SSA which is defined as Eq. (2) - Solving the differential with respect to time (t), produces the SSA decay model defined as Eq. (4), which follows the equation structure from of Eq. (1).

\[
\frac{dSSA}{dt}SSA(t) = -0.54 (SSA_0 - C) e^{-0.54t} + 14.69 C
\]

\[
SSA(t) = (SSA_0 - 26.8) e^{-0.54t} + 26.8
\]

Where SSA(t) is the SSA measurement at a given time, SSA_0 is the initial SSA value, and -0.54 \( \alpha \) is the decay rate, and C is the constant. The decay rate, determined by the slope of the linear regressions in Fig. 4, is higher for moderate-wind SSA decay events (-0.53 m^2 kg^{-1} day^{-1}) than for low-wind SSA decay events (-0.41 m^2 kg^{-1} day^{-1}). To account for a non-zero decay constant, the value 26.8 m^2 kg^{-1} C describes the 'background' SSA state which is defined by the value of x when the linear regression crosses the y-axis (y-axis in Fig. 4a). The SSA decay model describes rapid decrease in SSA based on empirical data from EastGRIP, Greenland.
3.2.1 Model evaluation

Decay Model Construction and Predictions: All samples for all events are included in plot a) showing the relationship between the rate of change in SSA per day (ΔSSA day⁻¹) against the daily absolute values. Points are coloured by the daily mean surface temperature. The linear regression is based on values for surface temperatures between –25°C and 0°C, and daily mean wind speeds below 6 m s⁻¹. b) shows a comparison between the model predicted SSA values using Eq. (4), against the SSA observations. The marker colour represents the day of the events (DOE). Marker style represents the sampling year to assess inter-annual variability for 2017 (○), 2018 (x) and 2019 (□). c), d) and e) show all included events in full line and f), g) and h) show the model predictions as the dashed line. E1-E21 refers to events as listed in Table A. Missing data day 1 E1.

Model performance is tested by comparing daily predicted decrease to the 10 daily observations. Model data residuals for daily data are normally distributed, suggesting no systematic errors in model predictions. Figure 4 shows the construction of the model (Fig.4,a,b) and prediction of SSA decay (Fig.4,f-h), equal to 21 m² kg⁻¹ and 24 m² kg⁻¹ for low- and moderate-wind events respectively. Note that events are here named E1, E2—consistent with Fig. 2 and also listed in etc. consistent with Table A.

There is a minor tendency for the model to underestimate the SSA decrease and thus overestimate the predicted values of SSA as seen in Fig. 4b. Model limitations are most evident during the first day, as seen in Fig. 4, where the modelled decay consistently underestimates the magnitude of decrease. The model has limited ability to predict observations below in the lower range of SSA observations as seen in Fig. 4f, g and h, where the modelled and observed values are compared for each event.

Following our definition in Section 3.1 the events have an extent of 2-5 days. To assess model performance in predicting magnitude of SSA decrease for events of different time periods, we compare the predicted versus measured SSA. E9 in 2018 is poorly represented by the moderate-wind SSA decay model from this study. The mean air temperature for this event was ~20.8°C, 5°C less than the next coldest (E11 at ~15.3°C). Fitting the model for E9 alone gives a decay rate of 0.44 m² kg⁻¹ day⁻¹, similar that of the low-wind events. For rapid events lasting 2 days the model tends to underestimate the rate of decrease. This is most apparent on Day 1 (24h after peak) for 2017 and 2018, while for 2019, Day 1 SSA is accurately predicted, with residuals increasing on Day 2. In comparison, events lasting 5 days show an underestimation for 2017 with negligible daily change in residuals, while the model overestimates. We therefore observe a temperature dependence of SSA decay similar to Cabanes et al. (2003). Based on limited number of events, we document low-winds having a similar effect to air temperatures below -20°C on the SSA decay rate. Our results indicate a slower rate of decay under decreased wind-speed conditions. A similar effect is observed for low temperature, as the single SSA decay event in the moderate-wind category but with mean air temperature below -20°C followed the decay rate of E14 in 2018. However, field documentation suggests intermittent snow fall during Day 2 of E14, causing increase in SSA. Consideration for environmental context is explored in Section 2.5.1. E16 is characterised by the highest initial SSA values, and the largest residuals, suggesting the model is limited at very high initial SSA values.

The model requires only initial SSA as a parameter and predicts SSA decrease at EastGRIP within the defined conditions with an averaged low-wind events.
### 3.2.1 Model evaluation

#### Table 2. RMSE - Model comparison

<table>
<thead>
<tr>
<th></th>
<th>Low-wind</th>
<th>Moderate-wind</th>
</tr>
</thead>
<tbody>
<tr>
<td>~</td>
<td>Mean</td>
<td>Individual</td>
</tr>
<tr>
<td>~</td>
<td>$m^2\text{kg}^{-1}$</td>
<td>$m^2\text{kg}^{-1}$</td>
</tr>
<tr>
<td>This Study</td>
<td>3.64</td>
<td>4.76</td>
</tr>
<tr>
<td>FZ06</td>
<td>3.45</td>
<td>7.08</td>
</tr>
<tr>
<td>T07</td>
<td>6.34</td>
<td>7.11</td>
</tr>
</tbody>
</table>

This Study uses the respective $\alpha$ and $C$ for the low- and moderate-wind events, using daily (mean) and individual samples. FZ06 parameters $\tau$ and $n$ are defined by the look-up table from Flanner and Zender (2006). T07 uses the mean surface temperature for each event.

Model performance is tested by 1) comparing daily predicted decrease to the 10 daily observations, and 2) comparing results from this study to previous models from Flanner and Zender (2006) and Taillandier et al. (2007). Model-data residuals are normally distributed, suggesting no systematic errors in model predictions. The root mean squared error (RMSE) of 5.6 between model predictions and observed SSA is $4.76\, m^2\text{kg}^{-1}$ when considering all sample sites individually. The model predicts SSA decay over 2-5 day periods ($\tau^2 = 0.89$), with the highest RMSE of 6.17 and $3.50\, m^2\text{kg}^{-1}$ for 2019 compared to 4.97 the low-wind and moderate-wind SSA decay events.

Using the physical-based decay model from Flanner and Zender (2006), hereafter referred to as FZ06, the influence of wind-speed on observed SSA decay rate can be assessed. Low-wind SSA decay events (E10 and E11) are most accurately predicted by FZ06 using the parameter values of $\tau = 3.50$ and $n = 4.72$ $m^2\text{kg}^{-1}$ for 2017 and 2018 respectively. The model adequately predicts rapid SSA decay at EastGRIP within the temperature range, while for colder temperatures, the decay rate is the same but the intercept is significantly higher (Fig. 4a). Overall, for all included events during the three sampling years, behaviour of SSA decay is clearly captured by the model (Fig. 4c,d and e). Exploring temperature conditions alone we find that the model performs well when daily mean surface temperatures are between −25°C and 0°C. 6.1 from the look-up table (Flanner and Zender, 2006). Both the empirical model from this study, and the model from Taillandier et al. (2007), hereafter T07, underestimate the rate of decrease for low-wind decay events, most apparent during the first day of the event E10 (see Fig. A3).

#### 3.2.2 Environmental-conditions during SSA decay events

Intuitively, environmental conditions would be considered to play a role for surface snow metamorphism and the rate of SSA decay. To explore this, hourly weather measurements from the PROMICE AWS and field report weather observations are analysed to provide environmental context to SSA decay events. Weather station data shows no systematic influence of basic
weather variables, relative humidity, surface temperature and wind speed on the model-data residuals, with linear regressions resulting in \( r^2 < 0.1 \) for all variables. An overview of event conditions using field observations are presented in Table A. Temperatures below \(-25^\circ C\) are characterised by the same slope defined by the model \(-0.54\). The data indicates that in natural conditions, wind-speed (between \(6\,\text{m}^2\text{kg}^{-1}\text{day}^{-1}\)), but with a significantly higher intercept of \(29\,\text{m}^{-1}\text{day}^{-1}\) compared to \(14.7\) and \(7\,\text{m}^{-1}\text{day}^{-1}\) for temperatures above \(-25\) increases the surface SSA decay rate \((\alpha \,^\circ\text{C})\). Significant wind drift is expected when hourly mean wind-speed exceeds \(6\,\text{m}_{-0.53}\,\text{a}^{-1}\), which happens during 144 days out of the total 258 sampling days from 2017, 2018 and 2019. Results indicate weather has no systematic influence on SSA decay during the first 2-5 days exposed at the surface, and that conditions vary for each event. The model is able to predict all defined decay events between \(-25\) and \(0\)\(^\circ\text{C}\), indicating mechanisms of decay are the same. Daily mean values are more accurately predicted by the SSA decay model than individual sample sites due to snow surface variability. In-homogeneous surface snow is especially important to consider for isotopic composition, because there is potential for samples to contain snow from different precipitation and/or deposition events\(-0.41\). RMSE values presented in Table 2 indicate that FZ06 predicts decay with the least error, for both wind-speed categories. Moreover, all models have lowest errors when predicts events in the moderate-wind category.

#### 3.2.2 Surface snow spatial variability

*Figures of snow isotopes and SSA* timeseries of \(\delta^{18}\text{O}\) (a), \(\delta\) excess (b) and SSA (c) for 2017, 2018 and 2019 sampling seasons. (d) shows the principle components of each parameter with colors corresponding to the color used to show absolute values. The black vertical lines indicate a break in the x-axis. Each faded line represents individual sample site values, and the thick line is the daily mean. Grey shaded regions indicate periods of high spatial variability in isotopic composition.

#### 3.3 Isotopic change decay during events

The characterization of the SSA decays provide a basis to explore how snow metamorphism of surface snow plays a role for the alteration of isotopic composition of Greenland snow after deposition. A recent study at EastGRIP has shown the significant homogeneity in surface snow due to post-depositional reworking of the snow (Zuhr et al., 2021). The focus for this manuscript is to identify signal coherence between physical properties and isotopic composition of surface snow subject to precipitation/deposition and post depositional processes. Autocorrelation analysis shows that isotopic composition values are spatially decorrelated after \(10\) m \((r^2 < 0.3\) after 10 m). Therefore, to avoid attenuation of isotopic signal, each sample is treated as independent. Isotopic composition is measured from each SSA sample containing snow from the top 2.5 cm of the snow-surface, potentially containing snow deposition layers from multiple precipitation events. Surface heterogeneity is considered by using Empirical Orthogonal Function (EOF) analysis to determine the dominant mode of variance for each sampling year. *Figure 2D* shows timeseries of-

The rate of change in SSA during low- and moderate-wind events is explored with respect to the rate of change in isotopic composition, given the covariance identified from EOF analysis. The rate of change in \(\delta\)-excess is plotted against the rate of change in SSA (Fig. 5), considering 1- and 2-day time intervals. We here include analysis of 2-day to allow isotopic
Figure 5. Isotopic change during all the analysed events are shown, with each point indicating a specific sampling site. The daily change in d-excess (d_{ex}) and SSA is presented in a), with 0 indicated with the grey dotted lines. The change in d-excess and SSA over a 2-day period is shown in b), while the change in d-excess is plotted against the absolute d-excess values is shown in c). Linear regressions are presented from daily change (light green) and 2-day change (dark green).

Equilibration between the existing surface snow and snow deposited in the day preceding the event. The change after 2-days is presented in Table 3 for each low- and moderate-wind event.

All events have an overall change in isotopic composition, with the percentage change in d-excess being an order of magnitude higher than that of $\delta^{18}$O(a), d excess (b) and SSA (c) with faded lines showing each sample site. The first principal components (PC1) of $-\delta^{18}$O, d excess and SSA are presented in Fig. ??d. All parameters continuously change throughout the field seasons of 2017, 2018 and 2019. Isotopic composition measurements (Fig. ??a, b) have larger spatial variability than SSA (Fig. ??c).

Inter-annual variability is observed corresponds to decreasing d-excess in 5 out of 8 events. E9, E11 and E13 deviate from this pattern, E9 and E13 both exhibit increases in $\delta^{18}$O with seasonal mean values of -31.6‰, -32.7‰, and -27.3‰, for 2017, 2018 and 2019 respectively (Fig. ??a-c). Note that the 2019 field season started approximately 15 days later than 2017 and 2018, resulting in a bias towards mid-summer conditions. Throughout the season d-excess, whereas E11 is characterised by a slight decrease in $\delta^{18}$O follows a gradual increasing trend from May to August. Some cases of abrupt decreases (-10 and 27% decrease in d-excess.

Using a significance level of 0.05, the relationship between change in d-excess ($\Delta$d-excess) and change in SSA ($\Delta$SSA) is assessed. The results are presented in Fig. 7. Firstly, the $\Delta$d-excess over 1-day are normally distributed around a mean of -0.3‰ are observed in the late summer, for example at July 12th in 2018 and July 25th in 2019. No clear seasonal trend is observed in d-excess (Fig. ??b) but with periods of gradual decreases. Total daily spread in $\delta^{18}$O and d excess is -15‰.

During 2017, 2018 and 2019 SSA has one dominant mode of variance (PC1) explaining 64 $\Delta$d-excess values <0.77 4% and 72% tend to correspond to smaller $\Delta$SSA (-15% of the total variance in the respective datasets. PC1 of $\delta^{18}$O explains 69 m$^2$kg$^{-1}$, 83 % and 75% of the total variance for the respective years. While PC1 of d-excess explains 47%, 51% and
for 2017, 2018 and 2019 respectively. PC1 of δ\textsuperscript{18}O and d-excess show strong coherence from May to early June during 2017 and 2018, while for the second half of the season, and throughout 2019, PC1 of d-excess corresponds to PC1 of SSA (Fig. 22d m\textsuperscript{2}kg\textsuperscript{-1}), suggesting that large decreases in d-excess occur after an extended period of exposure. This feature is highlighted in Fig.

Surface variability due to post depositional reworking of the snow is shown by a wide spread in SSA values during a given day. Time periods with low spatial variability indicate largely homogeneous snow cover over the transect, shown in Fig. ?? as shaded regions. High variability is defined by periods where the 5-day running mean of spatial variance in δ\textsuperscript{18}O is greater than one standard deviation. During periods of low spatial variability in isotopic composition, there is greater coherence between PC1 of SSA and PC1 of d-excess, due to a reduction of noise in the dataset. PC1 of SSA and d-excess show a coherence during 2018 and 2019 seasons, while the signal is less clear during 2017-7b, where d-excess decreases in 59 out of the 80 sites after two days of exposure to surface processes. Initial d-excess is observed to have a significant influence the magnitude of d-excess decrease over the defined period (Fig. ??b). However, the reduced signal coherence is concurrent with high spatial variability in isotopic composition-7c), with high initial d-excess corresponding to the largest decreases in d-excess.

A clear relationship between PC1 of SSA and PC1 of d excess is observed when there is a relatively homogeneous snow layer over the sampling transect, defined by low spatial variance in δ\textsuperscript{18}O.

Table of isotopic change for decay events

<table>
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<tr>
<th>Event</th>
<th>Initial Day-2</th>
<th>2-Day</th>
<th>% Difference</th>
<th>SSA %</th>
<th>d-excess %</th>
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</tbody>
</table>

Behaviour of snow parameters during decay events are defined. The initial, 2-day and percentage isotopic change over a 2-day period are presented for δ\textsuperscript{18}O, d-excess.

3.3.1 Low-wind event analysis

3.3.2 Isotopic change during decay events
Relationship between SSA and d-excess after the second day of each event: The relationship between the rate of change in SSA (ΔSSA 2days⁻¹) and d-excess (Δd-excess 2days⁻¹) over a 2-day period for a) individual samples for events presented in Table 3 for 2017 (○), 2018, (x) and 2019 (□), b) the same values colour coded by initial d-excess from each event. c) shows the relationship between change in d-excess after 2-days plotted against the initial d-excess value, with the linear regression line in black.

Figure 6. Isotopic change analysis for low-wind events, E10 and E11. Panel a) shows daily change in d-excess against change in d18O for E10 and E11 with corresponding linear regressions, b) shows change in d-excess against change in SSA, and c) shows change in d18O and change in SSA. The r- and p-value for each regression are indicating in the corresponding colours.

Figure 7. Latent heat flux (LE) (grey), relative humidity with respect to ice (purple), air-surface temperature gradient (TG) (red) and surface-10cm subsurface TG (red) for the low-wind SSA decay events, E10 (a) and E11 (b) (Table Dark grey shading in LE indicates sublimation and light grey shows deposition.
Using all 10 sample sites as independent values, the behaviour of isotopes during defined SSA decay events is analysed. To determine the isotopic change in the surface snow during rapid SSA decays, the rate of change in d-excess is plotted against the rate of change in SSA (Fig. 5). The change in SSA over a 2-day period is used. The daily mean change over the first 48h of each event is presented in Table 3.

In all events, the isotopic composition is observed to change, with δ¹⁸O increasing after 2 days but mostly limited to As mentioned in section 3.1, ground fog preceded the SSA peak in E11, concurrent with negligible accumulation recorded. In contrast, approximately 1±1 % mean increase, with the exception of E17 and E21 in 2019 (See Table 3). E17 is characterised by significant ground fog and snowfall during the event, while E21 has negative LHF (net deposition) measured from the eddy-covariance system over the event. The percentage change of d-excess is an order of magnitude higher than δ¹⁸O—expected due to the definition of d-excess—and similar to SSA, with 14 out of 19 events showing a decrease in d-excess during the first 2 days of each event. Further analysis looks specifically at the relationship between d-excess and SSA given the coherence observed between their PCs, and the significant change observed in Table 3. cm of snow was accumulated during the day prior to E10, corresponding to observation of snowfall.

SSA decreases by between 30% and 53% during the first 2 days, the largest change corresponding to the highest initial SSA value of 7.4 m² kg⁻¹ as defined by the decay model. Using a significance level of 0.01, Figure 6 shows the relationship between change in d-excess after the second day of each event (Δd-excess and change in SSA over the same time period (ΔSSA) is assessed. Events presented in Table 3 are shown in Fig. 7a. in isotopic composition and SSA. For E10, both Δδ¹⁸O and Δd-excess and ΔSSA and Δd-excess have significant negative correlations (r of decreases in SSA correspond to decrease in d-excess when treating each sample as an independent value. All large decreases in SSA correspond to high SSA values, as the model describes. Increases in d-excess are observed at 12 samples sites, 6 of which are during 2017 and all correspond to initial d-excess values < 5‰. (Fig. 7b) −0.5, r ≈ 0.8). Thus suggests either low d-excess of deposited snow, or old snow that has been re-exposed. In addition, initial d-excess is observed to significantly influence that magnitude of d-excess change over the subsequent 48h of rapid SSA decay (Fig. 7a and E10). The largest changes in d-excess correspond to high initial d-excess values. Moreover, increases in d-excess during rapid SSA decay follow very low initial d-excess values.

In summary, intuitively, Δδ¹⁸O and ΔSSA are positively correlated (r of 78 out of 108 samples) of cases decreases in SSA correspond to a decrease in d-excess of the snow sample during the first 2 days. Moreover, the magnitude of change in d-excess during rapid SSA decay shows a weak but significant dependence on the initial d-excess signal.

Significance of change in SSA and d-excess during events is tested by comparing the difference between the mean of daily changes for event and non-event periods using a t-test with 0.01 significance level. Background variability in d-excess is 0.1±2.5‰ for non-event periods, compared to 0.4±0.6), while no significant relationship is observed between the Δ-parameters during E11. All samples exhibit negligible change (<0.7‰ for events alone. Similarly for SSA, non-events daily change is 0.04) in δ¹⁸O during E11.

The dominant direction of vapour flux is assessed using air, surface and subsurface (10 cm depth) temperature data and LE between the snow and atmosphere. Net-sublimation is observed during both E10 and E11, with a total sum of 33.9 W m⁻² and 55.8 kg m⁻¹ compared to 7.7 W m⁻² for the respective events. The LE is controlled primarily by the temperature gradient (TG)
between the air and the surface, with strong sublimation (> kg m⁻¹) for events. SSA decay events exhibit significant difference in distribution to non-event daily changes (p < 0.01, t = 4.0070, df = 1715, Std. Err. = 0.125). Moreover, changes in d-excess during events are double the magnitude of background variability with a consistently negative sign for all years, supporting evidence that d-excess of recently deposited snow has a 72.10% chance of decreasing during surface snow metamorphism (SSA decay) during the first two days, according to our data.

Analysis shows that rapid SSA decay events correspond to decreases in d-excess over a 2-day period in 72 W% of the samples. Results from EOF analysis during periods of low spatial variance in isotopic composition over the sampling transect reveals a coherence between the dominant mode of variance of SSA and d-excess, suggesting that processes driving change in SSA also influence d-excess (m⁻²), corresponding a negative TG between the air and surface of 2.5°C on June 10th. A concurrent upwards vapour flux is indicated based on the TG between the subsurface and surface snow. Downwards LE flux up to 4 W m⁻¹ is observed each night corresponding to the transition from a negative to positive TG between the air and surface. The period between sampling on 9th June at 15:18 UTC and 10th June 10:40 UTC recorded net deposition, corresponding to a significant increase in δ¹⁸O and decrease in d-excess.

The amplitude of all parameters is during for E11 compared to E10. A negative surface-subsurface TG persists throughout the first day of E11, indicating a downwards vapour flux.

4 Discussion

Continuous daily SSA measurements at EastGRIP during the summer season of 2017, 2018 and 2019 have enabled quantification of variations in snow physical properties due to precipitation–deposition and snow metamorphism during summer. Understanding the relationship between rapid decreases in SSA and corresponding change in isotopic composition require clearly defined events and environmental context. Using a multi-day SSA decrease threshold, 21 events are defined from the summer field seasons of 2017, 2018 and 2019. All events are characterised by a peak and subsequent decay in SSA, the rate of which is proportional to the initial SSA value. Set of criteria, nine SSA decay events during precipitation-free periods are defined and used to construct an empirical decay model. We firstly discuss the behaviour of SSA decay in precipitation-free periods is driven by sublimation and vapour diffusion which is expected to influence the snow isotopic composition (Ebner et al., 2017; Hughes et al., 2021). Decay at EastGRIP compared to existing models. The isotopic change associated with low-wind SSA decay events is then considered, in the context of sublimation, vapour diffusion and wind effects (Ebner et al., 2017; Hughes et al., 2021).

In this study, we present an empirical SSA decay model for surface snow of polar ice sheets based on continuous daily SSA measurements. The model describes SSA decay under natural summer conditions on the ice sheet. The findings from this study agree with previous studies, that SSA decay is most accurately at EastGRIP. Decay model developments.

In this study, we present an empirical SSA decay model for surface snow of polar ice sheets based on continuous daily SSA measurements. The model describes SSA decay under natural summer conditions on the ice sheet. The findings from this study agree with previous studies, that SSA decay is most accurately at EastGRIP.
The empirical decay model defined in this study accurately predicts the SSA decay of surface snow at EastGRIP over a limited time-period. We find that rapid SSA decay events are best described by an exponential function (Cabanes et al., 2002), and indicates that the crystal structure of a new snow layer is a key driver of decay rate within the defined conditions over 2.5 day periods.

Comparison with weather station data showed that decay function, in agreement with observations from Cabanes et al. (2003). The expected temperature dependence on the SSA decay rate during events had no systematic influence from weather variables (wind speed, temperature and relative humidity). The only exception is for temperatures outside the set range for the model. Surface temperatures below -25°C were characterised by a significantly higher background SSA (defined as the mean SSA value of the final day of decay events) (Fig. 4), indicating high background SSA due to reduced snow metamorphism is apparent during E9, where the mean air temperature is less than -20°C, which is in agreement with the accepted knowledge that snow metamorphism is slower in colder conditions. This observation is supported by theory and observation that due to sublimation and deposition are being thermally activated processes (Cabanes et al., 2003). Taillardier et al. (2007) (T07) developed an SSA decay model with a surface temperature parameter in addition to initial SSA which is able to capture the behaviour of decay during the cold event, E7, at EastGRIP suggesting temperature is important to consider when predicting SSA outside the defined temperature range. However, (Cabanes et al., 2003; Legagneux et al., 2003; Flanner and Zender, 2006; Taillardier et al., 2007) The narrow temperature range of SSA decay events does not facilitate a conclusive definition of a temperature-dependent decay rate.

In addition, we focus on the influence of temperature on wind-speed of the SSA decay rate within the defined temperature range is negligible. Model observation comparisons show equal performance for the SSA decay model from this study ($r^2 = 0.89$) compared to T07 temperature gradient metamorphism model ($r^2 = 0.9$), and observe a more rapid SSA decay with increased wind-speed, potentially due to increased ventilation of saturated pore air acting as a catalyst for snow metamorphism (Cabanes et al., 2003; Flanner and Zender, 2006; Neumann and Waddington, 2004). Wind erosion cannot be definitively ruled out due to dis-continuous documentation of surface conditions. However, in some cases, high wind speeds are documented to increase SSA due to fragmentation and sublimation of suspended snow grains, which are then re-deposited and effectively sieved into the pore spaces of the surface snow layer (Domine et al., 2009).

The top 1 cm of the 2.5 cm SSA sample is measured by the Ice Cube device, and thus, is most likely to capture the precipitation signal (Gallet et al., 2009; Klein, 2014). Directly after precipitation, isothermal snow metamorphism is expected to be dominant due to to high surface curvature of fresh snow crystals (Colbeck, 1980). Alternative SSA decay models are proposed by Taillardier et al. (2007) to describe snow metamorphism under temperature gradient (temperature driven recrystallisation) and isothermal (curvature driven recrystallisation) metamorphism, with the surface temperature and initial SSA being variable parameters. Comparison to existing physical models allows for the assessment of the additional influence of wind-speed, not considered previously (Flanner and Zender, 2006). However, we find that all events are most accurately predicted using the temperature gradient decay equation, which accounts for the very low surface temperature observed in E7. The similarity in prediction for -25°C to 0°C suggests the EastGRIP SSA decays are not only driven by crystal curvature but by temperature gradient vapour diffusion as well.
The influence of snow metamorphism after precipitation during winter is expected to be reduced due to low temperatures and negligible temperature gradients during polar night. Based on this, the model is only recommended to use for polar ice sheet summer conditions only. Within the defined conditions, FZ06 most predicts the moderate-wind events with the lowest error. This is potentially due to the SSA decay model is a simple empirical-model-initial conditions for low-wind event E10 likely corresponding to surface hoar, while the models from the literature tend to describe SSA decay in the accumulation regions of the Greenland Ice Sheet, with dependence on the initial SSA alone, from precipitation. The initial SSA value of 46 m² kg⁻¹ for E10 is in agreement with documented SSA of surface hoar (Domine et al., 2009).

4.1 Decay model applications

Conditions for the model are expected to be applicable over the Greenland Ice Sheet interior under mean summer conditions. The model predicts decay events at EastGRIP with a r² of 0.89, compared to observation, within defined conditions. SSA estimates from satellites have previously been compared to ground observations and show a strong correlation between daily mean SSA and satellite-retrieved SSA at EastGRIP (Kokhanovsky et al., 2019). The SSA decay model has the potential to predict SSA decay. Modelling SSA decay using continuous in-situ measurements is associated with a number of limitations relating to surface perturbation by the wind and hoar formation, but nevertheless, is vital for studying surface energy balance and post-depositional change in isotopic composition. To test the model over the entire accumulation zone of the Greenland Ice Sheet-GIS using satellite data, the model can be evaluated for different sites to document the spatial variability in SSA over the entire ice sheet, and describe the summer SSA decay. This has additional benefits for quantification of surface mass balance and surface energy budget due to the relationship between snow microstructure and surface albedo. Following the methods in Kokhanovsky et al. (2019) would be an interesting future study, but is outside the scope of this manuscript.

4.1 Rapid SSA decay and d-excess Inter-annual variability

In this study, processes driving snow metamorphism are documented to influence isotopic composition of the snow after precipitation, supporting experimental observations and theoretical understanding (Ebner et al., 2017; Wahl et al., 2021; Hughes et al., 2021). Results from this study suggest that surface snow metamorphism following precipitation events corresponds to change in isotopic composition, most clearly observed in d-excess (Table 3). The surface snow over the 90 m sampling transect is often non-homogeneous due to uneven distribution of accumulation. EOF analysis is used to account for spatial variability at each site, and to determine covariance between the parameters SSA, δ¹⁸O, and d-excess. The positive mode of PC₁SSA is associated with depositional events, such as precipitation, surface hoar formation, and wind-fragmented snow drift, causing an increase in SSA (Domine et al., 2009), while the negative mode is associated with snow metamorphism or wind scouring (Cabanès et al., 2002, 2003; Legagneux et al., 2003, 2004; Taillandier et al., 2007; Flanner and Zender, 2006). Based on our results, rapid decreases in SSA correspond to decreases in d-excess of a new snow layer in 72% of cases during the first 2 days of rapid SSA decay. This interpretation, correlations between PC₁SSA and PC₁d-excess or PC₁δ¹⁸O suggests the aforementioned mechanisms controlling SSA variability also influence the isotopic composition.
Using the eddy-covariance latent heat flux measurements, we observed net sublimation during all decay events (with the exception of E21) used for isotopic analysis, which is in agreement with recent studies that document fractionation during sublimation results in slight increases.

Accumulation intermittency and temperature conditions are proposed as a potential explanation for the change in regime from a coherence between PC1$\delta^{18}O$ and PC1$_{d-excess}$ in 2018 and PC1$_{SSA}$ and PC1$_{d-excess}$ in $\delta^{18}O$ and decreases in d excess (Madsen et al., 2019; Hughes et al., 2021; Wahl et al., 2021). However, sublimation is not the only process occurring. Vapour pressure gradients due to surface curvature drive snow metamorphism via vapour diffusion through the pore space and thus kinetic fractionation is expected to influence the isotopic composition. A larger influence is expected for d-excess than $\delta^{18}O$ because kinetic fractionation influences $\delta D$ more than $\delta^{18}O$ ($d-excess = \delta D - 8 \cdot \delta^{18}O$) with a stronger influence on d-excess than $\delta^{18}O$, which can explain the covariance between d-excess and SSA observed most clearly during 2019 (Cappa et al., 2003; Dadic et al., 2015).

Our approach to 2019, Casado et al. (2021) show that during low precipitation periods in Antarctica, the isotopic signal is strongly modified during snow metamorphism. Approximately 10 cm of accumulation is recorded in both 2018 and 2019, but a gradual increase during 2018 suggests multiple small deposition events, whereas 2019 is characterised by step-like increases. Therefore, the change over a 2-day period instead of daily change allows for increased propagation of the isotope signal during SSA decay to account for the strong correlation between PC1$_{SSA}$ and PC1$_{d-excess}$ can be attributed to increased surface exposure and warmer temperatures facilitating snow metamorphism, in agreement with findings from Casado et al. (2021).

Low accumulation during 2017 presents a caveat to this interpretation, with results from 2017 showing PC1$_{d-excess}$ to be influence by both PC1$_{SSA}$ and PC1$\delta^{18}O$ during different periods. The period from May 15th to June 10th follows the regime observed during 2018 and corresponds to a negligible temperature gradient between the air, surface, and 10 cm representation from Ice Cube SSA measurements, compared to the 2.5 cm bulk isotope measurements (Gallet et al., 2009; Klein, 2014). A significant relationship is observed between change in d-excess and change in SSA during the first 2 days compared to daily analysis (with an additional relationship observed during 2019 between daily change in d-excess and daily change in SSA).

Decreases in d-excess are observed during rapid SSA decay, driven by a combination of sublimation, deposition and vapour diffusion through the pore space.

Surface snow metamorphism is not confined to rapid SSA decreases, and thus isotopic composition change is observed continuously. However, results from this study indicate that d-excess changes during rapid SSA decay have significantly different distribution than the background non-event fluctuations. Our findings are in agreement with a study from Antarctica which showed a significant relationship between d excess and physical snow properties with depth, while negligible relationship was observed for $\delta^{18}O$ (Dadic et al., 2015). Our study has selected rapid SSA decays fitting to subsurface (Fig. A4). In contrast, the period from July 1st onwards is characterised by a near-constant upwards vapour flux, indicated by a negative temperature gradient between the decay model to address how changes in snow crystal morphology after precipitation relates to change in isotopic composition. Future studies would benefit from using isotope flux models to account for the influence of sublimation and deposition, to determine unexplained isotopic composition change air, surface, and subsurface. PC1$_{d-excess}$ covaries with PC1$_{SSA}$ during this period, much like 2019, suggesting that vapour diffusion driven by temperature gradients modifies the
snow isotopic composition. This agrees with previous studies documenting kinetic effects during snow grain growth resulting from pore space diffusion (Neumann and Waddington, 2004; Casado et al., 2016; Ebner et al., 2017; Casado et al., 2021).

An additional feature supporting the observation of processes driving surface snow metamorphism corresponds to a decrease in d-excess, is a clear relationship between substantial increases in SSA and increase in d-excess (Fig. ??). The upper 10th percentile of ∆SSA increases (14.7 m² kg⁻¹) corresponds to positive ∆d excess in 70% of cases (Fig. ??). Large increases in SSA are closely associated with precipitation, however, increases are observed in The opposing phases of the North Atlantic Oscillation (NAO) between the years can explain the different meteorological conditions. The NAO is in a number of other scenarios (Domine et al., 2009). Precipitation is expected to cause the largest SSA, suggesting that the d-excess of precipitation is most often higher than existing surface snow. Our results therefore suggest that the precipitation isotopic composition signal is not always preserved after snow metamorphism due to (kinetic) fractionation during sublimation and other surface processes—positive phase during 2018 and the majority of 2017 bringing below-average temperatures, as observed at EastGRIP (Hanna et al., 2015). The opposite is observed during 2019, corresponding to a positive phase in the NAO.

Change in d-excess per day (Δd-excess day⁻¹) vs. change in SSA per day (ΔSSAday⁻¹) The relationship between the rate of change in SSA per day (ΔSSAday⁻¹) and d-excess (Δd excess day⁻¹) for all summer seasons 2017-2019 (light grey), all events (dark grey) and selected events based on substantial accumulation (dark turquoise). The box indicates the values corresponding to daily decrease in d-excess during decrease in SSA, with 81% of selected events in this quadrant. Conclusive results from EOF analysis are limited by wind-effects, especially in the negative phase, corresponding to decrease in SSA, where wind scouring potentially removes the surface layer (Domine et al., 2009; Flanner and Zender, 2006; Hachikubo et al., 2014).

Decoupling snow metamorphism from wind scouring is considered in the following section on isotopic change during low-wind SSA decay events.

4.2 Influence of event conditions on isotopic change

Surface conditions prior to and during SSA decay events

Three key mechanisms are expected to drive the rapid SSA decays, 1) large grains growing at the expense of small grains (Legagneux et al., 2004; Flanner and Zender, 2006), 2) diffusion of interstitial water vapour (Colbeck, 1983; Ebner et al., 2017; Touzeau et al., 2016). The dominant mechanisms can theoretically be identified by a combination of the change in isotopic composition - indicating the fractionation effect - and the LE and temperature gradient data.

In theory, mechanism 1) causes minimal change in the bulk isotopic composition of a snow layer under isothermal conditions (Ebner et al., 2017). Therefore, observations of SSA decay corresponding to negligible isotopic composition change could be explained by this mechanism. We observe no events with consistent isotopic composition throughout. In the instance of 2) interstitial diffusion, light isotopes are preferentially diffused, while the heavy isotopes will be preferentially deposited onto the cold snow grains (Colbeck, 1983; Ebner et al., 2017; Touzeau et al., 2016). Thus, diffusion of water vapour in the
pore space causes a decrease in $d$-excess and slight increases in $\delta^{18}O$ due to kinetic fractionation (Flanner and Zender, 2006). Sublimation has been widely documented to cause an increase in $\delta^{18}O$ of the remaining snow-mass due to equilibrium fractionation, and a significant decrease in $d$-excess due to kinetic fractionation (Ritter et al., 2016; Madsen et al., 2019; Hughes et al., 2021; 

An overall increase in $\delta^{18}O$ and decrease in $d$-excess during E10 can be attributed to a combination of 2) and 3) based on observation of net-sublimation and high amplitude diurnal temperature gradient variability indicating vapour transport within the pore space. The period between 9th June at 15:18 UTC and 10th June 10:40 UTC recorded net deposition corresponding to an overall decrease in $\delta^{18}O$ during SSA decay events vary, with a number of events having no measured accumulation or observed snowfall (Fig. ??). Removing events with non-homogeneous increases in surface height and events where additional precipitation or significant snowdrift are observed, reveals that during rapid SSA decays following significant precipitation, there is increased likelihood of observing concurrent decrease in $d$-excess during the first day (Fig. ??). This observation combined with results presented in Fig. 7a strongly suggests that initial snow metamorphism after precipitation and minimal decrease in $d$-excess, potentially due a deposition of atmospheric water vapour (Stenni et al., 2016; Feher et al., 2021; Casado et al., 2021)

A 30% decrease in $d$-excess corresponds to a decrease in $d$-excess of in the surface snow, negligible change in $\delta^{18}O$ during E11. Net-sublimation double that of E10 is measured, but with reduced amplitude in both TGs. Moreover, the largest decrease in $d$-excess occurs after the first day when the surface-subsurface TG is consistently negative, indicating that vapour diffusion plays a role in modifying the isotopic composition, and the effect of equilibrium fractionation during sublimation from the surface only weakly influences the bulk isotopic composition over the 3-day period (Casado et al., 2021). Decoupling the influence of atmosphere-surface exchange and diffusion from subsurface snow requires additional measurements of isotopic composition of atmospheric water vapour and precipitation isotopes, which is outside the scope of this study.

### 4.3 Spatial variability of snow surface

Low accumulation rates at EastGRIP result in the potential for winter snow layers to influence the isotopic composition in the 2.5cm surface snow. Accumulation heterogeneity causes uneven mixing of layers at each sample site, which is observed clearly in the large spatial variability in isotopic composition measurements in Fig. ??a and b. EOF analysis is used to account for spatial variability at each site, and a coherence is observed between the principal components of $d$-excess and SSA. PC1 is weaker when spatial variability is high, and during these periods the coherence between $d$-excess and SSA are muted. During the start of 2017 and 2018 PC1 of $d$-excess is coherent with PC1 of $\delta^{18}O$, and decoupled from PC1 of SSA. At the start of the season, the 2.5cm sample will contain winter snow layers which are less influenced by snow metamorphism (Libois et al., 2015; Town et al., 2008), and thus, a coherent signal between $d$-excess and $\delta^{18}O$ is observed. The transition to a coherence between PC1 of $d$-excess and PC1 of SSA can be explained by summer snow layers, influenced by snow metamorphism, causing $d$-excess to appear to become decoupled from $\delta^{18}O$, which is less influenced by kinetic fractionation than $\delta D$ (Masson-Delmotte et al., 2005) during snow metamorphism.
4.3 Implications to ice core interpretation

4.3 Implications and perspectives

Documented changes in snow isotopic composition during surface snow metamorphism have potential implications for interpretation of stable water isotope records from ice cores, given that the current interpretation assumes the precipitation signal is preserved (Dansgaard, 1964). Seasonal transition from a coupling of PC1. Our results suggest that processes driving snow metamorphism modify the isotopic composition of d-excess and PC1 of δ¹⁸O, to a coherence between PC1 of d-excess PC1 of SSA at the latter part of the season, suggest that summer snow metamorphism causes d-excess to appear to decouple from δ¹⁸O. Kinetic fractionation during sublimation-the snow while exposed at the surface, supporting experimental observations and theoretical understanding (Ebner et al., 2017; Wahl et al., 2021; Hughes et al., 2021). We find that d-excess is mostly influenced by vapour fluxes in the pore space, driven by temperature gradients. Net-sublimation appeared to have less influence on the isotopic composition, but this is expected to be the cause a decrease in d-excess in the snow, given the different diffusivities of HDO and and H₂¹⁸O (Masson-Delmotte et al., 2005). due to the depth of the sample and the short duration of both low-wind events.

Seasonal signals are influenced by millennial scale insolation variability (Masson-Delmotte et al., 2006; Laepple et al., 2011). An inverse relationship is observed between obliquity and d-excess over the past 250 ka years at Vostok which is attributed to the insolation gradient between high and low latitudes causing increases moisture transport from low-latitude relative to high latitudes (Vimeux et al., 2001, 1999). Results presented in our study document decreases in snow d-excess during surface snow metamorphism. Millennial scale local insolation variability has a strong influence on temperature gradients in the snow (Hutterli et al., 2009). Thus, it is possible that local insolation variability may also influence d-excess due to temperature gradients in the snow driving snow metamorphism at the surface—

Our results highlight the need to consider the influence of surface snow metamorphism on isotopic composition in stable water isotope records as the traditional interpretation of d-excess ice core signal does not account for any post-depositional signal. The findings of this exploratory study reiterates the importance of quantifying the isotopic fractionation effects associated with processes driving snow metamorphism during precipitation free periods. Moreover, the inter-annual variability observed at EastGRIP between 2018 and 2019 suggests that precipitation intermittency and temperature (gradients) play a role in isotopic change, which is not so readily identified in the surface snow SSA data. Future work to decouple the processes driving change in d-excess-d-excess (sublimation from surface or interstitial vapour diffusion in the pore space) is vital for modelling the change in isotopic composition down to the close-off depth in the firm (Touzeau et al., 2018; Neumann and Waddington, 2004). In addition, it would be beneficial to obtain Future studies would benefit from obtaining direct measurements of the isotopic composition and SSA of precipitation of precipitation and surface hoar, to determine the fraction of precipitation such deposits in the SSA samples. Furthermore, a quantitative representation of vapour fluxes in the surface snow would provide a basis from which to quantify the the relative influence of fractionation during sublimation and interstitial diffusion.
This study addresses the rapid SSA decay driven by surface snow metamorphism. In particular, the study aims to explore how rapid SSA decay relates to changes in isotopic composition of the surface snow in the dry accumulation zone of the Greenland Ice Sheet. Ten individual snow samples were collected on a daily basis at EastGRIP in the period between May and August of 2017, 2018 and 2019. SSA and isotopic composition was measured for each sample. Periods of snow metamorphism after precipitation-deposition events are defined using SSA measurements to extract periods of rapid decreases in SSA.

An exponential SSA decay model \( \text{SSA}(t) = \text{SSA}_0 e^{-0.54t} + 26.8 \text{SSA}(t) = \text{SSA}_0 e^{-0.54t} + C \) was constructed to describe surface snow metamorphism under mean summer conditions for polar snow, with surface temperatures between \(-25^\circ\text{C}\) and \(0^\circ\text{C}\) and wind speeds below \(6 \text{ m s}^{-1}\). The empirical model can be applied to remote areas of polar ice sheets and requires only initial SSA as the parameter, making it simple to use. Two categories were defined to assess the influence of wind-speed on the SSA decay rate. The relationship between defined events of snow metamorphism and corresponding snow isotopic composition was then explored.

We observe changes in isotopic composition corresponding to post-depositional processes driving rapid SSA decay. Principal components from EOF analysis for SSA and d-excess indicate that under near-homogeneous surface snow conditions, d-excess varies in phase with SSA throughout a large proportion of the sampling seasons. This suggests that post depositional processes and precipitation influence both physical snow structure and isotopic composition concurrently. SSA decay is observed in all events. Over the first 2 days of rapid SSA decay events, d-excess is observed to decrease significantly from the initial value for most events, at the same time we observe net sublimation. Significant changes in surface snow d-excess are observed during days following a precipitation event, suggesting that precipitation d-excess signal is altered after deposition, together with changes in physical snow properties (SSA)\(^{-1}\). Analysis of SSA decay events with consistent low wind speed indicates that the combined effects of vapour diffusion and diurnal LE variability causes isotopic fractionation of the surface snow in the absence of precipitation.

In summary, our results suggest that the precipitation isotopic composition signal is not always preserved due to isotopic fractionation during the processes driving surface snow metamorphism. Observations of post-depositional decrease in d-excess during rapid SSA decay hints to local processes influencing the d-excess signal and therefore an interpretation as source region signal is implausible.
## Appendix A

### Table A1. SSA Decay Event Conditions

Duration and conditions for all 21 events defined by the threshold. 'Initial Conditions' refers to the conditions during the day (24h) before the event, while 'SSA Decay Event Conditions' describes the dominant conditions for the event duration, based on field observations. 'Surface Temperature' is the mean surface temperature during the event. 'Comments' highlight any significant weather behaviour during the event.

<table>
<thead>
<tr>
<th>Date</th>
<th>Event No.</th>
<th>Surface Temperature</th>
<th>Initial Conditions</th>
<th>Event Conditions</th>
<th>Comments</th>
</tr>
</thead>
<tbody>
<tr>
<td>2017</td>
<td>27/05 - 01/06</td>
<td>E1 -17.3</td>
<td>No clear driver</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>19/06 - 24/06</td>
<td>E2 -13.6</td>
<td>Snowfall</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>30/06 - 02/07</td>
<td>E3 -14.0</td>
<td>Snowfall</td>
<td>Overcast</td>
<td>Snow drift Day-0</td>
</tr>
<tr>
<td></td>
<td>10/07 - 15/07</td>
<td>E4 -13.2</td>
<td>Snowfall</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>18/07 - 19/07</td>
<td>E5 -11.7</td>
<td>Snowfall</td>
<td>Overcast</td>
<td></td>
</tr>
<tr>
<td></td>
<td>21/07 - 23/07</td>
<td>E6 -11.2</td>
<td>Snowfall</td>
<td>Overcast</td>
<td></td>
</tr>
<tr>
<td>2018</td>
<td>07/05 - 10/05</td>
<td>E7 -33.7</td>
<td>Drift and fog</td>
<td>Clear/ice-fog</td>
<td>Snowfall Day-2</td>
</tr>
<tr>
<td></td>
<td>14/05 - 15/05</td>
<td>E8 -19.8</td>
<td>Snowfall</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>16/05 - 18/05</td>
<td>E9 -21.5</td>
<td>Snowfall and fog</td>
<td>Overcast</td>
<td></td>
</tr>
<tr>
<td></td>
<td>09/06 - 11/06</td>
<td>E10 -14.9</td>
<td>Ground fog</td>
<td>Overcast</td>
<td></td>
</tr>
<tr>
<td></td>
<td>27/06 - 29/06</td>
<td>E11 -15.3</td>
<td>Ground fog</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>30/06 - 03/07</td>
<td>E12 -11.2</td>
<td>Wind drifted snow</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>04/07 - 06/07</td>
<td>E13 -10.2</td>
<td>Snowfall</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>16/07 - 21/07</td>
<td>E14 -14.3</td>
<td>No clear driver</td>
<td>Clear-sky</td>
<td>Dusting of snow</td>
</tr>
<tr>
<td></td>
<td>23/07 - 27/07</td>
<td>E15 -14.1</td>
<td>Ground fog</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td>2019</td>
<td>17/06 - 20/06</td>
<td>E16 -11.4</td>
<td>Snowfall</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>27/06 - 30/06</td>
<td>E17 -9.5</td>
<td>No clear driver</td>
<td>Overcast</td>
<td>Fog and snow</td>
</tr>
<tr>
<td></td>
<td>02/07 - 05/07</td>
<td>E18 -7.0</td>
<td>Snowfall</td>
<td>Overcast</td>
<td></td>
</tr>
<tr>
<td></td>
<td>06/07 - 08/07</td>
<td>E19 -10.0</td>
<td>No clear driver</td>
<td>Clear-sky</td>
<td></td>
</tr>
<tr>
<td></td>
<td>18/07 - 20/07</td>
<td>E20 -7.6</td>
<td>Ground fog</td>
<td>Overcast</td>
<td></td>
</tr>
<tr>
<td></td>
<td>28/07 - 31/07</td>
<td>E21 -6.5</td>
<td>No clear driver</td>
<td>Clear-sky</td>
<td></td>
</tr>
</tbody>
</table>

Duration and conditions for all 21 events defined by the threshold. 'Initial Conditions' refers to the conditions during the day (24h) before the event, while 'Event Conditions' describes the dominant conditions for the event duration, based on field observations. 'Surface Temperature' is the mean surface temperature during the event. 'Comments' highlight any significant weather behaviour during the event.

Accumulation at each sample site: Accumulation measurements from each sample site over the 90 m sampling transect is shown here for 2017, 2018 and 2019 respectively. Each line represents an individual site. Negative values indicate a decrease in surface height, and positive values suggest precipitation or deposition adding to the surface height. The grey bars show the individual events defined in Section 3.1.
Figure A1. **Wind-speed Distribution**

Histograms showing a) the daily maximum values and b) the 10-minute mean values for all sampling days of 2017, 2018 and 2019. The black line indicates the mean.

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**Data availability.** The SSA, density and accumulation data for all sampling years is available on the PANGAEA database with the DOI:***. Snow isotope data is also available on the PANGAEA database with the DOI:***. Data from the Programme for Monitoring of the Greenland
Figure A3. *Decay Model Construction and Predictions*

A comparison between the observations, the decay models from this study and the existing decay models from Flanner and Zender (2006), FZ06, and Taillandier et al. (2007), T07. The 10-minute averaged wind-speed is shown on the secondary y-axis, with the 6 ms⁻¹ thresholds indicated. The low wind events E10 and E11 are shown in a) and b), and examples of two moderate-wind events are show in c) and d).

Ice Sheet (PROMICE) 400 were provided by the Geological Survey of Denmark and Greenland (GEUS) at http://www.promice.dk. Eddy Covariance Tower measurement are available on the PANGAEA database with the DOI: https://doi.org/10.1594/PANGAEA.928827.

Author contributions. HCSL, AKF and RHS designed the study together. AKF, SW, MH, MB, AZ, SK and HCSL carried out the data collection and measurements. RHS, AKF and HCSL worked directly with the data. RHS, AKF and HCSL prepared the manuscript with contributions from all co-authors. AKF contributed largely to the manuscript text and structure. HCSL designed and administrated the SNOWISO project.
Figure A4. *Air, surface and subsurface temperature time-series*

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