Report #1

Suggestions for revision or reasons for rejection (will be published if the paper is accepted for final publication)

General comments

First of all, I am glad to see the authors have put great effort into improving the (methods behind the) manuscript, partly based on my suggestions. My main point of concern was that their energy balance model did not allow for changes in incoming longwave radiation, surface temperature and albedo, with associated feedbacks. The authors have now extended their model to include all these processes, which makes the results more consistent and more convincing. Furthermore, they have clarified the model description and revised the section based on the NARR data, such that it is in line with the rest of the manuscript.

Due to these changes, the methods behind the manuscript have improved significantly. However, I still have some comments on the manuscript, as outlined below.

The most important is that the revised manuscript is considerably longer than the first version and I do not think this is an improvement in all places. For the model description, it was indeed necessary to include more detail. The Introduction has also increased in length. Although it reads smoothly, it is very long for an introduction and I suggest to make it more concise. The Discussion section has been rewritten completely. The present form reads like an evaluation of the results from the three different experiments grouped per perturbed variable. In my opinion, it is primarily a repetition of previously presented results with little discussion added. The last two paragraphs deal with suggested model improvements (without new subtitle...) and are more at place in a discussion section. On the other hand, some results are (also) discussed in the Conclusions section, which is also rather lengthy. I would suggest the authors to look critically at the Discussion and Conclusions sections and rewrite them. They should make sure that repetition is kept to a minimum and that no new discussion items are introduced in the Conclusions. Furthermore, they could perhaps address the representativity of their results for other parts of the glacier and shortly discuss the applicability of their methods on other glaciers.

Thanks to the reviewer for these suggestions. We rewrote much of the manuscript for the first resubmission, and it is true that parts of the discussion and conclusion were redundant. We have restructured the discussion and conclusion following the suggestions of both reviewers, and it is now shorter. Most of the 'summary' content in the discussion and conclusions that was covered in the results has been removed. The new content about insolation in the conclusions has been removed, and we bring in a little bit about glacier-wide applications in the discussion (ll. 970-985). The introduction has also been shortened by about 10%, but remains longer than the original submission in the interest of giving proper attention to some previous work on energy balance sensitivity studies.

The effect of the snowpack depth at the beginning of the melt season is now also investigated, which is definitely interesting. However, I am not convinced by the method used. It is not clear to

me how the experiment is conducted, it seems like the model is only run for one year with averaged daily meteorological values (also mentioned below). The new figure (Fig. 7) shows small effects of initial snowpack depth for depths above 1.2 m w.e., but the values fluctuate around zero. I do not understand where this variability comes from, it seems random. Is it because snowfall events are prescribed randomly and would it not be better to keep the timing of snowfall events the same for all runs? Furthermore, I am a bit surprised by the low albedo prescribed for snow remaining all summer. It is only slightly larger than the ice albedo; whether the surface consists of snow or ice after day 220 makes little difference for the energy balance. Is this realistic?

This is useful feedback, and we are certainly open to ideas here. The reviewer is correct – we were running for only one year, using average daily weather conditions for the period 2002-2012. Our wish is to isolate the effects of the initial snowpack, so this seems like a sensible way to do this: repeat the same weather but for different initial (May 1) snow conditions. But it is true the averaging gives a weather time series that is not real. We now do this differently, running for the 11 years of actual weather but over the suite of initial snowpacks. In the end it does not change much, as we are averaging the final result for presentation.

Yes, the random element in these graphs comes from the summer snowfall events – they are an internal part of the code and we left this on through these experiments. This could be specified to control for this, to make the graph cleaner, although the process is separate from and independent of the winter snow extent so it is not systematically interfering with the experiment. It just gives some variability between realizations. We left this on, but comment on it and note that the summer snow gives 'internal variability' of about $Q_N = 1 \text{ W/m}^2$, averaged over the summer (control experiments, not shown).

The final point, concerning the minimum snow albedo, is insightful and the reviewer is quite right that this 'old snow' value (0.3 for us) is very low and explains why the difference between exposed ice and old snow did not matter very much in the previous submission. Our number is based on observations from Haig Glacier firn, and our default treatment was to set the minimum snow albedo to that of firn. But this is may not be appropriate – firn on this glacier has an accumulation of impurities, similar to what occurs in ablation zones, so it is darker than old seasonal snow. Values of wet, impurity-rich, late-July snow at the AWS site on Haig are about 0.36. This may still be darker than values higher up in the glacier accumulation area, or at other glacier sites. To be a bit more 'typical' we set a new default value of 0.4 for the minimum snow albedo. All experiments in the paper now use this value. Figure 7c is new, showing the net energy sensitivity to winter mass balance for both $\alpha_{min} = 0.3$ and 0.4. Results are indeed sensitive to this choice for the late-summer energy balance. As expected, higher values of α_{min} give a stronger influence of a deep winter snowpack, although the graph is still asymmetric – a shallow winter snowpack leads to large increases in summer melt, while a deep winter snowpack moderately reduces melt at this site. This result depends on how close one is to the ELA – lower on the glacier, the result would be reversed. This is discussed in the text.

Regarding the figures, I would strongly recommend the authors to add legends explaining the different lines/symbols. Every figure shows new variables, many with multiple lines. Determining the meaning of all lines from the (sometimes incorrect, see below) captions is complicated and unnecessary. Related to this, I would suggest to show a smaller variation of fluxes (with standard colours) to make the figures more consistent. For example, the turbulent fluxes are sometimes shown separately, sometimes combined and then in another figure combined with the subsurface heat flux.

This is a good suggestion – there is not always room to add legends, so we put this information in the captions and tried to stay consistent, but agreed that we have too many lines and colours, and it changes every plot. We now have legends for each figure, where applicable. We also simplified a bit, e.g. removed Q_C in Figure 3, and radiation and turbulent fluxes are now combined in Fig 6 to reduce the amount of information.

Where the figures are discussed in the text, the authors sometimes refer to the specific line colour in the text. This should be avoided, it should be easy to derive from the figures, by means of a legend. In general, the authors may try to refer less directly to the figures and tables, by only adding references in brackets and not in the main text.

Revised as suggested throughout the text.

Some detailed comments

16 (and elsewhere): The authors now investigate the effect of changes in winter snowpack depth. However, they refer to this variable as 'winter snowpack', while they should add a measurable quantity like 'depth' or 'thickness'.

Revised as suggested, 1.16 and in Section 4 (discussion of Figure 7).

127-128: Positive net energy will not drive subsurface warming, as this has already been taken care of by Q_C.

Revised, 1.112

133-134: If the unit is given for Eq. (3), the unit for Eq. (2) should also be given (W m-2), as one follows from the other.

Added, 1.105

245: 'phi_t(z)'

Revised, 1.231

249-260: How is the refreezing rate calculated?

We added a brief explanation on this, ll.234-238. In essence through an enthalpy model. If liquid water is present in the pore space and conductive cooling gives an energy deficit, the available

'negative energy' is diverted to latent heat of freezing; temperature are not allowed to cool in a layer until liquid water content $\theta w = 0$.

293: In what sense is the grid fixed, with respect to the surface or a reference layer? How are changes in snow depth and ice surface lowering incorporated? Are layers added/removed or is the grid shifted?

Good questions, we briefly explain this now, 1.282 and 11.293-296. It is a fixed grid with respect to the surface, to a depth of 10 m (irregularly spaced, with nz=33). Near-surface layers are 10 cm thick. Snow depth d is modelled in a sort of Lagrangian sense, to the mm, so it is allowed to continuously accumulate, melt, or undergo densification (on a daily time step). Then at depth d below the surface, the grid cell has a weighted combination of thermal properties and densities to reflect the mixture of snow and either firn or ice in that layer. There is also a discrete step involved: every time 10 cm of snow accrues or ablates, the grid is shifted to propagate up/down the internal density/liquid water/ice layer structure. We hope this makes sense – happy to explain this further but we don't want to go sideways in this paper on the details of the subsurface model. It probably needs to be described elsewhere in proper detail.

340-341: The reference to Eq. (11) is not correct.

Revised, 1.330

355-356: Why is no aging included for summer snow events like for the seasonal snow pack?

It is, the clock starts again and albedo will decay. But this does not happen much as summer snows usually melt within 1-2 days. Clarified, 1.346

360: I am a bit confused that internal melting can occur in this model. The main source would be penetration of solar radiation, but this is not included here. Where does the melt energy come from and is it a large term?

Quite right, there should be no internal melting since we don't account for shortwave radiation penetration or meltwater/rainwater temperatures above 0°C. It is built into the code as an option (when we get to some attempt to include these processes), but is a 'latent' option right now. This statement has been removed, 1.353.

442-449: The main reason to use JJA for the theoretical sensitivities is that the surface is at the melting point, as a good approximation. This is not mentioned here. Please also replace 'here' by 'in this section' or 'to calculate the theoretical sensitivity' to stress the contrast with 'the next section' with the 'modelled sensitivity', where the full melt season is considered.

True of course, now noted, 1.433. Clarification on 11.444-447 for the second point.

679 (and Table 4): Not MJJAS melt energy as mentioned before?

No, we are trying our best to make it shorter and more focused where we can - so as of the first re-submission we now report only JJA, although we run the model year-round for the 11-year period (including May and Sept melt). The sensitivities in MJJAS are not so different from JJA, so we are sticking to that to increase the focus.

758-759: How is the energy balance model forced with this mean annual record? Do I understand it correctly that this record spans one year and has the mean value over the entire period for each day and each variable? This seems rather artificial to me, it does not represent real meteorological situations anymore. Why not run over the entire period using the same winter snow pack depth for each year per run?

Discussed above. It is true, averaging makes for an unrealistic time series in lots of ways, not too hot and not too cold. This can certainly be done as suggested, and it is consistent with the other perturbations (i.e. 11 realizations for each meteorological anomaly). We reworked this section thoroughly, starting at 1.710.

789-793: Why are standard deviations not compared over the same period (2002-2012)?

That is a fair point, we are interested in the variability over the full NARR period, 1979-2014, but it is probably not appropriate to compare those numbers to the 11 years of observation. We revised this to report the NARR variances over the common period, 2002-2012, ll.761-764. Values did not change much.

1021: Refreezing of melt water acts as an energy source (not sink) through release of latent heat.

Thanks yes, this was loose language. We were thinking of the energy sink as the positive net energy that is required to re-melt refrozen meltwater. At night when there is sometimes refreezing, the energy that is released is often dissipated (e.g., as QC to the surface, LW emissions, etc). Then the next day new energy is required to melt some 'recycled' meltwater (the overnight ice crust). This text has been removed as part of the discussion rewrite.

759-760: The range is 0.35-2.35 in the caption of Fig. 7.

Revised, we ran in the end for 0.36-2.36 (the observed mean, 1.36, \pm 1)

Fig. 2b: As also suggested before: Either note that KG indicates Kwadacha Glacier, which is mentioned once in the paper or remove the dot and zoom in on the map around Haig Glacier. I suggest to do the latter.

Revised as suggested

Fig. 3: Better show net shortwave and net longwave radiation instead of incoming shortwave radiation and net radiation, then all fluxes are shown exactly once. I would also suggest to use a long horizontal axis for both plots and show them above/below each other instead of next to each other. Then the interannual variation is more clearly visible and corresponding days can be compared.

These are good suggestions, adopted. QC is also removed.

Fig. 6: Net radiation is mentioned twice in the caption.

The black line is net energy, thank you for noticing, it is revised now.

Figs. 9 and 10: Numbering of the figures is incorrect, the numbers are still from the previous manuscript.

Apologies, amended. These figures were not meant to be included – we removed this part of the analysis and discussion in the first revision, but the figures lingered.

Fig. 10: Why are the turbulent fluxes both underestimated with the NARR forcing? The meteorological variables show good correspondence...

These figures and the associated results are no longer discussed, as we chose to focus just on the sensitivities. But for the reviewer's interest, this problem went away with the revised code – it was because we were assuming a melting surface in the initial submission. Now that Ts is internally modelled with the subsurface temperatures, the modelled turbulent fluxes are much improved. But no longer discussed....

Report #2

Referee review for manuscript tc-2016-06-manuscript-version-3, submitted for publication in The Cryosphere

"Surface Energy Balance Sensitivity to Meteorological Variability on Haig Glacier, Canadian Rocky Mountains" by S. Ebrahimi and S. J. Marshall

General Comments

The revised manuscript is an improvement on the initial submission. In particular, the model now includes more dynamic feedbacks which has increased the confidence with which the results can be interpreted. The re-setting of the NARR results as an extension of the sensitivity analysis is good to see and improves the focus of the paper. The authors appear to have considered and addressed the main points in their response to my first review. However, it was difficult to assess the actual changes made to address each point as no text changes were given in the response and a marked up version of the manuscript was not supplied.

Apologies, the overall was so large that we did not quote specific revisions or invoke 'track changes'. The first draft was almost completely written in accord with the reviewers' suggestions, so this did not seem productive. In our second round of revisions, we use track-changes mode and the specific changes can be seen – still very extensive in the rewritten discussion, but possible to examine specifically through the rest of the text.

Also, the paper still contains many ambiguities of method and much inference that isn't always supported by robust results. The main result appears to be an extremely large increase in temperature sensitivity when albedo is allowed to vary. While one would expect the sensitivity to increase, the magnitude of the increase here needs to be better supported by the validation of the albedo scheme against measurements and further discussion or analysis around the role of impurities to justify the low minimum value for snow albedo used. The choice of minimum values of albedo for snow and ice also impacts the conclusion that winter balance is of less importance, as with higher minimum albedo values the contrast between snow and ice is larger and this will increase the sensitivity. The authors also need to comment on the processes driving this sensitivity - to what extent the sensitivity is driven by decreases in snow albedo, earlier transition to ice and summer snowfall.

The temperature sensitivity with albedo feedbacks is still high, doubling the response, but it is much reduced with the various model changes that went into the revisions. Now it is well in line with previous literature. We made numerous changes, but mostly we now introduce all parameterizations for consistency, in particular cloud feedbacks (via the clearness index), such that shortwave radiation decreases with increased humidity (more cloud), offsetting the increase in longwave radiation. This and the increase in minimum snow albedo to 0.4 have reduced the sensitivity. With the revised experiments, the only difference is temperature forcing with and without albedo feedbacks, i.e., the other temperature feedbacks such as length of the melt season, rain vs. snow, are the same in all temperature experiments (cases 1-4 in Table 4).

The use of parameterisation for the radiative fluxes that responds to various drivers is encouraging. There is still a need for a validation of the melt produced by the fully parameterised model against that driven with measurements if the results are to be trusted. It is also ambiguous which version of the model is being used for any particular run and this needs to be carefully detailed and explained.

We had not brought in a discussion of the energy balance and subsurface model performance and validation vs. observations or a 'reference' model, mostly because these models are based on Marshall (2014) and Ebrahimi and Marshall (2015), for the longwave parameterization, and the current paper is already long. We could spend some time on a model validation section, but fear that this would be a diversion. The reviewer is correct though, confirmation that the parameterized daily model gives a good representation of the summer energy balance and melt is needed in order to trust the results. We added two paragraphs on this (Section 3, 11.407-423).

The new text gives a summary of the parameterized model performance in a sample dataset from summer, 2015, for which the melt model was not tuned or calibrated. We observed/measured a total summer melt of 3.1 ± 0.1 m w.e. at the AWS site from May to September, 2015, based on a May snowpit (measured snow water equivalent), the AWS SR50 record, and ablation stake data. The energy balance and subsurface snow temperature/ drainage model are run in two modes, forced by the 30-minute AWS data (best case or reference model) and with the parameterized model, that degrades the AWS data to daily forcing with parameterized albedo (Eq. 20), incoming longwave radiation (Eq. 8), and diurnal cycles of temperature and shortwave radiation from Eqs. (17) and (18). The reference model that is driven by 30-minute AWS data gives a total summer melt of 3.04 m w.e., and the parameterized daily model gives 2.98 m w.e., with a small underprediction of melt due to over-estimated summer albedo values (figure R1). The RMS error in daily melt totals is 3% (0.7 mm, relative to a mean summer melt value of 23 mm w.e.). Daily melt predictions from each model are shown here:

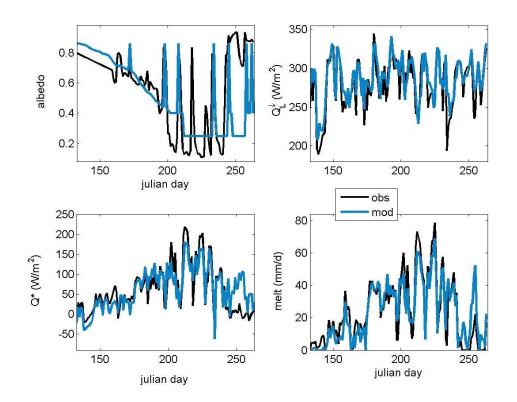


Figure R1. Parameterized model performance (blue) vs. the reference model (black), that is driven by 30-minute AWS data. There can be large departures in the melt daily melt rates (lower right), mostly attributable to discrepancies between the measured and modelled albedo (top left). The model overestimates albedo in summer 2015 (May-August) and underestimates it in September, when new snow (an early start of winter) is not adequately captured. Note that the downwardlooking shortwave radiometer was not properly wired in summer 2015 from May 12-June 6 (day 133-157), so albedo is estimated for this portion of the record. Bare-ice albedo at the AWS site was close to 0.1 in summer 2015, but is not always this dark.

We do not include this figure or other model calibration discussions in the main text, as we don't want to lengthen it or take a tangent, but we summarize these statistics on model performance to address the question of whether model skill is adequate for the sensitivity analyses that are our main focus. This is just one summer, but the model was tuned for the period 2002-2012 and the performance is similar here; it is not perfect, especially for the albedo, but we believe the parameterized model has reasonable skill and is a reliable tool for our present purpose.

The authors identify a series of significant feedbacks between air temperature, humidity and the incoming fluxes of short and longwave radiation. It would appear most of these are all connected through cloud cover, and it would be much clearer and insightful for the authors to explicitly examine changes in the frequency and attenuation of cloud cover alongside with air temperature and humidity, rather than inferring these from the sensitivity of incoming shortwave radiation in the analysis of the in-situ dataset. This should be possible as the authors already have a

parameterisation of cloud cover that includes humidity. If not, then the variability in incoming longwave also needs to be included in Table 4.

This is a good point and suggestion. We have rerun everything for the revised submission, with a simplified (identical) set of model parameterizations in all runs and with consistency of the shortwave and longwave radiation perturbations, through the transmissivity \tau. This is effectively our cloud proxy, and is now described as such. So we perturb \tau (\pm 0.1) and describe this as changes in cloud cover, which have opposite influences on incoming LW and SW radiation. Table 4 includes this now. Also, all experiments with a change in humidity see a corresponding change in \tau, hence incoming SW and LW. This is now internally consistent.

Moreover, all sensitivity studies in section 5 now use the parameterized version of the code, in the daily energy balance/subsurface temperature model. That is to say, we do not use the raw 30-minute AWS data to force the model, but rather the parameterized diurnal cycle, albedo model, modelled surface temperature, parameterized incoming longwave radiation, and parameterized clearness index (shortwave transmissivity), \tau. This allows clear understanding of our experiments and comparability across results. All of our numbers in Table 4 are revised as a result of these changes. Most have not changed much, but the temperature sensitivity with feedbacks is now much less – now more in line with observations and the NARR results.

In general, the paper is fairly well written though some of the text in the results and discussion is quite methodological and repetitive and perhaps is better suited to a methods section. Further work is needed to distill the main results from a rather large body of work and succinctly present them here.

Agreed, our results and discussion were repetitive and we did not do a good job of distilling the main results. We have rewritten and shortened this. We did keep some of the methodological details within the results, e.g., the details concerning the NARR forcing in section 6. The paper covers a lot, and we initially had much of this material in section 2 (methods), but it was far removed from the eventual results or NARR experiments and made for difficult flow/reading. The same holds true for the partial derivatives/theoretical sensitivities in Section 4. They are tiresome but are relevant locally (section 4) and not in the other results' sections. Hence we choose to combine methods and results to some degree in Sections 4-6, where specific to that section. We did our best to clean up the discussion, however.

The authors have managed to refocus their analysis but the new results and novelty of their results are often quite hidden. I would suggest re framing the discussion around 1. A summary of the important sensitivities and feedbacks that are observed on the Haig, 2. A discussion of the utility of the theoretical sensitivity based on mean summer conditions (good correspondence with full summers when feedbacks are omitted) 3. A discussion of the utility of exploring sensitivity to inter-annual variability with reanalysis datasets (mixed results). This structure would alert the reader to what is new and avoid some of the repetition.

The discussion and conclusion have been rewritten, somewhat but not fully along the lines suggested. We hope that the main results are now more clearly presented.

Line comments

204 - replace 'profile method' with 'bulk aerodynamic method'. The profile method uses wind speeds and temperature at two heights.

Revised, 1.187.

244 - the symbol psi has a different case from the equation.

Revised, 1.227.

249 - please include an explanation of how the refreezing rate is calculated and what constraints are put on the volumetric water content.

Added as per request of both reviewers, 11.234-238 and 11.248-251. The upper constraint on the volumetric water content is the porosity of the snow or firn, but drainage in the seasonal snow is efficient enough that this is never approached; tracking θ w evolution through the summer, it reaches a maximum of about 10%. There is a minimum (irreducible) water content in temperate snow, associated with capillary pressure, which we set to 3%.

278 - for consistency please state the model uses a variable timestep from 10 minutes to 1 hour to allow for stability of the subsurface temperature prognosis.

Revised as suggested, ll. 271-272.

349 - the minimum value for snow is very low. For the same site, Marshall (2014) gives 0.4 as a minimum. This will have a large impact on the sensitivity. Other authors have used values around 0.5 (Oerlemenas and Knap, 1998) and further justification of this low value is needed.

This is well observed, and R1 also questioned this. As discussed in the response to R1, this does of course impact the sensitivity; we now include experiments with both 0.3 and 0.4 as the minimum snow albedo (Figure 7), and have reverted to 0.4 as the default for the sensitivity studies (1.340). We initially used 0.3 because our observations at Haig Glacier indicate a value of 0.3 for firn, but old seasonal snow is closer to 0.4 at our site (see, e.g., the black line in Fig R1 above, just before the glacier ice is exposed). We sometimes see values below 0.4 for old snow at the Haig Glacier AWS site, in late summer, but 0.3 is probably too low to be representative of the accumulation area or the region, in general. It was our oversight to treat aged seasonal snow as firn; the latter has had more time to accumulate impurities.

405 - need to provide more discussion and evidence for the reasons for the decrease in albedo - i.e. is the increase in particulate concentrations documented?

The increase in particulate concentration is certainly observed, both in old snow and exposed firn and ice, but has never been quantified or supported by measurements. Empirically though, the seasonal albedo decrease at this site is strong, as has been documented elsewhere and as evident in e.g. Fig. R1 and Fig. 4a. Also see Fig. R2 below, for the average summer albedo evolution from 2002-2012. We assume that the seasonal decline is due to the water content, impurities, and grain growth in the temperate summer snowpack (after the conventional wisdom in Cuffey and Paterson and elsewhere), but it is fair to say that we are speculating as we don't have measurements or process studies to attribute the causes of the seasonal albedo decrease. We just observe a snow albedo decline from ~ 0.8 to ~ 0.4 each summer, before it drops to ~ 0.2 on the bare ice (Fig. R2). A detailed study of this may be warranted, but is out of scope here. For now, we modify the text (ll.394-398) to note the observations and empirical record but stop short of attributing cause.

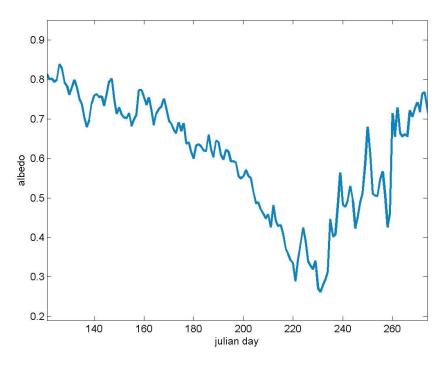


Figure R2. Average summer albedo evolution at the AWS site, 2002-2012.

419 - the contribution to melt should really be computed for melting periods only, as non-melting periods will bias these fractions towards the sensible heat flux (see Conway and Cullen, 2016). Please either show the fluxes for melting periods only or discuss only the contribution to the energy balance and not to melt.

Point taken, we now refer just to the contribution to the energy balance, 1.404.

423-430 - please make it clear that the feedbacks and NARR analyses are presented in the following two sections, rather than the current section.

Revised; the NARR discussion has been removed from here as it was evidently distracting.

434 and 439 - some more context is needed to justify conditions on the Haig being 'typical' of other mid-latitude glaciers. Please add either a table showing this or some references to papers with similar climatologies.

This is fair; we suspect but cannot substantiate that all of the weather conditions here are 'typical' of mid-latitude glaciers. For instance, the elevation (hence pressure and vapour pressure) are lower

than for glacier in the Alps and higher than the ablation zone of coastal mid-latitude glaciers; winds at other sites are often stronger than here; etc., etc. It is difficult to argue that mean conditions here are typical, and that is a loose qualitative word. In the end we removed this line and a bit of text around it from the manuscript, near 1.430.

444 - need further analysis in order to justify that the JJA has more impact on melt? Table 3 shows an almost identical combined sensitivity.

Yes, it is true in the results – this is why we report it, as we did not necessarily expect MJJAS sensitivity to be the same. But JJA conditions have more impact on melt simply because 80% of the melt occurs in these months. A cold May or Sept does not have the same impact on summer mass balance as a cold July. Net energy sensitivities may not change much, but the impact on melt does. In any case, this is now N/A as the MJJAS sensitivities are no longer reported, at the reviewer's request (see below).

442-448 - this is quite a confusing paragraph as the rationale for including/excluding months changes from the start to the end of the paragraph and results are introduced, but not referenced properly. Please either point to the figures/tables that justify these statements or move this text to the discussion.

Paragraph has been revised and simplified, 11.444-447, sorry for the confusion.

453 - Are these perturbations calculated as the average of positive and negative deviations from the mean? Some of the text (e.g. 518) seems to suggest that only positive perturbations were considered, which is not ideal.

This is a bit interesting. Yes, in Section 5, with anomalies introduced into the numerical model, perturbations are always introduced as positive and negative deviations from the mean. Here where we consider theoretical sensitivities, the values are based on derivatives at a point (the mean state), so sensitivities can be considered to apply only at this point, i.e. for infinitesimal negative or positive perturbations from the mean. The result is the same for either sign of perturbation, as it is essentially the slope at the point. If the relationship is nonlinear, it becomes invalid for large negative or positive perturbations.

503 - I am not sure at this stage it is appropriate to transfer the calculations of net energy to melt, as in reality not all periods will have melting conditions. Perhaps it is better to state the increase or decrease in the net energy available. Along with this I would remove the melt column from table 3.

We agree, of course, and transfer the net energy perturbation to melt with a cautionary note (ll.504-508), but have retained this in Table 3 and the discussion because melt rates give a more intuitive idea of the potential impact, or lack thereof. The values can also be compared across the different perturbations. Moreover, JJA mean Q_N translates well to JJA melt at the site; the reference value is 97 W/m², which gives 2.30 m w.e. melt if this is converted directly to melting. The reference JJA melt is 2.32 m w.e. (Table 2), so within 1%. 516 - please be consistent with the symbols used - the text uses vapour pressure while table 3 shows specific humidity.

Revised, specific humidity throughout now, for the perturbations

604 - Please be explicit this is top of atmosphere solar variability.

This subsection has been rewritten, hopefully clear now.

620 - For comparing the relative importance of each variable - it would be more useful to present the individual sensitivities relative to 1 standard deviation perturbations in Table 3. Agreed – we have added the individual sensitivities to 1- σ perturbations and written this into the results. This also facilitates comparisons with Table 4 (modelled sensitivities).

630 - The way these variables have been perturbed is not meaningful as they are physically unrealistic. For example - some of the standard deviation in vapour pressure will be due to increases in relative humidity, but you also increase incoming shortwave in this experiment - which as you noted earlier is likely to decrease with increased relative humidity. Thus, the experiment is contradictory. Please exclude these last two lines in Table 3.

This is a fair point, we agree that it is inconsistent for all variables to change in the same sense. This was meant only to explore joint variability of multiple weather variables, as occurs in reality, but as implemented this was not meaningful. These two lines have been removed. It is better to explore meteorologically-meaningful covariability through NARR (section 6) or another means.

677 - It is still ambiguous which variables are held constant at their measured values and which are parameterised for each run (in particular incoming longwave and surface temperature). Please provide a comprehensive table.

We have clarified this in our numerical experiments and in the text. We now use the parameterized daily model for all experiments that we present; LW radiation, albedo, and surface temperature are all parameterized/internally calculated. The introduction hopefully clarifies this, ll.627-634.

680 - why were changes in incoming longwave not examined?

We now perturb the clearness index \tau, which jointly impacts incoming SW and LW radiation; experiments on the radiation fluxes are not considered. This is consistent with our effort to perturb observable weather variables (e.g., T, v, qv, clouds), and allows a sensible (albeit parameterized) co-variation of the incoming radiation fluxes.

764 - this result is likely to be strongly dependent on the choice of the minimum albedo of snow, which in this case does not differ much from an ice albedo. Either the sensitivity of this result should be tested, or a more thorough justification made for the very low value chosen here. (0.3)

Agreed and revised, as discussed above in some depth. The default minimum albedo is now 0.4, and Fig. 7c includes an illustration and brief discussion of the sensitivity to this parameter.

776 - are the anomalies calculated with respect to the mean in-situ conditions? I suspect you took the daily anomalies of NARR from the NARR climatology, then applied these daily anomalies to the mean in-situ conditions - please clarify.

Yes, this is correct. Clarified, 1.751.

792 - it would be useful to see the standard deviation of relative humidity included here.

We don't actually use NARR relative humidity; it is derived from the specific humidity and temperature, for thermodynamic consistency. Hence any errors and variability in RH will flow from the NARR-derived T and qv. But for interest, the NARR RH is 65 pm 3% (sigma = 2.8%).

801 - this line needs more context to link it with the previous sentences.

Revised, 1.781.

970 - do the interannual anomalies in SWin and LWin from NARR correlate with the anomalies from the in-situ dataset? If not, it is hard to see how the NARR represents realistic interannual variability in these fluxes. This severely limits the inferences that can be made from model runs made with these anomalies.

We more or less agree – these are only weakly correlated, 0.52 for incoming SW radiation and 0.17 for incoming LW. Variables like temperature are much stronger (0.81). These are for only 11 years of mean JJA conditions, so the data is a bit limited. In summary, incoming SW is OK and incoming LW is very weak in the reanalysis. We note that NARR does not represent realistic interannual variability in these fluxes (1.765, 1.779). Noted again in the discussion.

1005 - this statement needs more justification. It would seem that the NARR based reconstructions performed satisfactorily in describing interannual variations in net energy flux, but that this is based on the accidental cancelation of errors in the radiative fluxes driving the model (Figures 5 and 6 from the original manuscript. The approach is worthy of further exploration, but a more thorough evaluation of the performance of the model, including biases and areas for improvement, is needed here.

Yes, agreed, this is still the case that further work is needed if one wishes to drive glacier mass balance reconstructions with NARR forcing. We deliberately stopped short of that here, and revised the discussion to stay within the sensitivity study and recommend further work, as per the reviewer's suggestion.

1413 - needs a legend describing the colours. Also, box 4 could be better as a separate figure, as the colours indicate the change in Qn due to different forcing, while in the other boxes the colours show the response of different fluxes to the same forcing.

Legend now added and the figure has been simplified to show the same fluxes in each case.

1415 - both black and red are listed as net radiation - please fix.

Revised.

1416 - add 'please note the different y scales'.

Revised as suggested.

1416 - please clarify in the caption which scenarios these figures relate to in Table 4.

Revised as suggested.

1277 - net melt has decreased 7-8% while net energy fluxes have remained similar. It would be useful to show the fraction of time the surface is diagnosed as melting in each month to provide some justification here.

We no longer discuss the M-S conditions, to shorten and focus the ms. Agree though that this is interesting.

1388 - I am not sure this figure adds much as you cannot see the detail in the daily values over the 11-year period, and the results presented in the figure are not discussed in the text. I would suggest either removing the figure, or modifying the figure so it is readable and discussing the results further. If the figure is kept the size of each box needs to be expanded and the line weight reduced to make a more readable figure.

Figure has been modified to better illustrate the data. It is meant to give a sense of the observed/driving data. It is now discussed a bit more in the main text, ll.383-387

1394 - please use thinner lines on these figures. Also, as months and not day of year are discussed in the text, it would be good to have months as the x-axis label, or at the very least, further tick marks that are at monthly or 30 day intervals.

On panel d, the median + interquartile range would better present the seasonal variation of melt rate. As it is, the mean appears to be greatly influenced by individual large melt events (such as around day 230).

Revised as suggested, also panel d (now median and interquartile range). It is true in the early spring or late fall, the mean is influenced by large events, though not so much in JJA.

1430 - the y axis of figure 7a should be m w.e. in line with the text and Figure 7c.

Revised as suggested

1469 - 1495 - were these figures meant to be included?

Our mistake, apologies. Remnants of the first submission.

Surface Energy Balance Sensitivity to Meteorological Variability on Haig Glacier, Canadian Rocky Mountains

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10 Abstract

Energy exchanges between the atmosphere and the glacier surface control the net energy available 11 for snow and ice melt. This paper explores the response of a mid-latitude glacier in the Canadian 12 Rocky Mountains to daily and interannual variations in the meteorological parameters that govern 13 14 the surface energy balance. We use an energy balance model to run sensitivity tests to perturbations in temperature, specific humidity, wind speed, incoming shortwave radiation, glacier surface 15 16 albedo, and winter snowpack depth. Variables are perturbed (i) in isolation, (ii) including internal 17 feedbacks, and (iii) with co-evolution of meteorological perturbations, derived from the North 18 American regional climate reanalysis (NARR) over the period 1979-2014. -Summer melt at this 19 site has the strongest sensitivity to interannual variations in temperature, albedo, and incoming shortwave radiation (i.e., cloud cover). Fluctuations in specific humidity also impact summer melt 20 extent, while interannual variability influctuations in cloud cover, wind speed, and winter 21 22 snowpack depth have less influence. Feedbacks to temperature forcing, in particular summer 23 albedo evolution, strongly amplifydouble the melt sensitivity to a temperature change. When meteorological perturbations co-vary through the NARR forcing, summer temperature anomalies 24 25 remain important in driving interannual summer energy balance and melt variability, but they are reduced in importance relative to an isolated temperature forcing. Covariation of other variables 26 27 (e.g., clear skies, giving reduced incoming longwave radiation) may be partially compensating for 28 the increase in temperature. -The methods introduced in this paper provide a framework that can 29 be adapted extended to compare the sensitivity of glaciers in different climate regimes, e.g., polar, 30 maritime, or tropical environments, and to assess the importance of different meteorological 31 parameters in different regions.

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34 1. Introduction

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Glaciers and icefields are thinning and retreating in all of the world's mountain regions in response to global climate change (e.g., Marzeion et al., 2014). This is reshaping alpine environments, affecting regional water resources, and contributing to global sea level rise (e.g., Radić and Hock, 2011). Melting of glaciers has drawn great attention, and it is important to understand the meteorological controls of snow and ice melt to correctly project glacier response to climate variability and change.

A glacier's climate sensitivity can be expressed in terms of the energy or mass balance response
to a change in meteorological conditions (Oerlemans and Fortuin, 1992; Oerlemans et al., 1998).
In a study of 12 glaciers byFor instance, Oerlemans et al. (1998);) defined the static or fixedgeometryglacier sensitivity to temperature, S_T, is expressed as:

$$S_T = \frac{\partial B_m}{\partial T} \approx \frac{B_m (+1K) - B_m (-1K)}{2}$$

46 where $B_m(\delta T)$ denotes the mean specific mass balance corresponding to the temperature 47 perturbation δT . Mass balance sensitivity to precipitation perturbations, $S_P = \partial B_m / \partial P$, iscan be 48 calculated in the same way.

49 Braithwaite and Raper (2002) extended the static sensitivity approach to regional scales, with the 50 idea that glaciers within a given climate regime should have similar mass balance sensitivities to 51 variations in temperature and precipitation. This framework has been used in numerous studies to 52 describe glacier sensitivity to climate change (e.g., Dyurgerov 2001; Klok and Oerlemans, 2004; 53 Arendt et al., 2009; Anderson et al., 2010; Engelhardt et al., 2015). Oerlemans et al. (1998) and 54 Oerlemans (2005) also introduced measures of dynamic sensitivity, characterizing glacier volume

55 and length changes as a function of changes in temperature.

56 Most studies to date have concentrated on glacier mass balance response to changes in temperature and precipitation. This is sensible, as these two fields are perhapsgenerally the most important 57 58 meteorological parametersvariables affecting glacier mass balance. Temperature and 59 precipitationThese two fields are also commonly measured, with long-term records available in 60 somemany regions, and extensive effort has gone into modelling and downscaling these two fields for a wide range of climate change impacts studies, including glacier modelling. Related to this, 61 62 Temperature and precipitation have also received the most attention because regional- to globalscale models of glacier mass balance commonly employ temperature-index methods to 63 parameterize glacier melt (e.g., Marzeion et al., 2014; Clarke et al., 2015). This is appealing 64 65 because temperature index models require only temperature and precipitation), with only these

66 <u>variables</u> as inputs.

Fremperature-While temperature index methodsmodels have been demonstrated to have reasonable skill in estimating seasonal melt (Ohmura, 2001; Hock, 2005), but they are nonetheless missing much of the physics that govern snow and ice melt. Because temperature-index models estimate snow and ice melt as a function of air temperature, these modelsAlso, they may be overly sensitive to changes in temperature, and may notwithout effectively <u>eapturecapturing</u> the impact of shifts in other <u>elimate</u> variables such as wind, humidity, or cloud cover. Internal processes and feedbacks that are important to glacier melt may also be absent, such as the surface albedo evolution that is Formatted: Font color: Auto

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- 74 observed on glaciers during the melt season, may also be absent, since degree-day melt factors
- 75 are usually taken to be static. Such feedbacks are critical to glacier melt (e.g., Brock et al., 2000;
- 76 Klok and Oerlemans, 2004; Cuffey and Paterson, 2010).
- 77 It is uncertain whether variability in glaciometeorological variables other than temperature and
- 78 precipitation is important to glacier energy and mass balance. While most large-scale glacier 79 change projections are rooted in temperature sensitivity (as built into temperature-index models), 80 it is generally recognized that the complete surface energy balance is important to glacier melt-and 81 its sensitivity to elimate change. For instance, net radiation has been identified as the main source 82 of melt energy for continental glaciers, accounting for ~70-80% of the total melt energy (e.g., Greuell and Smeets, 2001; Oerlemans and Klok, 2002; Klok et al., 2005; Giesen et al., 2008; 83 84 Marshall, 2014), with shortwave radiation providing the principal energy source. NetIncoming 85 shortwave radiation is not directly dependent on temperature. Nor are the turbulentAs another 86 example, latent heat fluxes, which can be are a significant source of energy in maritime and tropical environments (Wagnon et al., 1999, 2003; Favier et al., 2004; Anderson et al., 2010), and their 87 88 strength is a function of humidity and wind conditions, which are not strongly correlated with temperature fluctuations. This calls for a broader exploration of glacier sensitivity to climate 89 variability and change, beyond just the influence of temperature. 90
- 91 Energy balance models have been used extensively on individual glaciers (e.g., Arnold et al., 1996;
- 92 Klok and Oerlemans, 2002; Hock and Holmgren, 2005), and severalSeveral studies that estimate 93 glacier sensitivity to temperature change use complete models of surface energy balance (e.g.,
- 94 Klok and Oerlemans, 2004; Klok et al., 2005; Anslow et al., 2008; Anderson et al., 2010).
- 95 Assessments of glacier mass balance sensitivity have concentrated on changesThe influence of
- 96 <u>other meteorological variables has been explored</u> in temperature and precipitation, however, and
- 97 fluctuations in other variables are seldom considered a few studies. Gerbaux et al. (2005)
- exploreexamine the role of other meteorological different variables (e.g., temperature, moisture, wind) in energy balance processes and climate sensitivity of glaeiers in the French Alps. Giesen et
- al. (2008) note the importance of cloud cover in modulating interannual variability in summer melt
- 101 on Midtdalsbreen, Norway. Sicart et al. (2008) examine the surface energy budget on three glaciers
- 102 in different latitudes/climate regimes. Variations in net shortwave radiation, sensible heat flux, and
- 103 temperature each contribute to differences in glacier sensitivity to climate variability between these
- 104 three-locations.
- 105 We build on these studies through a systematic examination of glacier energy balance and melt 106 sensitivity to meteorological variability at Haig Glacier in the Canadian Rocky Mountains.. We 107 report the mean melt season (May to September, MJJAS) and summer (JJA) meteorological and 108 energy balance conditions on the Haig glacier in the Canadian Rocky Mountains for the period 109 2002-2012, based on an automatic weather station (AWS) in the upper ablation area (Marshall, 110 2014). These reference data are then-used as a baseline for theoretical and numerically modelled 111 sensitivity tests that assess the impact of changes in different meteorological parameters. The same 112 perturbation approach is then used to reconstruct variations in surface energy balance and melt for the period 1979-2014, based on North American regional climate reanalyses (NARR) (Mesinger 113 114 et al., 2006).

115 One of our Main questionsquestion is whether variables other than temperature and precipitation need to be considered to provide a realistic estimate of glacier sensitivity to climate 116 variability and change for mid-latitude mountain glaciers. Our analysis of summer energy and 117 mass balance sensitivity to meteorological variabilityin this study is limited to just one site in this 118 119 study, with a focus on the summer melt season (vs. annual mass balance). We examine the summer 120 energy balance in detail, however, to and evaluate the impact of different weather variables in 121 isolation and with more realistic covariance of meteorological conditions, and to control for direct 122 vs. indirect influences (i.e. feedbacks) on glacier melt at this site. The theoretical and energy-123 balance model framework for assessing glacier sensitivity that we introduce is applicable to other regions, and may help to understand regional differences in glacier sensitivity to climate variability 124 125 and change.....

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127 2. Surface Energy Balance and Melt Model

The energy budget at the glacier surface is defined by the fluxes of energy between the atmosphere, the snow/ice surface, and the underlying snow or ice. The surface energy balance can be written

$$Q_N = Q_S^{\downarrow}(1-\alpha) + Q_L^{\downarrow} - Q_L^{\uparrow} + Q_H + Q_E + Q_C, \qquad (2)$$

where Q_N is the net energy flux at the surface and $Q_S^{\downarrow}, Q_L^{\uparrow}, Q_L, Q_E$, and Q_C represent incoming shortwave radiation, incoming and outgoing longwave radiation, sensible and latent heat flux, and subsurface conductive energy flux, respectively. The energy fluxes have units of W m⁻². The surface albedo is denoted α and fluxes are defined to be positive when they are sources of energy to the glacier surface. This expression of the surface energy balance neglects We neglect the penetration of shortwave radiation and advection of energy by precipitation and meltwater fluxes.

140 The net energy Q_N can be positive or negative. When it is negative, as it is for much of the winter 141 and during the night, the snow or ice will cool or liquid water will refreeze. Positive net energy 142 will drive surface and subsurface-warming, or on a melting glacier surface with $Q_N > 0$, the net 143 energy flux is dedicated to generating surface melt, with For melt rate *m*, followingthis follows 144

$$\dot{m} = \frac{Q_N}{\rho_w L_f},\tag{3}$$

where ρ_w is the density of water and L_f is the latent heat of fusion. Melt rates in Eq. (3) have units of metres water equivalent per second (m w.e. s⁻¹).

Numerous studies have shown that incoming shortwave radiation is the dominant term in the energy balance during the melt season in most glacial environments. Incoming shortwave radiation (insolation) at the surface has three components: direct and diffuse solar radiation, along with direct solar radiation that is reflected from the surrounding terrain. Direct solar radiation is the radiative flux from the direct solar beam, which comes in at a zenith angle *Z*. It is a function of latitude, time of year, and time of day (e.g., Oke, 1987). Potential direct (clear-sky) incoming solar radiation on a horizontal surface can be estimated from

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$$Q_{\phi}^{\downarrow} = Q_0 \cos(Z) \varphi_0^{P/P_0 \cos(Z)}, \qquad (4)$$

for top-of-atmosphere insolation Q_0 , clear-sky atmospheric transmissivity φ_0 , air pressure P, and sea-level air pressure P_0 (Oke, 1987). Eq. (4) allows potential direct shortwave radiation to be calculated as a function of the day, year, latitude and elevation.

164 Longwave radiation can be estimated from the Stefan-Boltzmann equation,

$$Q_L = \varepsilon \sigma T^4, \tag{5}$$

where ε is the thermal emissivity, σ is the Stefan--Boltzmann constant, and *T* is the absolute temperature of the emitting surface. Snow and ice emit as near-perfect blackbodies at infrared wavelengths, with surface emissivity $\varepsilon_s = 0.98-1.0$. The longwave fluxes are then

$$Q_L^{\uparrow} = \varepsilon_{\rm s} \sigma T_s^4, \tag{6}$$

$$Q_L^{\downarrow} = \varepsilon_a \sigma T_a^4, \tag{7}$$

for surface temperature T_s , near-surface air temperature T_a , and atmospheric emissivity ε_a . Terrain emissions (i.e. from the surrounding topography) can also contribute to the incoming longwave radiation, particularly at sites that are adjacent to valley walls.

A spectrally- and vertically-integrated radiative transfer calculation is needed to predict the 180 incoming longwave radiation from the atmosphere, as this depends on lower-troposphere water 181 vapour, cloud, and temperature profiles. Because the requisite atmospheric data are rarely available 182 in glacial environments, Q_L^{\downarrow} is commonly parameterized at a site as a function of local (2-m) 183 temperature and humidity. Where available, cloud cover or a proxy for cloud conditions, such as 184 the atmospheric clearness index, are often used to strengthen this parameterization. Hock (2005) 185 and Lhomme et al. (2007) provide reviews of some of the parameterizations of atmospheric 186 emissivity that have been employed in glaciology. We found good results for regression-based 187 parameterization at two study sites in the Canadian Rocky Mountains (Ebrahimi and Marshall, 188 189 2015),

$$Q_L^{\downarrow} = (a + be_v + ch) \,\sigma T_a^4 \tag{8}$$

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$$Q_L^{\downarrow} = (a + be_v + c\tau) \, \sigma T_a^4,$$

Here *a*, *b*, and *c* are regression parameters (different in Eqs. (8) and (9)), e_v is vapour pressure, *h* is relative humidity, and τ is the clearness index, calculated from the ratio of measured to potential direct incoming shortwave radiation.

Solar radiation and cloud data are less commonly available than relative humidity, so Eq. (8) is a slightly less accurate but more portable version of this parameterization (Ebrahimi and Marshall, 201 2015). Multiple regressions of ε_a containing both relative humidity and clearness index were rejected, as these are highly (negatively) correlated. All-sky longwave parameterizations using either of these variables are reasonable, with root-mean square errors in mean daily incoming
 longwave radiation of about 10 W/m².

Relative humidity can also be used as a proxy for clearness index if shortwave radiation data arenot available. Summer (JJA) observations at Haig Glacier follow the relation:

$$\tau = 1.3 - 0.01h \,, \tag{10}$$

for mean daily values of τ and h ($R^2 = 0.5$). We draw on this below when we need to estimate perturbations in sky clearness index that are consistent with changes in atmospheric humidity. In accord with the observational basis of Eq. (10), the clearness index is constrained to be within 0.3 and 1 ($h \in [30, 100\%]$); if daily mean humidity drops below this, we set $\tau = 1$.

Turbulent fluxes of sensible and latent energy in the glacier boundary layer are commonly parameterized from an eddy-diffusivity model of turbulent exchangea bulk aerodynamic method (e.g., Andreas, 2002), also known as the profile method:):

$$Q_{H} = \rho_{a} c_{p} k^{2} v \left[\frac{T_{a}(z) - T_{s}}{\ln(Z/z_{0}) \ln(Z/z_{0H})} \right], \qquad (11)$$

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$$Q_E = \rho_a L_v k^2 v \left[\frac{q_a(z) - q_s}{\ln(z/z_0) \ln(z/z_{0E})} \right].$$
(12)

Here ρ_a is the air density, c_p is the specific heat capacity of air, L_v is the latent heat of evaporation, 225 k = 0.4 is von Karman's constant, v is wind speed, and q refers to the specific humidity. 226 Measurements of temperature and humidity are assumed to be at two levels, height z (e.g., 2 m) 227 and at the surface-air interface, s. For a melting glacier surface, $T_s = 0$ °C, and q_s can be taken from 228 229 the saturation specific humidity over ice at temperature T_s . We estimate T_s from an inversion of Eq. (6), using measurements of outgoing longwave radiation. In sensitivity tests, where we depart 230 from the observational constraints, T_s is internally modelled within a subsurface snow model (see 231 below), taken from the temperature of the upper snow layer. 232 233

234 Parameters z_0 , z_{0H} , and z_{0E} refer to the roughness length scales for turbulent exchange of momentum, heat, and moisture. We adopt fixed values for each, equivalent for both snow and ice 235 236 $(z_0 = 43 \text{ mm}; z_{0H} = z_{0E} = z_0/100)$, based on closure of the surface energy balance with reference to 237 observed melt (Marshall, 2014). Atmospheric stability adjustments can be introduced in Eqs. (11) and (12) to modify the turbulent flux parameterizations for the stable glacier boundary layer (e.g., 238 239 Hock and Holmgren, 2005; Giesen et al., 2008). In this study we We do not apply stability 240 corrections. Marshall (2014) was, as we are able to attain closure in modelled and measured 241 summer melt at this site without including stability corrections, and othersthis. Others have argued that stability corrections may lead to an underestimation of the turbulent fluxes on mountain 242 glaciers (e.g. Hock and Holmgren, 2005). This may be related to the low-level wind speed 243 maximum that is typical of the glacier boundary layer, which introduces strong turbulence and is 244 245 not consistent with the logarithmic profile of wind speed that is implicit in Eqs. (11) and (12). It 246 may also be that the effects of atmospheric stability are absorbed in the roughness values -

roughness values that are adopted to attain closure in the surface energy balance and melt calculations may be too low, implicitly accounting for the stable boundary layer.

Subsurface temperatures are modelled through a multi-layer, one-dimensional model of heat conduction and meltwater percolation and refreezing in the upper 10 m of the glacier. <u>Details of</u> the model are given in the next section. Ten meters is, the approximate depth of penetration of the annual temperature wave (Cuffey and Paterson, 2010). This depth includes the time-varying seasonal snow layer and the underlying firn or ice. The temperature solution follows

$$\rho_s c_s \frac{\partial T}{\partial t} = \frac{\partial}{\partial z} \left(-k_t \frac{\partial T}{\partial z} \right) + \varphi_t, \tag{13}$$

where ρ_s , c_s , and k_t are the density, heat capacity, and thermal conductivity of the subsurface snow, firn, or ice and $\frac{\psi_t(z)\varphi_t}{\psi_t(z)}$ is a local source term that accounts for latent heat of refreezing,

$$\varphi_t = \rho_w L_f \dot{r} / \Delta z \,. \tag{14}$$

163 In Eq. (14), the The refreezing rate \dot{r} has units m s⁻¹, φ_t has units W m⁻³, and Δz is the thickness 164 of the layer in which the meltwater refreezes.

266 We useRefreezing is calculated from a simplistic hydrological model that is coupled with the 267 subsurface thermal model. We track the volumetric liquid water fraction, θ_{w_a} in the snow/firn pore 268 space, and if conductive energy loss occurs in a subsurface layer where liquid water is present, this 269 energy is diverted to latent enthalpy of freezing, rather than cooling the snow. Temperatures cannot 270 drop below 0°C until $\theta_w = 0$. Liquid water is converted to ice in the subsurface layer. 271 We model meltwater drainage; by assuming that water percolates uniformly_a with hydraulic

272 We model meltwater drainage; by assuming that water percolates uniformly, with hydraulic 273 conductivity k_h and neglecting horizontal transport (i.e. assuming only gravity-driven vertical 274 drainage). For a volumetric liquid water fraction θ_w in the snow/firm pore space, localLocal water 275 layer thickness can be expressed $h_w = \theta_w \Delta z$. The local water balance is then

$$\frac{\partial h_w}{\partial t} = -k_h \frac{\partial h_w}{\partial z} - \dot{r} , \qquad (15)$$

279 where the final term accounts for water that is generated or removed through internal melting or 280 refreezing. In principle, this is a source/sink term that could also include internal melting (e.g., 281 from shortwave radiation penetration or percolation of warm rainwater), but we do not consider 282 these processes. We assume an irreducible water content of 3% for the melting snowpack 283 (Colbeck, 1974), and the maximum volumetric water content is equal to the porosity, θ , although 284 drainage in the seasonal snowpack is efficient and θ_w is always much less than θ .

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287 Numerical Energy Balance and Subsurface Temperature Model

For the energy balance sensitivity experiments in this study, we use a combination of directly observed and modelled glaciometeorological variables. Where we report the directly observed surface energy balance, for the 2002-2012 reference state, we drive the energy balance model with observed 30-minute data, including measured albedo and longwave radiation fluxes. Turbulent
 heat fluxes and subsurface heat conduction are modelled from Equations (11-15).

Where we do sensitivity tests or run the model with other meteorological input, such as from 294 climate models, we need to allow for internal feedbacks such as freely-determined albedo 295 evolution and changes in incoming radiation that will attend changes in atmospheric conditions 296 297 (e.g., cloud cover, humidity). The energy balance and melt model that we employ is based on daily mean meteorological inputs, in order to make our approach compatible with output from climate 298 299 models or reanalyses, as well as parameterizations that operate on a daily timescale (Eqs. 8-10). A parameterized diurnal cycle is introduced for temperature and shortwave radiation (see below), in 300 order to capture the effects of overnight refreezing and the fraction of the day that experiences melt 301 302 (when Q_N and $T_s > 0$). The model is run year-round with uses a nominal variable time step of from 303 10 minutes to 1 hour to allow for stability of the subsurface temperature prognosis.

305 The subsurface temperature model has 33 layers, with 10-cm layers until 0.6-m depth, 20-cm

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layers from 0.6-2 m, and 40-cm layers from 2-10 m. The upper boundary forcing comes from the 306 307 surface energy balance, which dictates either the melt rate or the temperature change in the upper surface layer. Heatconductive heat flux at the snow/ice-air interface, $Q_C = -k_t \partial T / \partial z$, is modelled 308 from a three-point forward finite-difference approximation of $\partial T/\partial z$. We use a two-step solution, 309 310 for the temperature (Eq. 13), then the meltwater drainage (Eq. 15). The temperature solution is 311 implicit for the temperature diffusion, with latent heat release from refreezing (the source term in 312 Eq. 13) calculated from the previous time step₇ within the hydrological model. Hydraulic conductivity in Eq. (15) is assigned the value $k_h = 10^{-4} \text{ m s}^{-1}$, near the low end of estimates reported 313 by Campbell et al. (2006), and meltwater). Meltwater is assumed to drain instantaneously when it 314 reaches the snow-ice interface. 315

The 10-m subsurface model consists of the seasonal snowpack of thickness $d_s(t)$, overlying either firn or ice. The grid is fixed with respect to the surface, and each layer is assigned a density, thermal conductivity, and heat capacity according to the medium (snow, firn, or ice). Snow/firn depth is updated daily, based on daily melt totals, snowfall events, and densification through the summer melt season, May to September (see below). Snow and firn density are modelled as a function of depth and the liquid water and ice content,

$$\rho_s = \rho_i (1 - \theta) + \rho_w \theta_w + \rho_i \theta_i , \qquad (16)$$

for porosity θ , liquid water fraction θ_w , and ice fraction θ_i . Densities ρ_s , ρ_i , and ρ_w refer to snow, ice and water, respectively. Porosity θ decreases We prescribe a decrease in porosity with depth following $\theta(z) = 0.6 - 0.05z$, parameterized to roughly represent the measured summer snow densities at the site ($\rho_s = 350-550 \text{ kg m}^{-3}$) and give reasonable estimates of firn density, up to $\rho_s =$ 820 kg m⁻³ at 10-m depth.

332 Snow accumulates, melts, or undergoes densification on a daily time step, with snow thickness d

333 varying continuously (vs. discretely) within the fixed-grid framework. At depth *d* below the

surface, the grid cell has a weighted combination of thermal properties and densities to reflect the

335 mixture of snow and either firn or ice in that layer. We do not have a model for snow

accumulation through the winter months. We treat this simply, and linearly accumulate snow

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from the start of winter until the start of the following melt season, with the accumulation rate set to give a match to the observed May snowpack <u>thickness</u> for each year. These data are available through annual winter mass balance surveys on the glacier, including a snowpit that provides

340 depth and density measurements at the AWS site.

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342 The steps in the energy balance and melt model are as follows:

1. Daily mean values are input for temperature, incoming shortwave and longwave radiation, airpressure, specific humidity, and wind speed, as well as minimum and maximum temperature.

2. A diurnal temperature cycle is parameterized as a cosine wave with a lag $\tau_t = 4$ hours to give the maximum temperature at 16:00, as per local observations, with an amplitude $A_t = (T_{max} - T_{min})/2$

348 (Figure 1a). For time t (hour of the day) and period $P_t = 24$ hours,

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$$T(t) = -A_t \cos\left[\frac{2\pi(t-\tau_t)}{P_t}\right].$$
 (17)

350 3. A diurnal cycle for incoming shortwave radiation is parameterized as a half-cosine wave with a period $P_{sw}(d) = 2h_s(d)$, where d is the day of year and h_s is the number of hours of sunlight on day d (Figure 1b). Defining lag τ_{sw} and amplitude A_{sw} ,

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$$Q_s^{\downarrow}(t) = \max\left\{-A_{sw}\cos\left[\frac{2\pi(t-\tau_{sw})}{P_{sw}}\right], 0\right\}.$$
 (18)

Sunlight hours are calculated as a function of latitude, θ , and day of year, based on the equation for the sunset hour h_{ss} (e.g., Liou, 2002):

$$\cos(h_{ss}) = -\tan(\delta)\tan(\theta), \tag{19}$$

where δ is the solar declination angle (solar latitude as a function of day of year). Sunlight hours 357 358 $h_s = 2h_{ss}$. The lag also varies with the day of year, and is calculated by setting peak shortwave radiation to occur at noon: $2\pi (12 - \tau_{sw})/P_{sw} = \pi$. This gives $\tau_{sw} = 12 - h_s$ hours. Amplitude A_{sw} is 359 calculated by integrating the area under the cosine curve and equating this to the average daily 360 incoming shortwave radiation, Q_{Sd}^{\downarrow} . This gives $A_{sw} = 12\pi Q_{Sd}^{\downarrow}/h_s$ W m⁻². This treatment implicitly 361 includes cloud effects that reduce incoming shortwave radiation on a given day (via Q_{Sd}^{\downarrow}), but 362 distributed evenly through the day. This neglects any systematic tendency for afternoon vs 363 364 morning clouds. For simplicity, we also neglect the effect of zenith angle on atmospheric 365 transmittance (i.e., lower transmittance for larger atmospheric path lengths in the morning and late 366 afternoon), although this could be built into a more refined model.

367 4. We assume that wind, incoming longwave radiation, air pressure, and specific humidity are

368 <u>constant through the day, held to the mean daily value. For sensitivity tests, Q_L^{\downarrow} is calculated</u>

following Eq. (8) and the daily mean value of Q_S^{\downarrow} is perturbed from Eq. (10) and $dQ_S^{\downarrow} = d\tau$.

- 370 Relative humidity has a diurnal cycle following temperature. This impacts incoming radiation
- 371 where we parameterize the sky clearness index from near-surface conditions (Eq. 10).

5. <u>5.</u> Relative humidity has a diurnal cycle following temperature, assuming constant daily
 humidity but adjusting *h* for consistency with the effect of temperature on saturation vapour
 pressure.

We assume that wind, incoming longwave radiation, air pressure, and specific humidity are constant through the day, held to the mean daily value. For sensitivity tests, Q_L^{\downarrow} is calculated following Eq. (11) and the daily mean value of Q_S^{\downarrow} is perturbed from Eq. (10) and $dQ_S^{\downarrow} = d\tau$.

6. Albedo is also modelled on a daily basis for the sensitivity studies. When the seasonal snowpack is melted away, albedo is set to the observed bare-ice value at the site, $\alpha_i = 0.25$. For fresh or dry snow, a fixed value $\alpha_0 = 0.86$ is used. The snowpack <u>thickness</u> is initialized on May 1 of each year, set to the observed value measured during the annual winter mass balance survey. During the melt season, which is assumed to start after this date, seasonal snow albedo decreases as a function of

(20)

cumulative positive degree days (ΣPDD) following Hirose and Marshall (2013),

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$$\alpha_s(d) = \alpha_0 - k_\alpha \sum PDD(d).$$

385 A minimum value of 0.34 is set for old snow-and firn, based on local observations. We also. We parameterize the effects of summer snow fall on albedo and mass balance through a 386 387 simplestochastic model of summer precipitation events, as described in (Marshall (, 2014). Precipitation events are set to occur randomly, with 25 events occurring from May through 388 389 September as the default setting. Precipitation totals vary randomly, between 1 and 10 mm in these 390 events, w.e., with snow at temperatures below 0°C, rainfall above 2°C, and rain/snow partitioning 391 increasing linearly over the range 0-2°C. Following a summer snow event, surface albedo is reset 392 to α_0 , and remains at this value until melting has removed the new snow its albedo begins to decay 393 following Eq. (20). This treatment allows a natural transition to end-of-summer conditions, when 394 fresh snowfall in September or October does not melt away.

7. Subsurface temperatures and the conductive heat flux, Q_c , are modelled with 10-minute to onehour time steps (chosen for stability of the temperature solution), driven by the energy balance (Eq. 2) at the upper surface. This gives surface and internal melt totals at each time step.). The updated surface temperature T_s is used for the calculation of outgoing longwave radiation (Eq. 6), sensible heat flux (Eq. 11), and latent heat flux (via q_s in Eq. 12) for the next time step.

8. The hydrology model is invoked to calculatecalculates meltwater drainage and refreezing. Annual meltwater runoff is then the sum of all meltwater that drains, while summer mass balance is equal to the meltwater runoff minus the total summer snowfall, nominally for the period May 1 to September 30 at this site. This allows for some meltwater retention as either liquid water or refrozen ice within the snow or firn. We neglect water storage in the englacial and subglacial hydrology systems.

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407 **3. Field Site and Observational Data**

408 Reference meteorological conditions, surface energy balance fluxes, and snow conditions are 409 based on *in situ* measurements at Haig Glacier in the Canadian Rocky Mountains for the period 410 2002-2012 (Marshall, 2014). Winter mass balance measurements are carried out each May. These observations provide an 11-year record of observed snow depth and summer melt from an
automatic weather station (AWS) located near the median elevation of the glacier, 2660 m (Figure
2). This is the upper ablation area of the glacier, which generally undergoes a transition from
seasonal snow to exposed glacier ice in August.

Table 1 summarizes the mean observed meteorological and conditions at Haig Glacier over the 11-year reference period. Data coverage is incomplete, particularly in the winter months, as we transitioned to summer only measurements (May-Sept) after 2009. For the 11 years, data coverage is as follows for most sensors (e.g., temperature, shortwave radiation): JJA - 90% (909 of 1012 days); MJJAS - 86% (1441 of 1683 days); annual - 63% (2519 of 4018 days). There are more missing longwave radiation data, as the sensor was not installed until July 2003. The corresponding numbers are: JJA - 76%; MJJAS - 70%; annual - 46%.

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423 The missing Missing data are gap-filled from a weather station that has operated continuously in the glacier forefield since 2001, at an elevation of 2325 m. The forefield AWS has more complete 424 425 data coverage than the glacier AWS, above 90% for all variables. Observational data are used to adjust for the altitudinal and environmental differences between the sites, through either a monthly 426 offset (e.g., $T_G = T_{FF} - \Delta T$), or a scaling factor β (e.g., $v_G = \beta_v v_{FF}$). Here, subscripts G and FF refer 427 to the glacier and forefield AWS sites. The monthly factors are calculated from the set of all 428 available overlapping data for the two stations. The temperature offset approach is equivalent to a 429 lapse rate, or can be expressed that way for distributed modelling over the glacier. In this study we 430 431 consider only the point energy balance at the glacier AWS site. Where If both stations are missing 432 data, gap-filling is done through assignment of mean daily observational data, in order to give 433 100% coverage.

To give a sense of the complete data record, Figure 3 shows examples of the full record, for air temperature (blue line in Fig. 3a), modelled surface temperature, and the radiationenergy fluxes (orange. Average June to August (JJA) air and red lines in Fig. 3b). Surfacesurface temperature
are 5.2°C and other surface -0.6°C, respectively, and 98% of JJA days reach surface temperatures of 0°C (melting conditions) in the 11-year record. The surface energy fluxes in Fig. 3 are modelled. The mean annual cycles of the energy fluxes3b illustrate the dominance of net radiation in governing net energy at this site (Table 2).

442 Mean daily values for the 11-year record are plotted in Figure 4. Mean daily values are plotted for 443 the 11-year record. The four components of the shortwave and longwave radiation in Figs. 4a and 444 4b are combined to give the net radiation (red line in Fig. 4e). As is typical for mid-latitude glaciers, 445 this net radiation is the main energy flux that drives glacier melt at this site (Fig. 4c).- Net radiation is negative in the winter, when shortwave inputs are low, albedo is high, and longwave cooling 446 447 gives a radiation deficit. During Net radiation is positive in the summer and increases through the 448 melt season, incoming. This is driven by increases in net shortwave radiation increases and as 449 snow albedo declines as snow becomes wet, particulate concentration increases, and the seasonal 450 snow at the site and then melts away to expose the underlying glacier ice (Fig. 4a). There is still a 451 net longwave deficit (Fig. 4b), but outgoing longwave radiation is limited (i.e. saturates at ~315 452 Wm^{-2}) when the surface is at the melting point. In combination, this gives surplus radiation, in

453 particular when low-albedo glacier ice becomes exposed4a). Measurements at the AWS site in late

454 July or August.indicate a seasonal snow albedo decrease from about 0.8 to about 0.4 each summer,

455 which may be due to a combination of increased snow water content, grain metamorphosis in the

456 temperate snowpack, and increasing concentration of impurities through the melt season (e.g.,

457 <u>Cuffey and Paterson, 2010).</u>458

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459 Table 2 summarizes the monthly surface energy balance fluxes at the Haig Glacier AWS site over 460 the melt season, May through September. Mean daily melt rates are plotted in Fig. 4d, as well as 461 the standard deviation, to give a sense of interannual variability over the period 2002-2012. Median 462 daily melt rates for the period 2002-2012 are plotted in Fig. 4d, along with the interquartile range. On average, 65% of the annual glacier melt occurs in the months of July and August. Net energy 463 peaks in August, when the low-albedo glacier ice is exposed. Sensible heat flux peaks in July, and 464 465 is the other main source of energy contributing to glacier melt. On average for June to August 466 (JJA), net radiation and sensible heat flux constitute 70% and 30% of the net energy-that is 467 available for melt, respectively. Latent heat flux represents a small sink of energy, and conductive 468 heat flux is a minor source of energy.

The energy balance and snowpack models have been developed and tested elsewhere (Marshall, 2014; Ebrahimi and Marshall, 2015), so we do not present the model validation in detail here.
Comparisons are favorable between AWS observations (*e.g.*, in situ albedo, SR50-inferred melt), the model driven with 30-minute AWS data, and the 'daily' version of the model used here, which includes parameterizations of albedo, incoming longwave radiation, and the diurnal temperature and shortwave radiation cycles (Section 2). The simplified daily model loses some reality, but its overall performance is excellent.

478 As an example, glacier AWS data from summer 2015 is used as an independent test of the model, 479 with its default parameterizations. Observed melt at the AWS site was 3.1 ± 0.1 m w.e. in summer 480 2015, while the melt model forced by 30-minute AWS data gives 3.04 m w.e. and the parameterized, daily version of the model gives 2.98 m w.e. Taking the 30-minute AWS-driven 481 results as the reference, the RMS error in the daily melt predictions for the parameterized model 482 483 is 3% (0.7 mm w.e., relative to a daily mean value of 22.7 mm w.e.). Departures from the 484 observations are primarily associated with the albedo, which is over-estimated in summer 2015. 485 Overall the parameterized daily model has good skill and is an appropriate tool for the sensitivity 486 analyses presented here. 487

488 4. Theoretical Sensitivity of the Surface Energy Balance

The meteorological variables in Tables 1 and 2 can be perturbed one at a time or in combination to examine the impact on modelled summer melt at the AWS site. We do this for the historical record (2002-2012) and also for the 35-year period 1979-2014, based on meteorological reconstructions from the North American Regional Reanalysis (NARR; Mesinger et al., 2006). The latter provides a more complete picture of interannual variability. Comparison of NARR predictions with measurements over the period 2002-2012 also allows us to assess the skill with Formatted: Font color: Auto

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496 which fluctuations in surface energy balance and summer melt can be captured in an atmospheric 497 model that does not explicitly resolve the alpine and glacier conditions.

498 Surface energy balance processes and summer melt rates depend on various meteorological-499 influences in (Eqs. (4-11). Warm summers generally cause high melt rates and promote negative 500 mass balance, but the energy balance is sensitive to other weather conditions as well. WeTo 501 examine these sensitivities for atmospheric conditions that are typical of, meteorological variables 502 in Tables 1 and 2 can be perturbed one at a time or in combination to examine the impact on 503 summer melt season on mid-latitude glaciers. For quantitative illustration, we adoptat the average 504 June to August (Haig Glacier AWS site. Perturbations are introduced with respect to the mean 505 JJA) meteorological conditions from 2002-2012 at Haig Glacier in the Canadian Rocky Mountains (Tables 1 and 2): $T_a = 5.1^{\circ}$ C, h = 67%, $e_v = 5.7$ hPa, $q_v = 4.8$ g/kg, P = 750 hPa, v = 2.6 m/s, $Q_S^{\downarrow} = 750$ hPa, Q_S^{\downarrow 506 226 W/m², $Q_{\psi}^{\downarrow} = 359$ W/m², $\tau = 0.63$, $\alpha = 0.55$, and $Q_{L}^{\downarrow} = 280$ W/m². Mean weather conditions 507 and surface energy balance conditions in Tables 1 and 2 are typical of mid-latitude mountain 508 509 glaciers.

510 511 Average JJA melt at Theoretical sensitivities are calculated in this section by differentiating the 512 513 Haig Glacier AWS site was 2320 mm w.e.net energy balance with respect to each meteorological variable. This is akin to generating a Jacobian matrix for Q_N , based on partial derivatives of the 514 515 dependent variables in the surface energy balance. One cannot gauge the most important meteorological influence on surface energy and mass balance from 2002-2012. the sensitivities to 516 a unit change in each variable. For instance, a change in specific humidity of 1 g kg⁻¹ equals 3.3 517 standard deviations, with respect to the interannual (JJA) variability (Table 1). In contrast, summer 518 temperature has a standard deviation of 0.8°C, so a 1°C temperature change is a smaller 519 perturbation. To allow a direct comparison of the theoretical sensitivities and to give a simple 520 representation of their natural, interannual variability, we perturb each variable by one standard 521 deviation, based on the values reported in Tables 1 and 2.

523 We consider the maincore summer months-here-, JJA, to calculate the theoretical sensitivity-524 because more the glacier surface is at melting point for most of this time (Fig. 3a), which is a 525 necessary condition to relate net energy to melt. More than 80% of the annual melt also occurs in 526 this season; (Table 2 and Fig. 4d), so meteorological forcing over this period has the highest impact 527 on glacier melt. Weather conditions also matter in the shoulder months, May and September, but 528 anomalies in these months have less impact on glacier melt and mass balance. We repeat the sensitivity analysis for MJJAS conditions, and present a summary of these results. Melt model 529 experiments in the next section consider the energy balance from May through September, in order 530 to capture the complete melt season. 531

533 Sensitivity to Temperature

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Air temperature appears directly in the expressions for Q_L^{\downarrow} and Q_H . Temperature change may also 535 influence the surface energy balance through influences on other variables, such as atmospheric 536 537

moisture (Q_E). For a melting glacier surface, where surface and subsurface temperatures are at 0°C, air temperature changes do not directly influence Q_L^{\uparrow} or Q_C . To estimate the magnitude of 538 539

temperature sensitivity, we differentiate each energy balance flux with respect to temperature.

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541 For incoming longwave radiation, Eq. (7), the resulting temperature sensitivity is:

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$$\frac{\partial Q_L^1}{\partial T} = 4\sigma \varepsilon_a T_a^3 + \sigma T_a^4 \frac{\partial \varepsilon_a}{\partial T}.$$
(21)

This general form applies to a range of formulations for ε_{a} , such as those of Brutsaert (1975), Lhomme et al. (2007), or Sedlar and Hock (2009). Adopting the parameterization in Eq. (8), which performs well at Haig Glacier,

$$\frac{\partial Q_L^{\ \flat}}{\partial T} = 4\sigma\varepsilon_a T_a^3 + \sigma T_a^4 \left(b \frac{\partial e_v}{\partial T} + c \frac{\partial h}{\partial T} \right).$$
(22)

The last two terms reflect potential feedbacks of temperature change on humidity. While we are only considering perturbations to temperature in this section, vapour pressure and relative humidity cannot both remain constant under a temperature change. We first assume that relative humidity *h* remains constant, under which conditions we assume that cloud cover and sky clearness will be unchanged. For constant *h*, e_v scales with temperature following the Clausius-Clapeyon relation for saturation vapour pressure, 557

$$\frac{\partial e_{\nu}}{\partial T} = \frac{h}{100} \frac{\partial e_s}{\partial T} = \frac{h}{100} \left(\frac{L_{\nu} e_s}{R_{\nu} T_a^2}\right) = \frac{L_{\nu} e_{\nu}}{R_{\nu} T_a^2},$$
(23)

560 where $R_v = 461.5 \text{ J kg}^{-1} \circ \text{C}^{-1}$ is the gas law constant for water vapour.

For the mean JJA meteorological conditions at Haig Glacier, Eqs. (22) and (23) give $\partial Q_L^{\downarrow}/\partial T =$ 4.7 W m⁻² °C⁻¹. Temperature increases affect Q_L^{\downarrow} through both the direct effect of higher emission temperatures and the indirect effect of higher atmospheric emissivity, with these two terms in Eq. (21) contributing 4.0 and 0.7 W m⁻² °C⁻¹, respectively.

567 The temperature sensitivity of sensible and latent heat fluxes follow

$$\frac{\partial Q_H}{\partial T} = \frac{\rho_a c_p k^2 v}{\ln(^Z/z_0) \ln(^Z/z_{0H})} , \qquad (24)$$

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$$\frac{\partial Q_E}{\partial T} = \frac{\rho_a L_p k^2 v}{\ln(Z/z_0) \ln(Z/z_{0E})} \left(\frac{\partial q_v}{\partial T}\right), \qquad (25)$$

574 where

and

$$\frac{\partial q_{\nu}}{\partial T} \approx \frac{R_d}{PR_{\nu}} \left(\frac{\partial e_{\nu}}{\partial T} \right), \tag{26}$$

577 for the dry gas-law constant $R_d = 289 \text{ J kg}^{-1} \text{ °C}^{-1}$ and air pressure *P*, under the assumption that air 578 pressure and density are constant for small changes in temperature. Table 3 gives the turbulent 579 flux sensitivities for mean JJA conditions at Haig Glacier. Perturbations to both Q_H and Q_E are 580 positive with an increase in temperature and the assumption of constant *h*. In combination with the increase in Q_L^{\downarrow} , net energy over the summer months is augmented by 12 W m⁻² for a 1°C increase in temperature. Interannual variations in summer temperature (1 σ) equal 0.8°C, giving a net energy perturbation $\delta Q_{N\sigma} = \pm 10$ W m⁻² (Table 3).

585 EnergyFluctuations in energy balance-perturbations can be related to melt rates through their 586 combined influence on Q_{N} , with $\delta \dot{m} = \delta Q_{N} / \rho_{w} L_{f}$. -Table 3 summarizes these impacts on summer 587 melt, assuming a JJA melt season (92 days). The $1^{\circ}C_{-\sigma}$ temperature increase ($\delta Q_N = 12_{\sigma} = 10$ W 588 m⁻²) is equivalent to 295236 mm of meltwater at the AWS site, if melting conditions prevail and 589 this energy can all be directed to snow/ice melt. This is a 13a 10% increase over the reference 590 levels of JJA melt, 2320 mm w.e. These are the direct impacts of higher temperatures, not 591 accounting for feedbacks or non-linearity in the seasonal evolution of melt conditions. -These 592 calculations assume that melting conditions prevail throughout the summer and all of this energy 593 can be directed to snow/ice melt, which is not strictly true. We include them because estimates of 594 the potential influence on summer melt provide an intuitive way to understand and compare 595 sensitivities. We consider more realistic relations between net energy and melt in the modelled 596 sensitivities of Section 5. 597

This <u>initial</u> scenario assumes that the warmer atmosphere contains more moisture, which is not necessarily the case. For instance, high summer temperatures in this region are commonly associated with ridging and subsidence, i.e. hot, dry conditions. If we assume that e_{wQv} is invariant with temperature (case 2 in Table 3), there is no feedback on the latent heat flux and the increase in net energy is less than with constant $h: \delta Q_N = 8.3_{\sigma} = 6.6$ W m⁻² and $\delta m = 196_{\sigma} = 157$ mm w.e.

604 However, this neglects there are additional feedbacks associated with relative humidity. If e_{yq_y} is 605 invariant, relative humidity must change to be consistent with the temperature perturbation. AnAs 606 an example, an increase of 1°C with no change in $e_x q_y$ corresponds to a decrease of 6% in mean 607 summer h at our site, to 61%. This lowers the atmospheric emissivity in Eq. (8), reduces the 608 incoming longwave radiation, and impacts $\partial \varepsilon_a / \partial T$ in Eq. (22). To be internally consistent, reduced humidity willanomalies should also be associated with decreased changes in cloud cover. For the 609 1°C temperature increase, the 6% decrease in relative humidity corresponds to an increase in 610 611 clearness index of 0.06 (Eq. 10), from 0.63 to 0.69.

The effects of these radiation feedbacks are given in Table 3. Reduced relative humidity decreases 613 Q_L^{\downarrow} and increases Q_S^{\downarrow} . The resulting increase in shortwave radiation partially offsets the decline 614 in Q_L^{\downarrow} , but there is an overall reduction in net radiation. For our parameterizations of the incoming 615 616 radiation fluxes as a function of humidity, the effect of drier air on longwave radiation is stronger than the shortwave radiation feedback. This reduces the overall sensitivity to temperature change 617 relative to the first two cases, with $\delta Q_N = 6.6\sigma = 5.3$ W m⁻² and $\delta m = 156\sigma = 125$ mm w.e. Note 618 that all of these temperature scenarios are all idealized, neglecting albedo feedbacks and other 619 620 indirect effects of a temperature change. These feedbacks are discussed and assessed in Section 5.

622 Sensitivity to Humidity and Wind

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624 Similar derivatives and energy balance sensitivities can be derived with respect to the other 625 meteorological variables, to explore the sensitivity of summer melt to different weather conditions.

626 The sensitivity of sensible and latent heat fluxes to wind perturbations follow:

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$$\frac{\partial Q_H}{\partial \nu} = \frac{\rho_a c_p k^2 (T_a - T_s)}{\ln(Z/Z_0) \ln(Z/Z_0 H)} , \qquad (27)$$

$$\frac{\partial Q_E}{\partial v} = \frac{\rho_a L_p k^2 (q_v - q_s)}{\ln(z/z_0) \ln(z/z_{0E})} , \qquad (28)$$

633 while the sensitivity to humidity is:

$$\frac{\partial Q_E}{\partial q_v} = \frac{\rho_a L_p k^2 v}{\ln(Z_{/Z_0}) \ln(Z_{/Z_0E})} .$$
⁽²⁹⁾

Incoming longwave radiation is also affected by perturbations in humidity, following:

$$\frac{\partial Q_{L}^{\downarrow}}{\partial q_{v}} = \sigma T_{a}^{4} \frac{\partial \varepsilon_{a}}{\partial q_{v}} = \sigma T_{a}^{4} \left(b \frac{\partial e_{v}}{\partial q_{v}} + c \frac{\partial h}{\partial q_{v}} \right).$$
(30)

Table 3 summaries the theoretical sensitivities for specific humidity and wind perturbations of 1 g kg⁻¹ and 1 m s⁻¹, respectively, assuming that temperature is unchanged. For the humidity, we present two scenarios: the first with perturbations to only the specific and relative humidity, and the second including the expected effects of an increase in relative humidity on cloud cover.

647 Changes in humidity directly impact the latent heat flux, and may also influence incoming 648 longwave radiation and cloud cover (hence, incoming shortwave radiation). We consider the 649 effects of a humidity perturbation with and without radiative feedbacks in Table 3. For $\delta q_v = 1$ g 650 kg⁻¹ and fixed temperature, <u>mean summer</u> relative humidity increases by 12%, to 79%; and <u>QE</u> 651 increases and <u>QN</u> increase by 10.5 Wm⁻². Interannual variations in q_v equal 0.3 gkg⁻¹, giving $\delta Q_{N\sigma}$ 652 = 3.2 Wm⁻², corresponding to a 25076-mm (443%) increase in summer melt.

Where radiation feedbacks are included, the increases in specific and relative humidity have a 654 strong influence on the atmospheric emissivity in Eq. (8), giving an increase in Q_L^{\downarrow} of 24 W m⁻². 655 This is partially offset by cloud feedbacks associated with the increased humidity. Following Eq. 656 (10), $\delta h = 12\%$ equates to a decrease in atmospheric transmissivity of 0.11, which strongly 657 658 attenuates incoming shortwave radiation. This reduces the radiative and net energyradiation by 19 659 Wm^{-2} , but the radiation feedbacks remain positive. The net impact of a <u>1- σ </u> humidity perturbation 660 661 melt.

662 663 Wind perturbations have straightforward linear effects on Q_H and Q_E , withgiving a theoreticalnet 664 sensitivity of $\partial Q_N / \partial v = +7 \text{ W m}^2$ for an increase in summer winds of 1 (m s⁻¹)⁻¹. Sensible heat 665 flux increases and evaporative cooling decreases slightly. -Winds have a low interannual variability 666 at this site, 0.2 m s⁻¹, so the associated net energy anomaly is $\delta Q_{N\sigma} = 2 \text{ W m}^{-2}$, equivalent to 50 667 mm w.e. in summer melt. 668 Formatted: Font: Not Italic

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669 *Changes in Net Shortwave* Sensitivity to the Radiation Fluxes

Net shortwave radiation is not directly dependent on air temperature, but is affected by variations in incoming shortwave radiationtop-of-atmosphere insolation, the clearness index (i.e. cloud conditions), and surface albedo.-Incoming shortwave radiation changes due to solar variability, e.g., sunspot cycles, or through variations in the atmospheric transmissivity or clearness index, though, e.g., the effects of aerosols and cloud cover. Our functional relationship for net shortwave radiation is $Q_{Snet} = Q_S^{\downarrow}(1-\alpha_s) = Q_{S\phi}\tau(1-\alpha_s)$, for potential direct insolation $Q_{S\phi}$ and clearness index τ . From Eq. (4), sensitivity to top-of-atmosphere insolation Q_0 follows

$$\frac{\partial Q_{Snet}}{\partial Q_0} = \tau \left(1 - \alpha_S\right) \cos(Z) \varphi_0^{P/P_0 \cos(Z)}, \qquad (31)$$

681 Sensitivity to the clearness index follows

$$\frac{\partial Q_{\text{snet}}^2}{\partial x} = Q_{\text{sb}}(1 - \alpha_{\text{s}}), \qquad (32)$$

685 and the albedo sensitivity is

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$$\frac{\partial Q_{\text{snet}}^-}{\partial \alpha_s} = -Q_{s\phi}\tau \,. \tag{33}$$

The insolation perturbation shown in Table 3, $\delta Q_S^{\downarrow} = 0.6 \text{ W m}^{-2}$, corresponds to a 1 W m⁻²-An anomaly of 1 W m⁻² in the top-of-atmosphere insolation, Q_0 , in Eq. (31). This is reduced to 0.6 W m⁻² as a result of the mean sky clearness index of 0.63, gives $\delta Q_S^{\downarrow} = 0.6 \text{ W m}^{-2}$, and the net radiation impact is further reduced to 0.3 W m⁻² by the surface albedo.

This is consistent with a direct estimate of sensitivity to variations in solar output through Eq. (31). For summer solstice at Haig Glacier (50.7°N, 2660 m altitude) and for $\varphi_0 = 0.84$ (clear-sky conditions), $\partial Q_{\text{snet}} / \partial Q_{50}$ in Eq. (31) can be integrated over the daily solar path. For a 1 W m⁻² change in top of atmosphere radiation, Q_0 , this gives a daily mean net shortwave perturbation of 0.25 W m⁻² at the surface. Even with clear-sky conditions, only 25% of the solar perturbation is felt at the glacier surface. The <u>The</u> net impact of daily and interannual top- of-atmosphere solar variability (e.g., such as sunspot cycles), is therefore small.

701 Incoming shortwaveIn contrast, incoming radiation fluxes and energy balance are morestrongly 702 sensitive to cloud cover, as captured through the sky clearness index, τ , although this is also muted by the surface albedo. An increase in τ of 0.05, from 0.63 to 0.68, translates to an increase in net 703 704 energy of 8 W m⁻² and a 5% increase in summer melt. Note that it could also be possible to work 705 with atmospheric transmissivity, φ , in Eq. (32), rather than the clearness index, but our 706 parameterizations of which in turn is largely governed by cloud cover. Direct, independent 707 variations in incoming shortwave and longwave radiation are reported in Table 3 for fluctuations 708 of 10 W m⁻² and for 1- σ variations in each. Sensitivity is moderate, of order 6% of the net energy. 709 710 It is more appropriate to consider co-variations of these radiation fluxes that can be expected in

association with changes in cloud cover-and. We can estimate through the sky clearness index, τ ,

712 as parameterized via Eqs. (9) and (10), which relate the atmospheric transmissivity are through τ

713 (Eqs. 9 and 10), so we use this framework here.emissivity and relative humidity to clearness index.

As an example, reduced cloud cover may be associated with a 1- σ increase in τ of 0.1, from 0.63

to 0.73. This translates to an increase in net shortwave energy of 16 W m⁻² (Table 3), but the

716 <u>change in cloud cover also impacts incoming longwave radiation. Clearer skies in the example of</u>

717 Table 3 give lower *h*, lower e_{v} , and lower Q_{L}^{\downarrow} . Latent heat flux also declines. The overall result is

718 <u>a reduction in net energy for an increase in τ . A 1- σ increase (+0.04) gives a 3% reduction in net energy.</u>

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721 <u>Sensitivity to Albedo</u>722

The sensitivity to albedo changes is comparatively high. An <u>increasechange</u> in albedo of 0.1 creates <u>a peakan</u> energy balance perturbation of more than 100 W m⁻² at local noon in midsummer. The magnitude of this effect varies with latitude, time of year, and atmospheric transmissivity. Integrated over the daily solar path and over the summer, an albedo increase of 0.1 reduces net solar radiation by -23 W m⁻², giving a 24% decrease in total summer melt. Measurements at the site indicate an interannual albedo variability of 0.06, equivalent to 14% of the net energy or $\delta m_{\sigma} = -323$ mm w.e.

731 One cannot gauge the most important meteorological variable to surface energy and mass balance 732 from the sensitivities to a unit change in Table 3, as some meteorological parameters are 733 intrinsically more variable. For instance, the sensitivity to a change in humidity appears to be 734 comparable to the sensitivity to temperature, but an increase in humidity of 1 g kg⁻¹-equals 3.3 735 standard deviations, with respect to the interannual (JJA) variability (Table 1). In contrast, summer temperature has a standard deviation of 0.8°C, so the 1°C temperature increase in Table 3 is a 736 737 weaker perturbation. Similarly, a sustained wind anomaly of 1 ms⁻¹ is a large perturbation, relative to a standard deviation of 0.2 m s⁻¹ in mean summer winds recorded at the site from 2002-2012. 738 739

To allow a direct comparison, we perturb each variable by one standard deviation (cf. Table 1) in the direction of increased melt: higher temperature, humidity, wind speed, incoming shortwave radiation, and a lower albedo. This might be representative of warm, sunny summer weather that causes high melt extent, but within the observed range of variability at Haig Glacier. The experiment assumes that weather conditions all align in a way to increase the net energy, which will not be true in general (e.g., warm summers are typically dry in the region).

747 Results are given in the last two lines of Table 3, for both mean JJA and mean MJJAS conditions. For the main summer months, JJA, Q_N is augmented by 34.6 W m⁻², giving a 35% (821 mm) 748 749 increase in summer melt. Increases in each component of the surface energy balance contribute to 750 this, but shortwave radiation is the strongest component, accounting for about half of the elevated 751 melt. This is due to both an increased clearness index (i.e. clear sky conditions) and the decreased 752 albedo. None of the surface energy fluxes is negligible in the perturbed energy budget. The 753 turbulent flux increases are mostly due to the increases in temperature and humidity. Over a 754 summer melt season, energy balance and melt anomalies are relatively insensitive to variations in 755 wind speed. This is not true on short timescales, where windy periods strongly affect the turbulent 756 heat fluxes.

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 Results are similar for the 1- σ perturbation in MJJAS conditions, with an increase in Q_N of 33.5 W m⁻². The net shortwave radiation again accounts for about half of this perturbation. If this energy balance anomaly is maintained over a five month period (and assuming melt conditions for this whole period), it equates to an additional 1320 mm of melt, a 43% increase over the mean value for the period 2002–2012.

764 <u>Summary</u>

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Overall, the results indicate a strong sensitivity of the summer energy balance and melt to 766 temperature and albedo, with weaker influences from cloud conditions, humidity, and wind speed. 767 768 These theoretical sensitivities are obviously-idealized, however, and neglect some many important feedbacks and glaciometeorological interactions that occur in glacier environments. The next two 769 770 sections examines the energy balance sensitivity at Haig Glacier within an energy balancemelt model. This allows an estimate of feedbacks associated with the evolution of albedo, 771 772 interannual variability in weather conditions, and meteorologically-consistent covariance of 773 weather variables.

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776 5. Modelled Sensitivity of the Surface Energy Balance 777

778 We use a point model of surface energy balance, described in detail in Section 2. Depending on 779 the sensitivity study (i.e. controlling for different variables or letting them freely evolve), For all 780 numerical experiments described below, we use either: (i) the direct measurements of radiation fluxes and surface albedo, or (ii) the daily model with parameterizations of the longwave radiation 781 782 fluxes, atmospheric clearness, diurnal cycles of temperature and shortwave radiation, and surface albedo evolution, following Eqs. (6), (8), (10), (17), (18), and (20). Surface temperature is 783 784 modelled from the subsurface temperature model. The mean daily forcing for the energy balance and snowpack models is taken from the glacier AWS data, and the model is run year-round for the 785 786 period 2002-2012. The May 1 snowpack thickness (winter accumulation) is specified for each year 787 based on the measured winter mass balance at the AWS site.

789 Perturbations to the observed weather from 2002-2012 are used to repeat the sensitivity analyses 790 of section 4, but with a realistic evolution of each summer melt season rather than the mean 791 summer conditions. Meteorological variables are perturbed as follows: $\pm 2^{\circ}C$ for temperature, 792 $\pm 50\%$ for specific humidity and wind, $\pm 20 \text{ Wm}^2$ for incoming shortwave radiation, 0.1 for the sky clearness index (a proxy for cloud cover), and ±0.1 for albedo. Increments are set to give 41 793 realizations in each case, spanning the range of the perturbation. For example, temperature 794 795 increments of 0.1° C are applied for the range -2 to 2° C. Each perturbation is prescribed for all 796 days in the original data, and the energy balance program is run for the period 2002-2012. In each experiment, all other meteorological variables are held constant except for those that are direct 797 impacted by a perturbation (e.g., relative humidity changes with temperature). 798

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Table 4 lists the response of mean summer (JJA) net energy, Q_N , to the different meteorological perturbations. Changes in the energy fluxes can be examined in response to the perturbations, e.g., ΔQ_N as a function of temperature anomalies, δT . We plot these values to give sensitivity curves (e.g., Figures 5 and 6), and the slope of each curve is a measure of the sensitivity, e.g., dQ_N/dT . Values in Table 4 are calculated through linear regression. The relationships area generally nonlinear, so we compute the regressions for the region of the sensitivity curve within ± 1 standard deviation ($\pm 1 \sigma$) of the reference value for each variable. This samples a more linear range and allows a better comparison with the derivatives in Table 3. Standard deviations refer to the interannual variability, as reported in Table 1. Table 4 also lists the change in net energy associated with a 1- σ increase in each variable.

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There are multiple scenarios for temperature, shown in the first four cases in Table 4. These cases 811 812 represent different assumptions about the way in which atmospheric moisture and radiation fluxes respond to a temperature perturbation. The first two cases follow the assumption that relative 813 humidity does not change. Hence, a temperature change δT is attended by a change in specific 814 humidity, δq_{ν} , to maintain constant h. This impacts latent heat flux and atmospheric emissivity. 815 Cases 1 and 2 show the net energy sensitivity to this scenario without and with albedo feedbacks. 816 The next two cases include albedo feedbacks, but assume no change in specific humidity, $\delta q_v = 0$; 817 818 hence relative humidity must respond. Cases 3 and 4 are without and with atmospheric radiation 819 feedbacks to the changed relative humidity.

Summer melt sensitivity for the four different temperature perturbation scenarios is plotted in 821 Figure 5. Case 1, the purple line in Fig. 5, lacks albedo feedbacks and indicates corresponds to a 822 net energy sensitivity of 1013 W m⁻² C⁻¹, which is comparable to the theoretical temperature 823 sensitivities in Table 3. This is throughdue to direct temperature/humidity impacts on incoming 824 825 longwave-radiation fluxes, sensible heat flux, and latent heat flux. Cases 2-4 include albedo 826 feedbacks. This can be considered to be more realistic, and the albedo feedbacks have a powerful 827 (roughly 5two-fold) amplification effect on the temperature perturbation. Under constant h, $dQ_N/dT = 5527$ W m⁻² C⁻¹ (black line in Figures 5 and cf. Figure 6a), representing a 5728% increase 828 in summer melt for a 1°C warming. This decreases by 4-126-10 Wm⁻² C⁻¹ in cases 3 and 4, where 829 q_{y} is held constant. Some of the reduced energy comes from the elimination of latent energy 830 831 feedbacks. Case 4, with atmospheric radiation feedbacks, reduces energy further as increased decreased cloud cover (via higher τ) reduces incoming shortwavelongwave radiation 832 833 more strongly than it increase longwave increases shortwave fluxes in the model. Here too, the 834 numerical model gives a similar result to the theoretical prediction. 835

Figure 6a plots the response of the different surface energy fluxes for ease 2 above. Here it can be seen that net radiation (via the absorbed the reference model, case 2. Net shortwave radiation) dominates the temperature response, over Q_H , Q_E , and Q_L^{\downarrow} . The relationship is nonlinear, with a weaker response at higher and lower temperature perturbations. This is likely associated with weaker albedo feedbacks when the snowpack is either persistent through the full summer (large negative δT) or is removed quickly in the early summer (large positive δT).

Figures 6b-6d provide similar details for perturbations in humidity, wind, shortwave radiationclearness index, and albedo (cases 5-9 in Table 4). Sensitivity to humidity changes is relatively strong, through the combined impacts of latent and longwave fluxes (Fig. 6b). Case 6 is shown in this figure, including feedbacks on the atmospheric radiation. Incoming longwave radiation (orange line in Fig. 6b) is strongly augmented by the increases in absolute and relative humidity, and accounts for about 70% of the net energy sensitivity to specific humidity. It is partially offset by cloud feedbacks, however, such that the net radiation (red line) sensitivity is 850 much less, and is comparable to the latent heat flux sensitivity. which reduce incoming shortwave
 851 radiation.

For increases in both temperature and humidity, the mean summer latent heat flux switches sign from negative (Table 2) to positive; that is, latent heat flux becomes a source rather than sink of energy under warmer and wetter conditions. In contrast, latent heat flux remains negative, but small, under increases in wind speed (Figure 6c). Energy balance sensitivity to wind perturbations is primarily associated with the sensible heat flux.

Shortwave radiation Net energy perturbations due to albedo and clearness index in Figure 6d are independent of each other, but are plotted together for convenience. Net energy sensitivity to perturbations in incoming shortwave radiation are attenuated through the albedo, which reduces the impact of changes in top-of atmosphere insolation or atmospheric transmissivity on surface energy variability. Albedo sensitivity over athe range of ± 0.1 is relatively high, with a decrease in net energy of -2227 W m⁻² (2328%) for an increase in albedo of 0.1.

866 The sensitivities computed with the surface energy balance model are generally consistent with the theoretical sensitivities in Section 4, with the exception of the strong amplification introduced 867 by albedo feedbacks. For realistic perturbations, such as a $1-\sigma$ increase in each variable, sensitivity 868 869 to temperature is far higher than for the other variables. Interannual variations in albedo are the 870 next strongest influence, followed by incoming shortwave radiation (i.e. Changes in sky clearness index (atmospheric transmissivity) have a lower impact, due to the compensating influences on 871 872 incoming shortwave and longwave radiation. Reduced cloud cover) and humidity. Interannual variations in wind have only a minor influence on the summer energy budget. (higher τ) gives an 873 overall reduction in net energy at our site, as longwave radiation effects are dominant. 874

876 *Sensitivity to Winter Snow Accumulation*

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Changes in the winter mass balance <u>the spring snowpack have an additionalalso</u> influence on the evolution of the summer melt season, which we have not examined above. Interannual variability in the snowpackamount of snow is implicit in the simulations, as the spring (May 1) snowpack <u>depth</u> is initialized with the measured winter mass balance, b_{W_3} as measured at the AWS site_for_each Mayyear, b_W (Marshall, 2014). However, we have these experiments do not eontrolledcontrol for thisthe influence of snow depth on summer melt extent.

To examine this, we use the mean annual meteorological record from 2002-2012 and use this to 885 886 force the energy balance model throughover a range of winter mass balance conditions, $b_w \in [0.36,$ 887 2.36] m w.e. The This is ±1 m w.e. relative to the mean observed value from 2002-2013 was at the 888 AWS site, 1.36 ± 0.27 m w.e. at the AWS site. The melt model is run through 11 years of weather, 889 2002-2012, with the different values of winter mass balance as an initial condition. Figure 7 plots the modelledaverage evolution of the seasonal snowpack depth and albedo through the summer 890 891 melt season, from May through September, for this suite of experiments. The snowpack depth is 892 expressed in meters, as used for the subsurface temperature model. Fig. 7c shows the Transitions 893 from seasonal snow to ice span from early July to mid-September. Albedo spikes in Fig. 7b are due to summer snow events, which become more frequent as temperatures cool in September. 894 895

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896 The net energy balance perturbation perturbations that accompanies accompany these scenarios. 897 There is relatively little effect at this site are shown for the deeper snowpacks, once the seasonal 898 snow is thick two choices of the minimum snow albedo (Fig. 7c). Observations of late-summer 899 snow at the site are in the range 0.3-0.4, the two values presented here. The plot is asymmetric; net energy is more sensitive to reduced winter snow depths, which result in an earlier transition to 900 901 exposed glacier ice. A 20% (1 σ) reduction in b_w gives a net energy increase of about 4 W m⁻² (4%), 902 and the sensitivity increases non-linearly with increasingly lower snow depths. The influence from 903 a deep winter snowpack is comparatively muted: $1-2 \text{ Wm}^{-2}$ reductions in O_N for a 20% increase 904 in the winter snow thickness. Perturbations in Q_N asymptote once seasonal snow is deep enough 905 to survive through the summer. For low values of b_{w_s} net energy increases by about 10 W m⁻² 906 (10%) This is 907

908 The influence of the winter snowpack at this site is similar in magnitude to the net energy impacts 909 of interannual variations in the humidity or wind speed, but less important to the summer melt than 910 observed variations in summer temperature, albedo, or cloud cover. This result is partly due to the 911 relatively low contrast between late-summer snow albedo and bare-ice albedo at this site. If late-912 summer snow has a higher albedo, a deep winter snowpack is more effective at reducing the net 913 energy and summer melt. The shape of the sensitivity curve would change for locations with 914 higher-albedo snow, and also for sites in the lower ablation zone, where ice is exposed early in the 915 melt season. A heavy winter snowpack would have a comparatively stronger role in this case. The 916 result in Figure 7 is therefore more site-specific than for the other meteorological perturbations. 917

919 6. NARR-based Surface Energy Balance Reconstructions, 1979-2014

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921 To examine energy balance sensitivity over a longer time period and with joint variation in 922 meteorological variables, we run the energy balance model forced by North American Regional 923 Reanalysis (NARR) atmospheric reconstructions from 1979 to 2014 (Mesinger et al., 2006). This 924 provides a more complete picture of interannual variability, while comparison of NARR 925 predictions with measurements over the period 2002-2012 also allows us to assess the skill with 926 which fluctuations in surface energy balance and summer melt can be captured in an atmospheric 927 model that does not explicitly resolve the alpine and glacier conditions.

We use a perturbation approach as in Section 5, taking NARR daily meteorological fields as 928 929 anomalies relative to the mean NARR conditions for the period 2002-2012. Anomalies in near-930 surface temperature, specific humidity, wind speed, pressure, incoming shortwave radiation and 931 incoming longwave radiation are used to drive the model for the 36-year period 1979-2014. in situ 932 conditions for the period 2002-2012. The main difference from Section 5 is thatPerturbations are 933 introduced as anomalies relative to the mean observed conditions. NARR input fields allow us to 934 introduce multiple perturbations at once, with magnitudes that are physically meaningful and 935 meteorologically-consistent covariance of variables.

936 NARR has an effective spatial resolution of 32 km, and we extract mean daily data from the grid

cell over Haig Glacier. This grid cell has an elevation of 2214 m, about 450 m lower than the AWS
 site. Anomalies in near-surface temperature, specific humidity, wind speed, pressure, incoming

939 shortwave radiation and incoming longwave radiation are used to drive the model for the 36-year

940 period 1979-2014. By using daily weather anomalies, we avoid most biases associated with the

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different altitude of the NARR grid cell. However, variations in some fields such as specific 941 942 humidity, pressure, and temperature can be larger at lower elevations and over non-glacierized 943 land surface types. Since we use meteorological fluctuations as perturbations, this is potentially 944 problematic. Inspection of the summer variance in the different meteorological inputs over the 36yearreference period 2002-2012 indicates that this does not appear to be an issue, however. 945 Standard deviations of each variable, calculated from mean JJA values, are as follows: 946 temperature, 0.8°C; specific humidity, 0.32 g kg^{-1} ; wind speed, 0.23 m s^{-1} ; incoming shortwave 947 948 radiation, 76 W m⁻²; and incoming longwave radiation, 43 W m⁻². Temperature, humidity, and wind 949 values are identical equivalent to the observed range of variability from 2002-2012 (Table 1), but 950 the radiation fluxes are less variable. The effects of a lower elevation in the NARR grid cell appear to be less than those associated with systematic biases in the reanalysis, e.g., not enough variability 951 952 in cloud or surface conditions.

953 The energy balance model requires an estimate of winter snow accumulation. We base this on 954 cumulative NARR precipitation for the period September to May of each year, normalized to the 955 observed value of 1.36 m w.e. at the Haig Glacier AWS site. This permits interannual variability 956 in the winter snowpack <u>thickness</u> to be included in the simulations, by scaling the mean observed 957 value up or down based on the NARR winter precipitation totals. We use this as an initial condition 958 for the <u>summer</u> melt model. Our focus is on the <u>summer energy budget and (i.e., for May 1</u> 959 snow/ice melt, rather than annual mass balance reconstructions, depth).

960 We examine the sensitivity of net summer energy balance and melt to interannual variations in each weather variable in the NARR forcing. Table 5 reports the NARR-based surface energy fluxes 961 and melt for JJA and MJJAS, averaged over the period 1979-2014. Mean values are all within 2 962 W m⁻² of the reference surface energy fluxes (Table 2), derived from the in situ data, but there are 963 some significant differences in the standard deviation, which is a measure of the interannual 964 965 variability. As noted above, incoming shortwave radiation has about half of the variability in the 36-year NARR record as observed in the 11-year measurement period, and variance in incoming 966 967 longwave radiation is also less than observed. This implies more uniform summer cloud conditions in the reanalysis, compared to the observational period. 968

Average summer albedo is also less variable in the model than the observations. The, and the mean 969 970 value in the NARR-forced model is also lower than the observations, particularly too low for May 971 through September (0.55 vs. an observed value of 0.60). Most of this difference is associated with 972 a low value of September albedo in the model; we are likelygenerally underestimating September snow events and predicting too late a transition from end-of-summer to the winter accumulation 973 974 season. This transition occurs sometime in September or October each year in our study period. September is mixed on the glacier, with fresh snowfall alternating with periods of melting. This 975 raises the average albedo on the glacier, but our albedo parameterization does not fully capture 976 977 this.

978Figure 8a plots time series of the NARR-forced surface energy balance terms, and Figures 8b-8d979shows the relations between net energy and selected meteorological variables. These are equivalent980to the relations between total summer melt, and These provide a visual indication of the strength981of each variable as a predictor of summer melt. Regressions through these data points give982estimates of net energy sensitivity, e.g. $dO_N/dT \partial O_N/\partial T$, as seen in actual realizations of the summer

weather conditions. These values gradients can be thought of as the melt sensitivity to interannual
 variability or trends in each weather variable.

985 The resulting sensitivities are given in Table 6, as well as linear correlation valuescoefficients 986 between Q_N and all glaciometeorological variables that are used in the energy balance model. 987 These simulations are forced with NARR radiation flux anomalies, so we do not parameterize the 988 incoming longwave or shortwave radiation in these tests. The clearness index, τ , is not used, but it 989 can be calculated from the NARR relative humidity estimate, via Eq. (10), or more directly through 990 the fraction of incoming shortwave radiation relative to the clear-sky potential radiation. We test both approaches and find similar results. Values for $\partial Q_N / \partial \tau$ reported in Table 6 are averaged from 991 992 the two approaches. We also report the direct relation between NARR total cloud cover and net energy; cloud cover is available in the reanalysis, but we do not have in situ data to compare with. 993

994 Temperature and albedo have the strongest influences on summer energy balance and melt, while 995 incoming longwave radiation and. Fluctuations in specific humidity also emerge as significant 996 influences on and incoming longwave radiation also correlate strongly with interannual variability 997 in the summer energy budget. Other variables, includingWind speed, cloud conditions, and 998 incoming shortwave radiation, do not strongly contribute to the year-to-year variations in summer 999 melt over the NARR period. There is a weak, positive relationship between the clearness index and net radiation in the NARR-forced results, indicating that increased shortwave radiation 1000 1001 associated with reduced cloud cover has a stronger role than the associated reduction in longwave 1002 radiation.

1003 These sensitivities can be compared with those in Section 5 that include full albedo and 1004 atmospheric radiation feedbacks, (Table 4), but they differ in that the NARR forcing has multiple 1005 joint perturbations. This is realistic as the meteorological variables often co-vary systematically, 1006 but it means that it is not possible to isolate the role of a single variable, such as temperature. A 1007 temperature change impacts several of the energy fluxes, but coincident changes in e.g., humidity 1008 and radiation fluxes, may reinforce or reduce the temperature impacts. The results Results in Table 1009 6 should therefore be interpreted as the 'net' or 'effective' influence of each weather variable on 1010 the summer energy balance, and some of them (e.g., relative humidity) may have correlations that 1011 are more coincidental than casual. - Most results are nonetheless similar in magnitude to the 1012 theoretical and modelling results (Tables 3 and 4), which are based on the in situ data. The largest exception is the relation between clearness index (cloud cover) and net energy, which is opposite 1013 1014 in sign.

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1016 7. Discussion

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1018 We takehave taken three different approaches to estimate summer (JJA) energy balance and melt 1019 sensitivity at Haig Glacier in the Canadian Rocky Mountains: (i) theoretical, perturbing one 1020 meteorological variable at a time with reference to the mean meteorological conditions, (ii) 1021 through a numerical model of the surface energy balance, restricting model experiments to single 1022 perturbations but allowing for internal feedbacks to be modelled, and (iii) through reanalysis-based 1023 meteorological perturbations from a regional climate reanalysis, allowing multiple variables to change at once in a way that is meteorologically consistent. The latter also permits a longer time
 period to be modelled (36 years), to examine. Here we briefly summarize and interpret the role
 ofintegrated results from these different meteorological variables in interannual variability of
 glacier melt. methods.

In all three approaches to assess the sensitivity, perturbations in <u>Haig Glacier Energy Balance</u>
 Sensitivities and Feedbacks
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1032 Interannual variations in temperature and albedo have the strongest influence on summer melt 1033 extent.energy balance in all three approaches to assessing Haig Glacier melt sensitivity (Figure 9). 1034 Fluctuations in humidity and incoming shortwave and longwave radiation, via cloud cover, are 1035 also important to the summer energy budget, while variations in cloud cover (τ) , wind speed, and 1036 the winter snowpack thickness are less influential on the summer energy budget and melt extent 1037 at this site. We discuss each weather variable in a more detail in the next paragraphs.

1039 *Temperature Perturbations*

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1041 Temperature changes are generally thought of as the main driver of glacier advance and retreat, 1042 through various combined influences on the surface energy budget, snow accumulation, and 1043 summer melt season. Numerous studies have estimated glacier melt or mass balance sensitivitySensitivities to elimate warming. These sensitivitiestemperature are commonly 1044 expressed as the change in summer or net mass balance per unit warming. Sample mass balance 1045 sensitivities reported in the literature are -0.6 m w.e. °C⁻¹ on Morteratschgletscher, Switzerland 1046 (Klok and Oerlemans, 2004) and Illecillewaet Glacier, British Columbia (Hirose and Marshall, 1047 2013), -0.68 ± 0.05 m w.e. °C⁻¹ for a suite of glaciers in Switzerland (Huss and Fischer, 2016), and 1048 -0.86 m w.e. °C⁻¹ on South Cascade Glacier, Washington (Anslow et al., 2008). Values as high 1049 as -2.0 m w.e. °C⁻¹ are reported for Brewster Glacier, New Zealand (Anderson et al., 2010). 1050

These values are for the annual mass balance, but they are dominated by the summer melt response 1052 1053 to warming, and they. They represent a melt sensitivity of about $30\% \, {}^{\circ}C^{-1}$ for the examples in the 1054 Alps and western North America. When we introduce temperature perturbations in the absence of 1055 albedo feedbacks, we find a relatively muted energy balance response, about $\pm 10 \text{ Wm}^{-2} 13 \% \text{ °C}^{-1}$ 1056 averaged over. The increase in net energy is distributed about equally across the main summer 1057 melt season, JJA. This equates to a meltsensible heat flux, incoming longwave radiation, and latent 1058 heat flux, and we have similar results for both the theoretical and numerically-modelled 1059 temperature perturbations. Albedo feedbacks increase the net energy sensitivity of about 10to 28 % °C⁻¹ at this site, or -0.266 m w.e. °C⁻¹, in accord with previous studies. The exact number 1060 1061 depends on-the assumptions about humidity; if specific humidity increases with temperature (e.g., 1062 by holding relative humidity constant), temperature sensitivity is higher. This idealized warming 1063 is distributed about equally across the sensible heat flux, incoming longwave radiation, and latent 1064 heat flux, and we have the same result for both the theoretical and numerically modelled 1065 temperature perturbations.

1067 When The albedo feedbacks are activated in our model, the impacts of a feedback results from two 1068 main ways that temperature change are amplified dramatically at this site: roughly a five-fold 1069 increase in the melt sensitivity, to $\sim 50\% \, {}^{\circ}C^{-1}$. This is equivalent to about -1.2 m w.e. ${}^{\circ}C^{-1}$. The Formatted: Font: Italic

1070 relationship is nonlinear and is strongest near the observed present-day temperature. Several 1071 effects can contribute to this strong amplification of the temperature signal in the melt model. A 1072 longer and influences the seasonal albedo evolution. A more intense melt season gives rise to a 1073 lower albedo through higher impurity concentration and water contentsnow albedo and an earlier 1074 transition from seasonal snow cover to glacial ice. These positive feedbacks also operate (in 1075 reverse) under a cool perturbation. We do not explicitly model impurities or snow-albedo 1076 processes (e.g., grain metamorphosis, effects of snow-water content on the albedo), but we 1077 parameterize the seasonal albedo evolution as a function of cumulative PDD (Eq. 20). This is a 1078 rough proxy for cumulative melt effects that lower the albedo, and is empirically supported, but 1079 positive degree days are a direct function of temperature so this may make our albedo model 1080 overly20), which makes the model directly sensitive to temperature perturbations.

1082 Temperature changes have several additional, indirect impacts, including; (i) a longer melt season 1083 starting earlier and ending later, (ii) a greater fraction of time with surface temperatures at the 1084 melting point during the year, i.e., with reduced overnight cooling and refreezing, and (iii) an 1085 increase in the frequency of summer rain vs. snow events. Summer snow events have an important 1086 impact on surface albedo, with fresh snow strongly attenuating melt. Each of these processes 1087 contributes to the strong impact of increased temperaturestemperature anomalies on glacier melt. 1088 Combined with the albedo feedbacks, these processes and the model results help to explain why 1089 glaciers are sostrongly sensitive to temperature change, as they clearly are in natural settings (e.g., 1090 Marzeion et al., 2014).

1092 When multiple meteorological perturbations are introduced at the same time, in the NARR-based 1093 surface energy balance modelling, interannual temperature fluctuations appear to be weaker than 1094 the sensitivity experiments would suggest, ~14% °C⁻¹, although mean summer net energy and 1095 temperature are highly correlated (r = 0.84). All feedbacks discussed above are active in the 1096 NARR-based simulations. The impacts of temperature variability on net energy and melt could be 1097 partially compensated by other systematic changes in the energy budget. For instance, warm 1098 temperatures could be associated with calm, dry conditions that reduce the incoming longwave 1099 radiation and the turbulent fluxes. NARR mean summer temperature over the 36 year period is negatively correlated with wind speed (r = -0.11) and cloud cover (r = -0.50), which supports this 1100 1101 possibility.

1103 Albedo Perturbations

1105 Direct changes to albedo have an influence on summer energy balance and melt extent that is 1106 comparable to the temperature influence. The three different methods of gauging albedo sensitivity 1107 give similar results, a summer energy balance impact of 22-26 Wm⁻², ~17% for a change in albedo of 0.1 (Tables 3, 4 and 6). Interannual albedo fluctuations are associated with net energy and melt 1108 variations of about 12%, a large fraction of the interannual variability. 1109

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1111 equal to the interannual albedo fluctuations, 0.06. Mean summer albedo differences arise as a feedback to other meteorological forcings that drive the summer snowpack evolution, such as 1112 1113 temperature. Interannual snow melt, but interannual albedo variations can also occur more directly, 1114 as a consequence of frequent summer snowfall events or, as a resultfunction of low or high winter 1115

accumulation totals, which influence how long the seasonal snowpack will persist through the

summer. Haig Glacier is also vulnerableor due to impurity loading (e.g., black carbon deposition)).
The latter has been observed in association with forest fires in British Columbia. ExtensiveStrong
fire seasons have occurred twice during theour period of study-at this site, in 2003 and 2015, and
each left a visibly and measurably darker glacier surface. For instance, the average albedo recorded
at the AWS site in August 2003 was 0.13.

1122 Humidity Perturbations

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124 Changes in humidity directly affect the latent heat flux, but they also influence the incoming 125 radiation fluxes in our parameterizations, through the atmospheric emissivity and the clearness 126 index. Net energy has <u>We found</u> a large unit sensitivity to specific humidity, but interannual 127 variability at the site is relatively low, such that fluctuations in specific humidity do not strongly 128 influence summer melt extent. This is true for both the observational period and in the NARR-129 forced reconstructions.

1131 Mean summer latent heat flux was weakly negative through the observational period at the Haig 1132 Glacier AWS site, but increases in specific humidity increase this flux and it switches signs to a 1133 small positive flux with a 5% (~0.2 gkg⁻¹) increase in JJA humidity. Radiation fluxes are more 1134 strongly sensitive to humidity, at least as parameterized in our study, with relative humidity being 1135 the main influence. Atmospheric emissivity and incoming longwave radiation increase with 1136 humidity, while incoming shortwave radiation is reduced, based on the empirical link between 1137 relative humidity and cloud cover. These fluxes largely compensate and offset each other, but the 1138 longwave radiation has a stronger response in our results, for both the theoretical and modelled 1139 perturbations in humidity. There is a net cooling influence when specific humidity is reduced, as 1140 decreases in incoming longwave radiation exceed the attendant increases in shortwave radiation. 1141 Increases in humidity give an increase in net radiation, as gains in incoming longwave radiation 1142 again exceed the reductions in net shortwave radiation. The balance will depend on the surface 1143 albedo, which reduces the magnitude of shortwave radiation anomalies in the net energy budget.

1145 Incoming Shortwave Radiation

Top of atmosphere shortwave radiation fluctuations, i.e. solar variability, have only minor
 influences on glacier melt, as top of atmosphere forcing is diminished through atmospheric
 extinction and the glacier surface albedo. Fluctuations of ~3 W m⁻² are attenuated to 1 W m⁻², which
 is negligible relative to the daily and interannual variability associated with cloud cover.

1152 The latter does have a significant impact on year to year melt conditions. Surface level interannual 1153 variability in shortwave radiation forcing equates to fluctuations of about 6 W m⁻² (6%) of the JJA 1154 net energy budget, and can compound the effects of warm temperatures in associated with hot, 1155 dry, clear-sky periods on the glacier. This is empirically borne out at the site, but shortwave 1156 radiation fluctuations are less important in the NARR-driven energy balance than they are in the 1157 observations. NARR shortwave radiation variations correlate positively but weakly with summer 1158 melt, and interannual variability of incoming shortwave radiation is muted in the reanalysis. The 1159 NARR dataset may not be picking up some of the persistent ridging conditions which are observed 1160 to drive strong summer melt events at the site.

1162 *Winter Mass Balance*

1163 1164 We found only a minorweak influence of winter mass balance on the summer melt extent, based 1165 on observed interannual variability in winter snow accumulation as well as sensitivity experiments. 1166 . A low snowpack depth has a greater impact, through an earlier transition to low-albedo bare ice. A deep winter snowpack has the opposite influence, supporting a higher average summer albedo, 1167 but the influence is weaker because the AWS site is in the upper ablation area, where the seasonal 1168 1169 snowpack persists until late summer in most years. The effects of greater winter accumulation 1170 plateau once there is enough snow to survive the summer; beyond this point, additional snow has 1171 no effect on the summer albedo or melt extent. Sensitivity to winter mass balance would likely be 1172 stronger at lower altitudes on the glacier, and for the overall glacier mass balance. 1173

1174 *Multivariate Perturbations*

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1176 Meteorological variables do not vary as idealistically as in the simple experiments presented in 1177 this paper. In reality, meteorological variables all vary at once, and different weather systems will 1178 have tendencies for the combined meteorological perturbations to compensate (buffer) or 1179 accentuate (amplify) impacts on energy balance and melt. This is implicit in the NARR forced 1180 simulations, which sample a 36-year record of interannual variability with physically consistent 1181 covariance of meteorological variables.

1183 Humidity changes can also be considered a feedback to temperature, but this is not certain; specific 1184 humidity varies as a function of local- to synoptic-scale moisture sources and weather patterns, 1185 and these are not necessarily coupled to temperature conditions. For instance, warm conditions at 1186 Haig Glacier often accompany anticyclonic ridging in the summer months, during which time 1187 southerly flows and upper-level subsidence promote dry, clear-sky conditions (low q_{y} and h). At 1188 other times, westerly flows bring warm, moist Pacific air masses and humidity, temperature, and cloud cover co-vary. Interannual variability in specific humidity has a significant impact on 1189 summer energy and melt extent, an ~8% change for a perturbation of 0.3 g kg⁻¹ (1 σ). This effects 1190 1191 net energy through impacts on the latent heat flux and incoming longwave radiation. The latter is 1192 partially compensated by accompanying changes in incoming shortwave radiation. 1193

With all three methods, cloud cover shows up as a relatively weak influence on summer net energy at this site, ~4% for a 1-σ variation in the clearness index (Figure 9). This result is a consequence of the offsetting effects of cloud cover on the shortwave and longwave fluxes. The sign of the relationship is also uncertain. In isolation, interannual fluctuations in shortwave and longwave radiation have a moderate influence on the summer net energy (Figure 9), so these are important; they are just not simply related to the cloud cover index, τ.

1201 <u>NARR Results</u> 1202

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NARR results are broadly consistent with the <u>idealized experiments</u> <u>in situ-based</u> and theoretical sensitivities, in terms of the relative importance of different meteorological parameters to interannual variability in summer energy balance and melt. <u>The influence of interannual temperature fluctuations appear to be weaker than the other sensitivity experiments would suggest, ~15% °C⁻¹. All feedbacks discussed above are active in the NARR-based simulations. The impacts
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of temperature variability on net energy and melt could be partially compensated by other
 systematic changes in the energy budget. TheFor instance, warm temperatures are often associated
 with calm, clear-sky conditions that reduce the incoming longwave radiation and the turbulent
 fluxes.

1213Temperature nonetheless emerges as the most important variable explaining interannual variations1214in net energy. Mean summer net energy and temperature are highly correlated (r = 0.84). This1215reinforces the argument that temperature indices offer a good proxy for net energy and summer1216melt extent (e.g., Ohmura, 1987).

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1218 There are two main other discrepancies are that the temperature sensitivity and yearin the NARRforced results. Year-to-year variance in incoming shortwave radiation are less than expected. This 1219 1220 may be connected, as the highest observed summer mass losses have occurred in hot, dry summers, 1221 where there were strong positive anomalies in both shortwave radiation and temperature. We do 1222 notradiation fluxes is less than observed, pointing to poor representation of interannual cloud 1223 variability in the reanalysis. The variability is still positively correlated with the in situ data (e.g., 1224 r = 0.50 for the correlation between incoming JJA shortwave radiation in NARR and in the data 1225 from 2002-2012). Hence, NARR is picking up some of the observed variability, but it is muted. 1226 The sensitivities to the radiation fluxes may still be representative, as there is still some interannual 1227 variability for which on can assess the relation between Q_N and the radiation fluxes. However, the 1228 poor representation of the radiation fluxes and cloud conditions can be expected to reduce the skill 1229 of NARR-forced mass and energy balance reconstructions; this requires further study. 1230

1231 The other main difference with the NARR forcing is a switch in sign in the sensitivity to changes 1232 in cloud cover, as analyzed through either τ or the NARR-predicted total cloud cover. Clear-sky 1233 conditions have a positive relation with Q_N in the NARR-driven simulations, signalling that 1234 incoming shortwave radiation fluxes exert more influence than incoming longwave fluxes for net 1235 summer energy. Clear-sky conditions (less cloud cover) give increased shortwave radiation and a 1236 lesser decrease in longwave radiation, resulting in increased net energy. The theoretical and in situ 1237 sensitivities predict the opposite result, reduced net energy with clearer skies. The relationship is 1238 relatively weak, so it is possible that there are confounding variables in the NARR simulations 1239 once again, such as temperature effects masking the cloud relationship. 1240

1241 We do not test the ability and skill of NARR-forced energy and mass balance reconstructions here. 1242 This requires further study. In general, the perturbation method eliminates biases in the mean 1243 NARR variables, but a realistic representation of the variability and long-term trends in reanalysis 1244 fields is important to realistic representations of the glacier mas balance record and meltwater 1245 runoff. It would be instructive to analyze the synoptic weather patterns and weather anomalies in 1246 high-melt vs. low-melt summers in the NARR-driven simulations. AnWe recommend an 1247 investigation of specific weather systems and their associated meteorological and energy balance 1248 conditions is recommended forin followup work. 1249

1250 Daily NARR-based forcings for the surface energy balance and summer melt/mass balance worked 1251 well in this study, when taken as anomalies to the mean observed conditions. This method for 1252 calculating the surface energy balance is a general approach that can be adopted to explore 1253 meteorological influences on melt in different glacier environments, or to model variations in time 1254 at a particular site. The method could similarly be applied to climate model output for future 1255 projections. 1256 1257 1258 1259 Representativeness of the Results 1260 1261 We have designed the sensitivity approach and the model to be applicable in regional studies, e.g. in a distributed model of glacier energy balance, forced by climate model reanalyses or projections. 1262 However, we did not expand our scope to other sites within the present study. In principle, the 1263 1264 theoretical sensitivities (i.e. from the same set of equations) could be calculated for different 1265 baseline meteorological conditions, such as maritime or tropical environments. The method, rather 1266 than the specific Haig Glacier results, could be exported to other glacierized environments. 1267 1268 At regional scales, Haig Glacier energy balance sensitivities might be more transferrable, since 1269 similar summer climate conditions prevail across the Canadian Rocky Mountains (Ebrahimi and 1270 Marshall, 2015). Regional, multi-year reconstructions of glacier meltwater runoff might be 1271 feasible through a perturbation approach to summer mass balance, driven by meteorological 1272 anomalies from station data or climate models. This needs to be tested, however, for sensitive 1273 parameterizations such as the albedo model. It is uncertain whether the Haig glacier bare-ice and 1274 old-snow albedo are regionally representative. 1275 1276 Within Haig Glacier itself, our AWS site is in the upper ablation area, near the equilibrium ELA. 1277 Results are specific to the snow and ice albedo, snowpack depth, and meteorological/energy 1278 balance conditions at this location. We have not examined the representativeness of the results to 1279 other parts of the glacier, but summer melt extent and mass balance at the AWS site are strongly 1280 correlated with glacier-wide mass balance. We recommend additional work to calculate an average

set of glacier sensitivities and assess whether the values presented here are representative. We
 suspect that sensitivity of net energy to winter snow depth and the strength of albedo feedbacks
 will vary across the glacier.

1285 <u>Recommended Model Improvements</u> 1286

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1287 Model improvements are certainly possible, particularly recommended with respect to our 1288 treatment of the glacier surface albedo and precipitation modelling. Based on our systematic 1289 underestimation of September albedo, a better treatment of late-summer snow accumulation and 1290 the transition to the winter accumulation season is needed. The energy balance, albedo, and melt 1291 models perform well in the core summer melt season, June through August, when summer 1292 snowfall is infrequent and impacts on the albedo are transient. We systematically underestimate 1293 September albedo, however; better treatments of late-summer snow accumulation and the 1294 transition to the winter accumulation season are needed.

1296 Our meltwater drainage model is also simplistic. We assume that water drains efficiently from the 1297 glacier surface, but in fact water has been observed to pond and refreeze here. This acts as an 1298 energy sink and would reduceon the surface. Re-melting and of this superimposed ice consumes 1299 energy and reduces the total summer runoff from the site. A more realistic treatment of year-round snow accumulation is also needed in order to carry out model-based glacier mass balance reconstructions. We rely on observed winter mass balance for the studies here, but historical reconstructions and future projections require a way to reliably estimate snow accumulation from climate models. NARR precipitation in the Haig Glacier grid cell poorly represents the observed winter accumulation totals.

We have done tests to verify that the daily, parameterized model performs well relative to direct forcing with 30-minute AWS data, but some simplifications embedded in the daily model need to be examined. For instance, we assume constant cloud cover/clearness index over the day; systematic diurnal variations in cloud cover would affect the net radiation in ways that we do not capture. Overnight clouds serve to increase energy flux to the glacier, while daytime clouds reduce the incoming radiation. Effects like these become complicated to model or parameterize, but could bias our sensitivity results to cloud cover.

1315 8. Conclusions

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1317 Theoretical Sensitivity studies presented here extend the foundational work of Oerlemans, and 1318 numerical models exploring surface energy Fortuin (1992) and others, which has generally been 1319 done on glacier mass balance on Haig Glaciersensitivity to changes in the Canadian Rocky 1320 Mountains providetemperature and precipitation. Our study is limited to summer mass balance at one location, but our results offer insight into summer melt sensitivity to the influence of different 1321 1322 meteorological variables and energy fluxes, their year-to-year variability, and the role of isolated 1323 vs. collective forcings. The study is based on a 11-year record of glaciometeorological conditions from an AWS site in the upper ablation area, 2002-2012. Numerical experiments examine 1324 1325 perturbations to variables in isolation, with internal, feedbacks to the, and interactions on summer 1326 melt season evolution, and with multiple perturbations to meteorological variables, extent.

1328There is a good correspondence between the theoretical sensitivities and those derived from North1329American Regional Reanalyses from 1979-2014. the numerical energy balance model, when1330feedbacks are omitted. This supports the potential application of the theoretical sensitivities to1331explore energy balance sensitivities under different climate regimes. This method can be1332transferred directly to other sites.

The model runs year round, to simulate sub-surface temperature evolution in the winter snowpack
 and to include the complete summer melt season (May to September), but our analysis concentrates
 on mean summer (JJA) surface energy balance and melt. Just over 80% of the annual melt at the
 site occurs in JJA, and we find similar energy balance and melt sensitivity to meteorological
 variability when we look at MJJAS.

1340 Temperature and albedo variations exert the strongest controls on year-to-year variability in 1341 summer melt at our site. While albedo can fluctuate independent of temperature, e.g., through the 1342 influence of the winter snowpack <u>depth</u> or aerosol loading, it is also a powerful feedback 1343 mechanism to temperature and melt season evolution. In our model, albedo feedbacks give a 1344 <u>fivetwo</u>-fold increase in the net energy balance sensitivity to a temperature perturbation, 1345 amplifying the summer melt response from 1013% °C⁻¹ to ~5028% °C⁻¹. Temperature and albedo Formatted: Font color: Text 1

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1346 fluctuations are also the strongest influences on interannual melt variations in the NARR-forced surface energy balance, but the melt sensitivity to temperature variations is about 15% $^{\circ}C^{-1}$, 1347 1348 weaker than our result from the control experiments. This may be because the co-variation of other 1349 variables in the surface energy balance partially offsets the temperature forcing. For example, 1350 temperature increases are associated with lower relative humidity and cloud cover, which reduces 1351 incoming longwave radiation. It is also possible that NARR climate reconstructions are not 1352 adequately capturing the weather conditions and their interannual variability over the field site, as 1353 suggested by a poor representation of cloud conditions and radiation fluxes compared to in situ 1354 observations.

1355 Other meteorological variables cannot be neglected in the surface energy balance and its

1356 interannual variability. At Haig Glacier, incoming shortwave radiation fluctuations are particularly 1357 influential on summer melt extent. The strongest melt seasons and most negative mass balance

1357 Infinite function of summer more extent. The strongest metric seasons and most negative mass outlinee 1358 years on record at the site, 2003, 2006, and 2015 (not shown), were each associated with persistent 1359 anticyclonic ridging in the summer months, giving warm, dry, clear sky conditions, i.e. co-

1360 variance of strong positive anomalies in temperature and incoming shortwave radiation.

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Humidity fluctuations are also effective in influencing the net energy, through their impacts on latent heat flux and incoming radiation fluxes. Wind <u>speed</u>, <u>cloud</u> conditions, and the winter snowpack <u>thickness</u> are less important to the summer energy balance and melt extent at our site. The relationship with cloud conditions is statistically weak and we do not have confidence in the sign; we recommend further work to assess the influence of cloud cover on summer net radiation at this site and elsewhere.

1368TheseOur results apply to just one location, in the upper ablation area of a relatively smallsuggest1369that it is may be reasonable to model glacier. While this is a typical mid-latitude mountain glacier,1370other parts of the glacier (i.e. the lower ablation zone) and other glaciers will have different energy1371balance sensitivities to meteorological conditions. This contribution is an initial step, introducing1372an energy balance melt sensitivity approach to quantify glacier sensitivity to meteorologicalat this1373site to temperature forcing, while ignoring variability and elimate change. Further work is needed1374and recommended to extend this approach to different elimate regimes.

Qur analyses and results focus on the summer melt season; additional work is also needed to extend
 this to the broader implications for glacier mass balance, including winter mass balance and its
 sensitivity to meteorological variability and change. Winter snow accumulation is governed by
 synoptic-in other weather and storm track patterns more than surface energy balance conditions,
 at least in the Canadian Rocky Mountains (e.g., Shea and Marshall, 2007; Sinclair and Marshall,
 2009), so this is beyond the scope of the present study.

Sensitivity studies presented here extend the foundational work of Oerlemans and Fortuin (1992)
 and others, which has generally been done on glacier mass balance sensitivity to changes in
 temperature and precipitation. Our study is limited to summer mass balance at one location, but
 our results offer insight into the influence of different meteorological variables and energy fluxes,
 their year-to-year variability, and the role of isolated vs. collective forcings, feedbacks, and
 interactions on summer melt extent.

Results affirm the importance of temperature conditions such as a driving variable in wind speed
 and cloud cover. This is the implicit premise in temperature-index melt models, and they can be

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1392 tuned to work well at our site. We hesitate to recommend this though. Albedo feedbacks are crucial to include in assessments of glacier response to elimate change, amplified by numerous 1393 1394 temperature-related feedbacks in the melt season evolution. However, change, and are not 1395 physically represented in most variables that influence the surface energy balance have nontemperature-index models. Variations in humidity and their influence on melt are not negligible 1396 influences on, and all terms in the surface energy budget and summer melt extent. Temperature-1397 index methods of estimating melt neglect potentially important impacts from cloud cover, 1398 1399 humidity, and the seasonal albedo evolution. Caution is needed when applying these methods to 1400 future projections. contribute to the daily and interannual fluctuations in net energy. 1401

Our modelling approach for surface energy balance is well-suited to a distributed energy balance model, applying the perturbation approach to larger scales (e.g., mountain ranges). Climate models simulate all of the relevant meteorological fields, and both past reanalyses and future projections can be driven using the perturbation approach introduced here. Meteorological sensitivities under different climate regimes (e.g., maritime, polar, or tropical conditions) can also be explored using this framework, to help understand theregional differences in glacier sensitivity to climate variability and change in different regions.

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1557 Tables

Table 1. Mean monthly weather conditions \pm one standard deviation at Haig Glacier, Canadian1559Rocky Mountains, May to September 2002-2012. Data are from automatic weather station1560measurements at an elevation of 2660 m, in the upper ablation zone of the glacier.

1561					11		<u> </u>
1563	Month	<i>T</i> (°C)	h (%)	e_{v} (hPa)	$q_{v}\left(\mathrm{g/kg} ight)$	P (hPa)	v (m/s)
1564	May	-1.4 ± 1.1	73 ± 4	4.0 ± 0.4	3.4 ± 0.4	743.0 ± 2.4	2.8 ± 0.2
1565	June	2.6 ± 0.9	73 ± 6	5.5 ± 0.5	4.6 ± 0.4	748.1 ± 1.4	2.6 ± 0.2
1566	July	6.9 ± 1.4	62 ± 5	6.4 ± 0.4	5.3 ± 0.3	751.2 ± 1.6	2.8 ± 0.3
L567	August	5.9 ± 1.1	64 ± 7	6.1 ± 0.4	5.1 ± 0.4	750.8 ± 1.4	2.5 ± 0.2
1568 1569	Sept	2.1 ± 1.8	71 ± 10	5.0 ± 0.4	4.2 ± 0.3	748.4 ± 1.8	3.0 ± 0.4
.570	JJA	5.1 ± 0.8	67 ± 4	5.7 ± 0.4	4.8 ± 0.3	750.0 ± 1.1	2.6 ± 0.2
1571	MJJAS	3.2 ± 0.7	69 ± 4	5.3 ± 0.3	4.3 ± 0.3	748.3 ± 1.4	2.7 ± 0.2
572							

Table 2. Mean monthly surface energy balance terms \pm one standard deviation at Haig Glacier,1576Canadian Rocky Mountains, May to September 2002-2012. Radiation fluxes and albedo values1577are from automatic weather station measurements and the turbulent fluxes and subsurface heat1578conduction are modelled from the AWS data. Fluxes are in W m⁻² and melt totals are in m w.e.

Month	Q_{s}^{\downarrow}	α_s	${\mathcal Q}_{\scriptscriptstyle L}{}^\downarrow$	${{{\it Q}_L}^\uparrow}$	Q_H	Q_E	Q_C	Q_N	melt
May	249 ± 24	0.76 ± 0.04	258±12	299±4	7 ± 4	-11 ± 3	5±2	22±12	$0.20\pm$
June	237 ± 23	0.70 ± 0.05	276 ± 14	310 ± 2	17 ± 4	-5 ± 4	3 ± 1	56 ± 21	$0.45\pm$
July	240 ± 19	0.57 ± 0.06	275 ± 8	313 ± 1	38 ± 9	1 ± 5	1 ± 1	$109\pm\!27$	$0.88\pm$
August	205 ± 25	0.38 ± 0.07	273 ± 11	312 ± 1	32 ± 7	-1 ± 3	2 ± 1	123 ± 22	$0.99\pm$
Sept	140 ± 30	0.59 ± 0.09	$271\!\pm\!13$	306 ± 3	23 ± 12	-6 ± 3	3 ± 2	$42\pm\!21$	$0.34\pm$
JJA	227 ± 14	0.55 ± 0.06	275 ± 6	312±1	29 ± 3	-2 ± 3	2 ± 1	97 ± 19	2.32±
MJJAS	215 ± 17	0.60 ± 0.04	271 ± 7	$308{\pm}1$	23 ± 4	-4 ± 3	3 ± 1	71 ± 15	$2.86\pm$

Perturbation δm- (mm) σ	$-\delta Q_{S}^{\downarrow}$	δα	δQs^{net}	δQ_L^{\downarrow}	$-\delta Q_H$	$-\delta Q_E$	δQ_N	<u>δQ</u> _{Nσ}		
$\delta T = 1^{\circ}\mathrm{C}; \delta h = 0$	_0	0	-0	4.7	4.2	-3.5	-12.4	- <u>-2959.9</u>		
$\delta T = 1^{\circ} \text{C}; \ \delta q_{\nu} = \delta \tau = \delta \varepsilon_a = 0$	_0	0	-0	4.0	-4.2	-0	-8.3	<u> 1966.6</u>		
$\delta T = 1^{\circ}\mathrm{C}; \delta q_{v} = 0; \delta \tau, \delta \varepsilon_{a}$	-22.6	0	-10.2	-7.8	-4.2	-0	-6.6	- 156 5.3	_	Formatted: Font: 10 pt
$\delta q_{\nu} = 1 \text{ g kg}^{-1}; \ \delta \tau = \delta \varepsilon_a = 0$	-0	0	-0	0	-0	-10.5	-10.5	<u>-2483.2</u>		
$\delta q_v = 1 \text{ g kg}^{-1}; \delta \tau, \delta \varepsilon_a$	41.8	0		24.1	0	-10.5	15.7	<u> </u>	<	Formatted: Font: 2 pt
$\delta v = 1 \text{ m s}^{-1}$	-0	0	-0	0-	8.3	-1.4	6.9	<u>-1642.1</u>		Formatted: Font: 2 pt
50									\swarrow	Formatted: Font: 2 pt
$\delta Q_0 = 1 \text{ W m}^{-2}$	-0.6	0	-0.3	0	-0	-0	-0.3-	7		Formatted: Font: 2 pt Formatted: Font: 2 pt
$\delta \tau = \delta Q_S^{\downarrow} = 10 \text{ W m}^{-2}$		10.0-0	5			18	0	4.5 0-		
$\frac{003}{1000} -0 \frac{4.5}{1000}$	6.3 1	50	5			10	_0			Formatted: English (United States) Formatted: French (Canada)
$\delta Q_L = 10 \mathrm{W} \mathrm{m}^{-2}$	0	<u> </u>		0	10	0	0 1	0.0 6.0		
$\frac{143}{80}$ 1 102		26.0	0	1()	10.0	0	1.(0 2 2		
$\delta \tau = 0.1 - 192_{-76}$		36.0	0	16.2 -	-19.6	0 -	-4.0 -	-8.0 -3.2		
$\frac{1}{\delta \alpha_{\rm S}} = 0.1$	-0	0.1		0	-0	-0	-22.7			Formatted: Font: 3 pt
-539 -13.6 -323						-				
1σ, all (JJA)	14.0	-0.06	20.8	0.8	5.6	7.5		<u>—821</u>		
1 σ, all (MJJAS)	17.0	-0.04	16.0	5.5	4.6	7.3		<u> 1323</u>		

1647	1. $\delta T = \pm 2^{\circ}$ C; $\delta h = 0$; $\delta \alpha_S = 0$	$\partial Q_N / \partial T = -1013 \text{ W m}^{-2} (^{\circ}\text{C})^{-1}$	-+8+10 W m ⁻²		
1648	2. $\delta T = \pm 2^{\circ}C; \delta h = 0$	$\partial Q_N / \partial T = -5527 \text{ W m}^{-2} (^{\circ}\text{C})^{-1}$	+4421 W m ⁻²		
1649	3. $\delta T = \pm 2^{\circ} C$; $\delta q_v = \delta \tau = \delta \varepsilon_a = 0$	$\partial Q_N / \partial T = -5121 \text{ W m}^{-2} (^{\circ}\text{C})^{-1}$	$+41\underline{17}$ W m ⁻²		
1650	4. $\delta T = \pm 2^{\circ}$ C; $\delta q_v = 0$; $\delta \tau$, $\delta \varepsilon_a$	$\partial Q_N / \partial T = -4317 \text{ W m}^{-2} (^{\circ}\text{C})^{-1}$	+3413 W m ⁻²		
1651	5. $\delta q_v = \pm 50\%$; $\delta \tau$, $\delta \varepsilon_a = 0$	$\partial Q_N / \partial q_v = -9 \text{ Wm}^{-2} (g/kg)^{-1}$	$+3 \text{ W} \text{m}^{-2}$		
1652	$-6. \delta q_{\nu} = \pm 50\%; \delta \tau, \delta \varepsilon_{a}$	$\frac{\partial Q_N}{\partial q_v} = 1815 \text{ W m}^{-2} (\text{g/kg})^{-1}$	$+5 \text{ W} \text{m}^{-2}$		
1653	<u>6. $\delta q_y = \pm 50\%$; δτ, δε_a</u>	$\partial Q_N / \partial q_v = 25 \text{ W m}^{-2} (\text{g/kg})^{-1}$	$+8 \text{ W} \text{m}^{-2}$		
1654	7. $\delta v = \pm 50\%$	$\partial Q_N / \partial v = -814 \text{ W m}^{-2} (\text{m/s})^{-1}$	$+23 W m^{-2}$	/	Formatted: Font: 2 pt
1655	8. $\delta \tau = \pm 0.1$	$\partial Q_N / \partial \tau = -9 \mathrm{W} \mathrm{m}^{-2} (0.1)^{-1}$	$\underline{-} \frac{8}{8} \frac{\delta Q_{\text{S}}}{\delta = \pm 20}$		
1656	$W \cdot m^{-2} - \partial Q_N / \partial Q_S^{\downarrow} =$	$-0.4 \text{ W} \text{m}^{-2} (\text{W} \text{m}^{-2})^{-1} + 6 \text{W} \text{m}^{-2}$		_	Formatted: French (Canada)
1657	9. $\delta \alpha_s = \pm 0.1$	$\partial Q_N / \partial \alpha_S = -22227 \text{ W m}^{-2} (0.1)^{-1}$	-16 W m^{-2}		Formatted: Font: 12 pt
1658	$10. \delta b_{\underline{w}} = \pm 1 \mathrm{mw.e.}$	$\partial Q_N / \partial b_w = -12 \text{ Wm}^{-2} (\theta \cdot \mathbf{m} \cdot \mathbf{w} \cdot \mathbf{e} \cdot)^{-1}$	<u> </u>		Formatted: Font: 11 pt
1659 1660	$W m^{-2}$			\frown	Formatted: Font: 4 pt
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1666	Table 5. Summer surface energy b	alance fluxes on Haig Glacier as forced l	by the North American		
1667		nily weather fields, 1979-2014. NARR	2		
1668		l values. Melt is in m w.e., and all fluxes	1		
	r				

1669										
1670	Period	Qs^{\downarrow}	α_s	$Q_{\scriptscriptstyle L}{}^\downarrow$	${\mathcal{Q}_L}^\uparrow$	Q_H	Q_E	Q_C	Q_N	melt
1672 1673	JJA 0. <u>3242</u>	227 ± 7	$0.53\pm0.04\underline{05}$	275 ± 4	311 ± 1	27±4	-3±3	2±1	$95\pm\!14$	$2.28\pm$
1674 1675 1676 1676	MJJAS 0. <u>3850</u>	215 ± 6	$0.55 \pm 0.03 \underline{04}$	271 ± 4	308±2	22±3	-5±3	3±1	73 ± 10	2. 95<u>68</u>±
1679 1678										

Table 6. Correlation and sensitivity of different weather variables to the mean summer (JJA) net1681energy flux, Q_N , for the NARR simulations, 1979-2014. <u>'cloud' is the NARR total cloud fraction</u>.

Variable	Correlation	Sensitivity	δQ_N for $+1\sigma$
<i>T</i> (°C)	0.84	$\partial Q_N / \partial T = 14 \text{ W m}^{-2} (^{\circ}\text{C})^{-1}$	$+10 \text{ W} \text{m}^{-2}$
<u>-h (%)</u>	-0.33	$-\frac{\partial Q_N}{\partial h} = -2 \text{ W m}^{-2} (\%)^{-1}$	$-6 \text{ W} \text{m}^{-2}$
$q_v (\mathrm{gkg}^{-1})$	0.50	$\partial Q_N / \partial q_v = 25 \text{ W m}^{-2} (\text{g/kg})^{-1}$	$+7 \text{ W} \text{m}^{-2}$
$v ({\rm ms^{-1}})$	0. 07 <u>00</u>	$\partial Q_N / \partial v = -4 \text{ W m}^{-2} (\text{m/s})^{-1}$	$-1 \text{ W} \text{m}^{-2}$
Qs^{\downarrow} (W m ⁻²)	0.14	$\partial Q_N / \partial Q_S^{\downarrow} = 0.3 \text{ W m}^{-2} (\text{W m}^{-2})^{-1}$	$+2 \text{ W} \text{m}^{-2}$
Q_L^{\downarrow} (W m ⁻²)	0.64	$\partial Q_N / \partial Q_L^{\downarrow} = 2 \text{ W m}^{-2} (\text{W m}^{-2})^{-1}$	$+8 \text{ W} \text{m}^{-2}$
τ	0.25	$\partial Q_N / \partial \tau = 15 \text{ W m}^{-2} (0.1)^{-1}$	$+4 \text{ W} \text{m}^{-2}$
cloud	-0.19	$\partial O_N / \partial c = -8.1 \text{ W m}^{-2} 0.1)^{-1}$	$-3 \text{ W} \text{m}^{-2}$

1693 1694 1695 1696 1 698 1699	α_s b_{a^v} (m w.e.)	-0.83 -0. 05 15	$\partial Q_N / \partial \alpha_S = -26 \text{ W m}^{-2} (0.1)^{-1}$ $\partial Q_N / \partial b_w = -3 \text{ W m}^{-2} (\text{m w.e.})^{-1}$	-11 W m ⁻² -1 W m ⁻²		Formatted: Font: 12 pt, Superscript Formatted: Font: 6 pt Formatted: Subscript Formatted: Font: 12 pt, Superscript
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1700 Figures

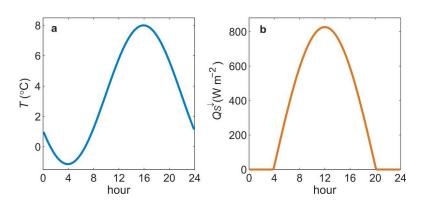
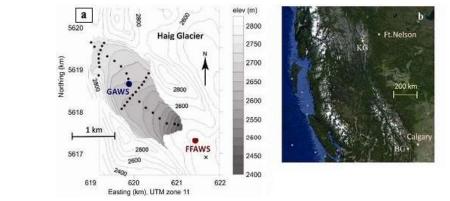


Figure 1. Idealized diurnal cycles of (a) temperature and (b) incoming shortwave radiation used
in the energy balance model. These two examples are for a sample day, July 1, 2010, parameterized
from daily minimum and maximum temperature in (a) and day of year plus mean daily incident
shortwave radiation in (b).





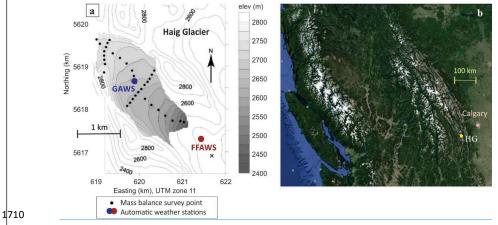
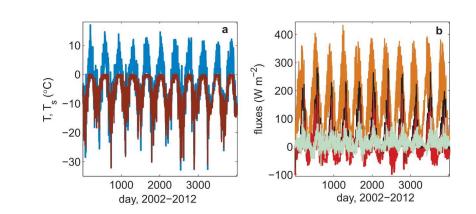


Figure 2. (a) The topography and automatic weather stations on Haig Glacier (GAWS) and the glacier forefield (FFAWS). The smaller black dots are mass balance survey points. (b) The location of Haig Glacier is labelled HG on the Google Earth map of Westernsouthwestern Canada.

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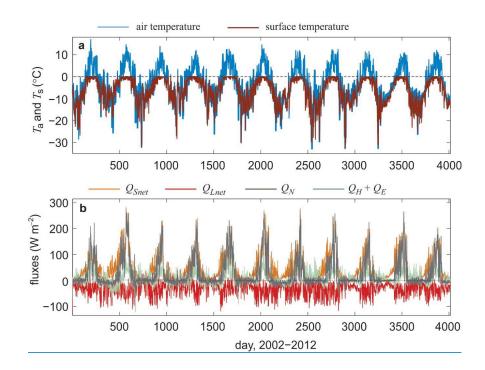
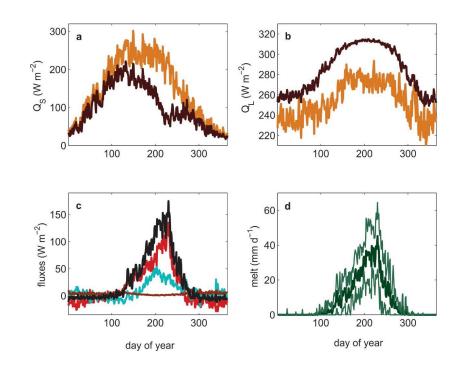
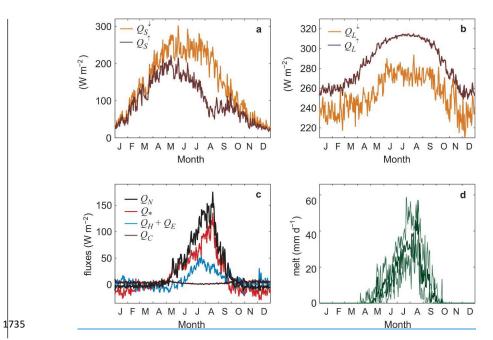
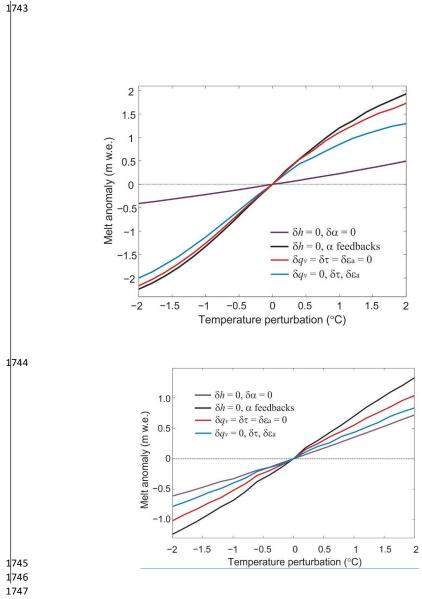


Figure 3. The 11-year record of (a) <u>air temperature, modelled surface temperature</u>, and (b) <u>surface</u> energy fluxes at the Haig Glacier AWS site. Daily mean values are plotted from Jan 1, 2002-Dec
31, 2012. (a) <u>Air temperature (blue) and modelled surface temperature (brown). (b) Incoming</u>
shortwave radiation (orange), net radiation (red), turbulent fluxes and Q_C (green), and net energy
(black).



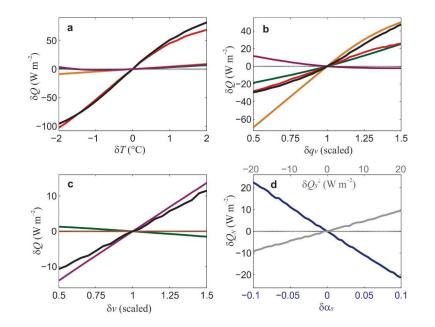


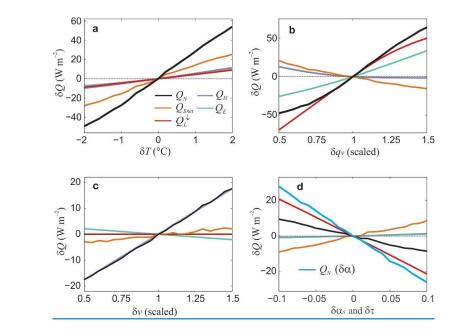
1736Figure 4. The average annual cycle of (a-c) surface energy fluxes and (d) daily melt at the Haig1737Glacier AWS. Daily mean values are plotted for the period 2002-2012. (a) Incoming (orange) and1738outgoing (brown) shortwave radiation. (b) Incoming (orange) and outgoing (brown) longwave1739radiation. (c) Net radiation (red), turbulent fluxes (green), Q_C (brown), and net energy (black). (d)1740MeltFor melt rates, mm w.e. per day. The the heavy line is the 11-year meanmedian value and the1741thin lines indicate the mean ± 1 standard deviation interquartile range.



- Figure 5. Sensitivity of modelled summer (JJA) melt to temperature perturbations for different assumptions, as per Table 4. Purple line: relative humidity is constant, no albedo feedbacks.

- Black line: relative humidity is constant, including albedo feedbacks. Red line: specific humidity
- 1750 1751 1752
- is constant, no atmospheric feedbacks. Blue line: specific humidity is constant, including atmospheric feedbacks. The reference The reference (mean 2002-2012) JJA melt is 2.32 m w.e.
- 1753





1756 1757 1758 1759 1760 1761 1762 Figure 6. Sensitivity of the surface energy balancefluxes at Haig Glacier to a-changes in (a) temperature, (case 2), (b) specific humidity, (case 6), (c) wind speed, (case 7), and (d) shortwave radiation (greyatmospheric transmittance (case 8) and albedo (blue). For plots (a)-(c), black lines indicate the net radiation, purple lines are the sensible heat flux, green lines are the latent heat flux, red lines are the net radiation, and orange lines are the incoming longwave radiation. line, case 9). 1763 All lines are anomalies relative to the baseline data from the period 2002-2012, and indicate the 1764 mean sensitivity of the different energy fluxes over this period. Please note the different y (δO) 1765 scales. 1766

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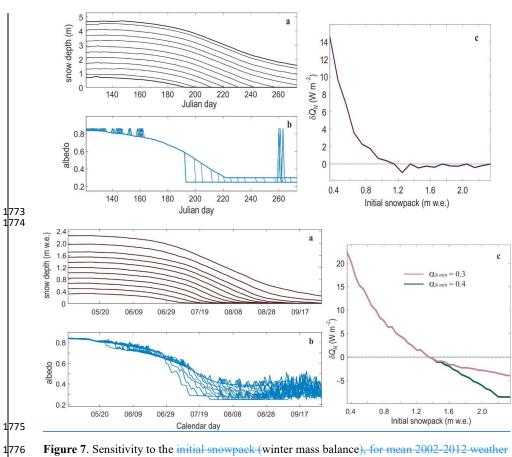
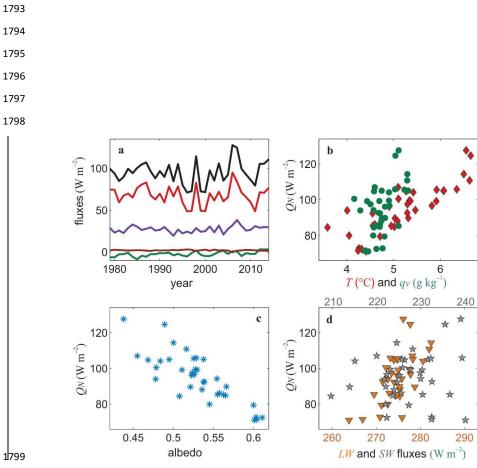


Figure 7. Sensitivity to the initial snowpack (winter mass balance), for mean 2002-2012 weather conditions but with, examined by varying May 1 snow varied_depth from 0.3536-2.3536 m w.e., relative to the reference value of 1.3536 m w.e. at the glacier AWS. (a) Snow depth (m) and (b) albedo through the summer melt season, May 1-Sept 30, for the different scenarios. Albedo
spikes correspond to summer initial snow eventsdepths. (c) Net summer (JJA) energy balance change as a function of the winter mass balance for two different settings of the minimum snow albedo.



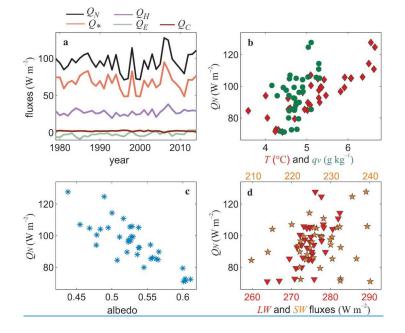


Figure 8. a) Mean summer (JJA) NARR-forced surface energy fluxes at Haig Glacier, 1979-2014.
 Black: net energy. Red: net radiation. Purple: sensible heat flux. Brown: conductive heat flux.
 Green: latent heat flux. (b) Mean summer net energy as a function of (b) temperature (red diamonds) and specific humidity (green circles). (c) Mean summer net energy as a function of albedor, and (d) Mean summer net energy as a function incoming shortwave radiation (grey stars) and incoming and longwave radiation (orange triangles). Table 6 gives the associated correlations.



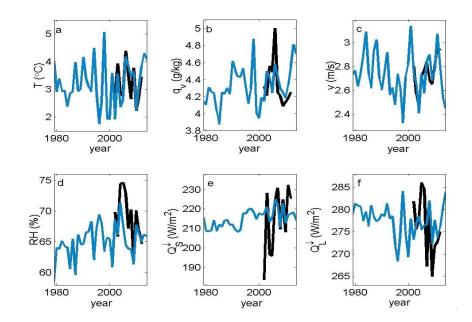


Figure 5. Mean melt season (MJJAS) weather conditions from the bias-corrected NARR output (blue), 1979-2014, and for the in situ data (black), 2002-2012: (a) temperature, (b) specific humidity, (c) wind speed, (d) relative humidity, (e) incoming shortwave radiation, and (f) incoming longwave radiation.

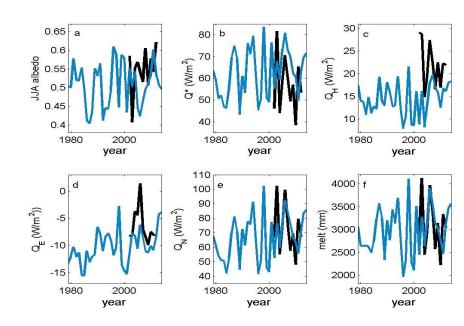


Figure 6. The evolution of modelled summer surface energy balance and melt from the perturbed NARR output (blue), 1979-2014, and from the in situ data (black), 2002-2012. (a) albedo, (b) net radiation, (c) sensible heat flux, (d) latent heat flux, (e) net energy, and (f) total summer melt (mm). All fields are for MJJAS except for albedo, which is shown for JJA, the

