Dear Dr. Matsuoka,

Thank you for reviewing our revised manuscript. We have adopted most of your suggested changes, as we describe below. Please let us know if you have any questions or concerns.

The revised version of the manuscript and the response letter is clear and well articulates the arguments that the authors have made. The both reviewers concerned the validity of sinusoidal fitting. The authors followed the suggestion of Dr. Gartner to use a bootstrapping technique and demonstrated that the best estimate of the sinusoidal approximation is robust. Also, the authors clarify in the response letter and the manuscript that it is likely that ice-flow variations are more complicated than the simple sinusoidal curve can capture. I am happy to see that these points are well presented. However, in my opinion, the authors do not adequately exclude a possibility that there is no seasonal variation of the flow speed at all. Thus, I would like to propose the authors to perform p tests with a null hypothesis that there is no seasonal variation of the flow speed. If this hypothesis is rejected with a very high confidence, it further supports the argument...

We have taken this suggestion and we now mention the correlation coefficient between observations and sinusoid fit along with the statistical significance of the correlation in the main text. The final paragraph of Section 2.2 now states,

...The resulting best-fit sinusoid is characterized by a 1601 m yr\(^{-1}\) mean velocity, an amplitude of 106 +/- 9 m yr\(^{-1}\), a maximum velocity on March 21 (1\(\sigma\) = 5 days), and a minimum velocity on September 19 (1\(\sigma\) = 5 days). The sinusoid provides a measure of periodicity at the 1 yr\(^{-1}\) frequency and matches observations to \(r = 0.472, p = 6 \times 10^{-33}\). A complete description of the sinusoid fit and a full uncertainty analysis are provided in Appendix A.

We have also added the following paragraph to the beginning of the uncertainty analysis section:

To verify the presence of seasonal variation in ice flow speed, we performed a p test using the null hypothesis that the amplitude of seasonal variability is 0 m yr\(^{-1}\). The 565 MODIS velocity measurements match the sinusoid fit by least squares with a Pearson correlation coefficient of \(r = 0.472\) and a corresponding \(t_{\text{statistic}} = 12.70\). The probability of the null hypothesis that there is no seasonal cycle is \(p = 6 \times 10^{-33}\), and thus we reject it in favor of the alternate hypothesis that cyclic seasonal behavior is present at TIS.

...In addition, I would request authors to include error estimates of individual flow-velocity measurements in Sections 2.1 and 2.2.

The short answer is that error estimates for individual flow velocity measurements are not directly available from template matching algorithms. For this reason, we deliberately designed our analysis to ensure that we do not to analyze individual
velocity measurements at any point in this manuscript. Our analyses and conclusions instead lean on the climatological sinusoid fit to the measurements. Error estimates of the sinusoid we analyze are quantified by bootstrap analysis, which gives a total estimate of error in the underlying measurements and the mismatch between true behavior and the sinusoidal approximation.

A brief description of the difficulty of obtaining error estimates for individual measurements follows.

In theory, template matching should be accurate to about a quarter of a pixel displacement, which is easy enough to convert to velocity uncertainty by dividing by the $dt$ between images. However, ice deformation, migrating sastrugi, snow accumulation, or partial cloud cover can complicate things quite a bit.

Part of the problem is that the algorithm might accurately track migrating sastrugi, whereas we are interested in the motion of the underlying ice. Quantifying the uncertainty in the difference between sastrugi migration and ice motion is not straightforward. In most cases, sastrugi is probably not the culprit of erroneous displacement measurements. Rather, uncertainty can result from indistinct peaks in the correlation or a high correlation peak that is surrounded by other highly correlated values. This figure from Scambos et al., 1992 depicts the concept of template matching:

![Template matching diagram](image)

The reference chip from the first image is chosen, and it is compared to the entire search area of the second image. Wherever the correlation index is the highest between the reference chip and the second image, that is assumed to correspond to the ice displacement. However, the cartoon above shows an ideal case where the correlation peak is quite distinct from its surroundings. Quite often, correlation values may be high, but the curve is nearly flat. Other times, correlations everywhere within the search area may be quite low, yet a small peak can provide
an accurate measure of displacement. Other times still, multiple correlation peaks can be seen within a given search area. Which peak is the correct one? How do you distinguish between a true peak versus a few pixels of noise? How broad of a correlation peak is too broad?

Because it is unclear what level of uncertainty results from these thresholding decisions in template matching, different data providers have gone about trying to quantify uncertainty by different means. The MEaSUREs annual velocity dataset (https://nsidc.org/data/nsidc-0720) averages several short-term displacement measurements collected each year, and presents two measures of uncertainty. They do provide estimated errors in the x and y directions, while cautioning that "these values should be used more as an indication of relative quality rather than absolute error." To give more insights into potential error, they also provide the standard deviations of x and y displacement and the total number of displacement measurements contributing to the average velocity calculated for each pixel. In other words, because the MEaSUREs annual velocity dataset is unable to provide uncertainty estimates for individual velocity measurements, they present only the statistics of the velocity measurements they obtain for each year. In our present manuscript, we are looking at how ice velocity varies within each year, so such values are useless.

The GoLIVE dataset (https://nsidc.org/data/NSIDC-0710/versions/1) we use provides individual displacement measurements, so unlike MEaSUREs, it cannot provide statistical measures of uncertainty. Thus, for each measurement, for each pixel, GoLIVE provides the value of the peak correlation, quantitative measures of the curvature of the correlation curves in the x and y directions which may be used to determine how sharp the correlation index peak is in each direction, and the value of the second-largest peak in correlation. GoLIVE then leaves it up to the users to decide, given these metrics, what type of quality threshold may be appropriate for any given work. No other metric of error is provided in the GoLIVE dataset.

Our MODIS velocity measurements were generated with a template matching algorithm that is similar to the one used by GoLIVE, so our quality metrics are similar to theirs and we have no direct measure of uncertainty for any individual measurements.

For the reasons described above, template matching is characteristically noisy, which we have dealt with by averaging the GoLIVE and MODIS velocity measurements over large regions of the ice shelf, with the idea that noise tends to cancel itself while the signal remains strong (as in Greene et al., 2017 and Greene & Blankenship, 2018). This averaging approach might suggest that we could get an idea of uncertainty by assessing the statistics of the displacement measurements within the averaging region for any given image pair. However, that approach would assume that errors from one pixel to the next are uncorrelated, which is almost certainly untrue. While some uncorrelated noise varies from pixel to pixel, reference chips and search areas partially overlap, clouds or other physical effects can
influence arbitrary regions of the averaging area, and geolocation errors in the underlying images affect entire measurements to an unknown degree. Without a clear way to assess the spatial correlation of errors, we cannot responsibly quantify the uncertainty for individual measurements. Thus, we analyze only the climatological sinusoid, whose uncertainties are well constrained.

The authors presented the uncertainty quantification in Appendix A to keep the manuscript structure as close as possible to the original form, which was well received by the reviewers. However, depending on the outcome of the proposed tests above, please consider reorganizing the manuscript and present these uncertainty quantifications together at the end of Section 2.2.

In the main text we show the presence of seasonal variability in two independent datasets, in two separate locations on the ice shelf. In the Appendix, the bootstrapping analysis we implemented in response to Dr. Gardner's suggestion effectively shows that the chance of no seasonal variability (amplitude 0 m yr\(^{-1}\)) lies 12-σ below the observed amplitude (106 m yr\(^{-1}\)) of the least squares fit. We also describe a p test in which we show that the possibility of no seasonal variability can be rejected at the significance level of p = 6x10\(^{-33}\). We feel that the multiple lines of evidence we present in the main text and in the Appendix are more than sufficient to convince readers that seasonal variability exists at Totten, and inserting multiple paragraphs into the main text as proof would be an unnecessary distraction from the physics at hand. We have contained most of the uncertainty analysis to the Appendix in this revision, but we are happy to move it into the main text if the editor feels it would belong better there.

Below, you find some very minor suggestions for improvement.

- Fig. 1: the calving front is also indicated by black, so revise the caption accordingly....

We have revised the caption, which now reads,

*The calving front and grounding line are shown in black, including a grounded ice rumple near the center of the ice shelf.*

...Also, revise the second sentence in the caption to “Towards the ice calving front right bottom of the image...” so the shape and ice flow direction of TIS is clear for all readers...

We have changed the text to the suggested wording.

...The label for the colorbar should be revised to “.... (m yr-1)” from “(m/a)”...

We have changed the colorbar label as requested.
...Strictly speaking, velocity is a vector and speed is a scholar. I feel that the authors tend to describe both velocity and speed as velocity, which is acceptable for me. However, then please keep the terminology simple and uniform; consider changing "speed" in the colorbar label and elsewhere to velocity or at least check the clarity of the entire text one more time.

We, too, lament this unfortunate convention in the glaciological community. We have nonetheless opted for convention rather than correctness, and we have changed the colorbar label to 'velocity' as suggested.

- Fig. 2: Unit of the ordinate should be “m yr-1”, not “m/yr”. And consider plotting the speed in the unit of km/a (to be more compatible with Figs. 3 and 7).

We have implemented both changes as requested.

- P6L6-7 and P8L10-11: do you indicate the plus/minus one sigma (i.e. 2 sigma) or only one sigma? Remove the plus/minus signs from the sigma, or add this sign to the right hand side of the embedded equation as well.

Ah, yes, we understand the ambiguity now that it has been pointed out. We have followed the suggestion by removing the +/- symbol.

- Fig. 5 caption: As far as I can read from the figure, I am not convinced that melt anomalies propagate in a clockwise fashion around the cavity. If this point is not important then please delete it. If important, revise the figure or explain more clearly.

This section focuses on the magnitude, timing, and spatial distribution of basal melt at TIS, and Figure 5 shows all three. The phase of the melt rate is at the heart of our evidence for eliminating basal melt from the list of potential causes of the seasonal velocity variability we observe. The direction of circulation is not integral to our conclusions per se, but we nonetheless describe the circulation in a few brief words to provide context for understanding how the melt rate signal propagates around the ice shelf cavity. The wording is brief enough that uninterested readers can skip by it without losing the main story, but descriptive enough that curious readers can explore the figure and see the pattern themselves.

Melt rate anomalies tend to initiate at the lower left hand corner of the ice shelf, where the figure shows a zero-day phase anomaly. Following the grounding line clockwise up the left side of the ice shelf, we pass the 10 day lag contour, then the 20 day lag contour, then the 30 day lag contour. Longer lag times extend from the deepest part of the grounding line toward the ice shelf front along the right side of the figure because anomalies reach the "three o'clock" position of the ice shelf only after passing the "nine o'clock" position. Near the circulation's exit on the lower right hand side where the ice shelf front abuts Law Dome, we see a 40 day contour, which is the longest lag contour and completes the circulation regime.
The pattern described above is shown in Fig 5, and it describes what we have observed in poring over decades of high-resolution melt rate data. It is also in keeping with the reports of clockwise circulation described in the Gwyther et al. 2014 reference we cite. Of course, the model attempts to replicate the complexities of reality, so the contour map we present in Fig 5 is by no means an idealized picture of the circulation. Nonetheless, the contour map presents the phase of the melt rate as elegantly as we know how, and the physical process it depicts is described in the caption, which reads,

...Gray contours show melt rate lag times in days relative to anomalies at the ice front, indicating melt anomalies propagate in a clockwise fashion around the cavity...

We believe that this statement is true, it is meaningful to anyone who wonders if the nonuniform timing of melt rate anomalies throughout the cavity might affect our results, and it is brief enough that readers can understand it and move on without distraction. We feel that figure captions should do more than simply identify the markers that are used in a figure, and should instead be used to help viewers understand the physical processes and important relationships depicted in the figure. We believe the wording above accomplishes this goal, and we prefer to keep the caption as it is.

- Spell out ECCO as “Estimating the Circulation and Climate of the Ocean” at least in the reference list of Fukumori et al.

We have included the full definition of ECCO where it first appears in the section describing the basal melt rate model.

- Fig. 6: add goldern ECCO location marker in the two most right satellite images taken in April and May 2016.

We have clarified the reason for the missing golden markers by stating in the figure caption that they are not shown for 2016, when no ECCO v4-r3 data are available.
Seasonal dynamics of Totten Ice Shelf controlled by sea ice buttressing

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Abstract. Previous studies of Totten Ice Shelf have employed surface velocity measurements to estimate its mass balance and understand its sensitivities to interannual changes in climate forcing. However, displacement measurements acquired over timescales of days to weeks may not accurately characterize long-term flow rates where ice velocity fluctuates with the seasons. Quantifying annual mass budgets or analyzing interannual changes in ice velocity requires knowing when and where observations of glacier velocity could be aliased by subannual variability. Here, we analyze 16 years of velocity data for Totten Ice Shelf, which we generate at subannual resolution by applying feature tracking algorithms to several hundred satellite image pairs. We identify a seasonal cycle characterized by a spring to autumn speedup of more than 100 m yr\textsuperscript{−1} close to the ice front. The amplitude of the seasonal cycle diminishes with distance from the open ocean, suggesting the presence of a resistive backstress at the ice front that is strongest in winter. Springtime acceleration precedes summer surface melt and is not attributable to thinning from basal melt. We attribute the onset of ice shelf acceleration each spring to the loss of buttressing from the breakup of seasonal landfast sea ice.

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1 Introduction

Totten Glacier in East Antarctica drains the Aurora Subglacial Basin, which is grounded well below sea level (Young et al., 2011; Roberts et al., 2011) and contains enough ice to raise the global sea level by at least 3.5 m (Greenbaum et al., 2015). Short-term observations have identified Totten Glacier and its ice shelf (TIS) as thinning rapidly (Pritchard et al., 2009, 2012) and losing mass (Chen et al., 2009), but longer-term observations paint a more complex picture of interannual variability marked by multi-year periods of ice thickening, thinning, acceleration, and slowdown (Paolo et al., 2015; Li et al., 2016; Roberts et al., 2017; Greene et al., 2017a). The current best estimates of Totten Glacier and TIS mass budgets have been calculated using a mosaic of surface velocity measurements collected at different times throughout the year (Rignot et al., 2013); however,
such estimates have been built on an unconfirmed assumption that ice velocity does not vary on subannual timescales. Where glacier flow varies throughout the year, it is possible that velocity measurements collected over short time intervals may lead to inaccurate estimates of annual mass balance or incorrect interpretation of interannual changes in velocity. Furthermore, most common methods of ice velocity measurement, such as satellite image feature tracking or in-situ GPS measurements taken over the course of a field season, are strongly biased toward summer acquisition and may not accurately represent winter ice dynamics. Wherever seasonal velocity variability exists, it is important to consider how ice velocity is measured, and how the measurements can be interpreted.

Seasonal variations in glacier velocity have been observed in Greenland and Antarctica (e.g., Joughin et al., 2008; Nakamura et al., 2010; Moon et al., 2014; Zhou et al., 2014; Fahnestock et al., 2016), and have been attributed to a number of different mechanisms. On grounded ice, surface meltwater can drain into crevasses or moulins, make its way to the bed, pressurize inefficient subglacial hydraulic systems, and allow the glacier to speed up until pressure is reduced (Sohn et al., 1998; Bartholomew et al., 2010; Moon et al., 2014). On floating ice, surface meltwater may also influence ice shelf velocity by percolating through and weakening the ice shelf shear margins (Liu and Miller, 1979; Vaughan and Doake, 1996; Cavanagh et al., 2017). Observations have shown correspondence between seasonal advance and retreat of marine-terminating glaciers and the presence of ice mélange at the glacier terminus (Howat et al., 2010; Cassotto et al., 2015; Moon et al., 2015). The exact mechanisms by which ice mélange can affect glacier dynamics are poorly understood, but modeling studies have shown that the back stress provided by sea ice can prevent calved icebergs from rotating away from the ice front (Amundson et al., 2010), and in some cases can shut down calving entirely (Robel, 2017) causing an appreciable effect on glacier velocity (Todd and Christoffersen, 2014; Krug et al., 2015). For example, the buttressing strength of ice mélange at Store Glacier in Greenland has been estimated at 30–60 kPa, which is an order of magnitude below the driving stress of the glacier, but is sufficient to cause observable subannual changes in glacier velocity up to 16 km from the ice front (Walter et al., 2012; Todd and Christoffersen, 2014).

In Antarctica, marine ice is known to strengthen the Brunt and Stancomb-Willis ice shelf system (Hulbe et al., 2005), and an ice shelf acceleration event observed there in the 1970s has been attributed to a reduction in stiffness of the ice mélange that connects the two ice shelves (Khazendar et al., 2009). Similarly, multi-year landfast sea ice is strongly mechanically coupled to Mertz Glacier Tongue (Massom et al., 2010) and may have delayed a major calving event that occurred there in 2010 (Massom et al., 2015). Closer to TIS, two recent major calving events in Porpoise Bay (76°S, 128°E) were attributed to the breakup of landfast sea ice at the ice shelf termini (Miles et al., 2017), and on the Antarctic Peninsula it has been shown that sea ice can protect ice shelves from fracture induced by ocean swell (Massom et al., 2018). At TIS, long-term changes in calving front position have been reported with a possible connection to local sea ice processes (Miles et al., 2016), but corresponding links to glacier dynamics have not previously been investigated. To our knowledge, there have been no reports of seasonal variability of TIS or any of the mechanisms that may drive TIS variability at subannual timescales. In this paper we find seasonal variability in two independent ice velocity datasets and we consider the potential roles of surface melt water, ice shelf basal melt, and sea ice buttressing, in influencing the flow of TIS at subannual timescales.
2 Surface velocity observations

We analyzed surface velocity time series using feature tracking algorithms applied to Landsat 8 and MODIS (MODerate-resolution Imaging Spectroradiometer) images. Each image dataset was processed separately, using different feature tracking programs, and the resulting time series represent two independent measures of TIS velocity. The 15 m resolution of Landsat 8 permits precise displacement measurements over short time intervals, but the relatively brief four-year Landsat 8 record and limited number of cloud-free images inhibits our ability to separate interannual velocity changes from seasonal variability. The MODIS record contains many cloud-free images per year from 2001 to present; however, the 250 m spatial resolution of MODIS images limits measurement precision where ice displacements are small between images. Thus, the two image datasets each offer incomplete, but complementary insights into the seasonal dynamics of TIS. Processing methods for each dataset are described below.

2.1 GoLIVE (Landsat 8) velocities

We used the Global Land Ice Velocity Extraction from Landsat 8 (GoLIVE) dataset (Scambos et al., 2016; Fahnestock et al., 2016), which is processed at 600 m resolution for most of Antarctica. We analyzed the high-confidence vx_masked and vy_masked velocity fields from late 2013 to early 2018 and limited the dataset to 143 image pairs separated by $16 \leq dt \leq 112$ days. Many of the image pairs overlap in time, providing several redundant, semi-independent velocity measurements, particularly throughout the summer months when each image may contribute to multiple image pairs.

To understand the spatial pattern of TIS seasonality, we developed characteristic velocity maps for spring and autumn separately. Spring velocity was taken as the mean of 76 velocity measurements whose image pairs were obtained between June 16 and December 15. Autumn velocity was taken as the mean of 67 velocity fields from images obtained during the remainder of the year. We discard all pixels where the mean ice speed is less than 250 m yr$^{-1}$. We also discard all pixels containing fewer than 10 high-confidence spring or autumn velocity measurements. The resulting difference between spring and autumn velocities is shown in Fig. 1.

The terminal $\sim$50 km of TIS accelerates each year from spring to autumn, then slows during the winter. Seasonality is strongest close to the glacier terminus and decays with distance from the open ocean. The relatively featureless nature of the inner TIS surface limits the number of high-confidence matches in that region of the ice shelf, but the available measurements indicate minimal seasonality upstream of the mid-shelf ice rumple identified by InSAR (Mouginot et al., 2017b). The grounded ice of the eastern tributary accelerates slightly throughout the summer, while some grounded ice of the western tributary exhibits a weak slowdown.

To assess the timing of the annual TIS acceleration, we generate a velocity time series for a region of TIS near the terminus shown in Fig. 1. We populate a velocity time series from the means of all GoLIVE velocity measurements within 5 km to 10 km from the ice front, considering only pixels with a mean velocity exceeding 1700 m yr$^{-1}$. The resulting TIS velocity time series is shown in Fig. 2.
**Figure 1. Summer ice shelf acceleration.** Toward the ice calving front (right bottom of the image), autumn velocity exceeds spring velocity by more than 100 m yr$^{-1}$. This image shows the difference between the means of 76 spring and 67 autumn GoLIVE velocity fields. Dark green vectors indicate the mean velocity, supplemented by MEaSUREs InSAR-derived velocity (Rignot et al., 2011) outside the range of Landsat path 102, row 107. Green and pink polygons indicate the bounds of velocity averaging for the velocity time series shown in Fig. 2 and Fig. 3, respectively. A gold marker shows the location of the ECCO sea ice thickness time series described in Sec. 5. The black line indicates the calving front and grounding line are shown in black, and includes a grounded ice rumple near the center of the ice shelf (Mouginot et al., 2017b).

The short record and low temporal resolution of the GoLIVE dataset make it difficult to identify the exact timing of the onset of acceleration in any given year, but a linear trend fit to all available measurements indicates a typical acceleration of 0.8 m yr$^{-1}$ per day from late September to early April. Further investigation into the timing of accelerations each year requires a more complete time series of TIS velocity, which we generate from MODIS images.
Figure 2. Ice front acceleration from GoLIVE (Landsat 8). The GoLIVE dataset contains many overlapping TIS velocity measurements captured between September and April of each year. The velocities here represent all displacements measured over \(16 \leq dt \leq 112\) days (indicated by gray lines), averaged within the green polygon in Fig. 1. The red trend line is a linear least-squares fit to the observations, indicating a typical spring-to-fall acceleration of \(0.8 \text{ m yr}^{-1}\) per day.

2.2 MODIS velocities

A MODIS velocity time series was generated from 672 pairs of cloud-free MODIS band 2 images (Scambos et al., 2001, updated 2018; Greene and Blankenship, 2018) acquired between 2002 and 2018. Each image pair was separated by 92 to 182 days and was processed at 250 m resolution using the ImGRAFT template matching software (Messerli and Grinsted, 2015) with Antarctic Mapping Tools for MATLAB (Greene et al., 2017b). Similar to the method described by Greene et al. (2017a), we used 2.5 km square templates with 4.0 km search boxes centered on locations predicted by InSAR-derived velocities (Rignot et al., 2017). To generate the MODIS velocity time series we averaged velocities from all pixels within 10 km to 30 km from the ice front, bounded on each side by the glacier shear margins identified by Greene et al. (2017a). We discarded any image pairs for which fewer than 99% of the pixels within the polygon contained valid displacement measurements, resulting in 565 valid MODIS velocity measurements in the time series. The polygon used for the MODIS time series is shown in Fig. 1.

Despite having measurements from dozens of MODIS image pairs most years, subannual template matching applied to 250 m resolution MODIS images produces such noisy velocity estimates that the timing of springtime acceleration cannot be accurately determined for any given year. However, by combining data from all years we can assess the characteristic cycle of ice shelf acceleration and slowdown that occurs throughout the typical year. Figure 3 shows the MODIS velocity time series overlaid on the mean seasonal cycle. Because no visible-band MODIS images are available during the dark winter months, no
Figure 3. Seasonal cycle of ice shelf velocity from MODIS. TIS velocity measurements from 565 MODIS image pairs separated by 92 to 182 days, averaged within the pink polygon shown in Fig. 1. Velocity measurements are shown at the mean of the acquisition times of their MODIS image pairs. The average seasonal cycle is shown approximated as a sinusoid, with 95% confidence intervals shaded.

Image pairs separated by 92 to 182 days are centered on any days in April, May, August, or September. However, 46 image pairs span the winter, providing velocity measurements centered on June and July.

We approximate the seasonal cycle of the TIS velocity as a sinusoid obtained by least squares fit to the 565 MODIS velocity measurements. To minimize the influence of interannual variability, the one-year moving average was removed before analyzing the seasonal cycle. The resulting best-fit sinusoid is characterized by a 1601 m yr$^{-1}$ mean velocity, an amplitude of 106±9 m yr$^{-1}$, a maximum velocity on March 21 ($\pm 1\sigma$ $=\pm 5$ days), and a minimum velocity on September 19 ($\pm 1\sigma$ $=\pm 5$ days). The sinusoid provides a measure of periodicity at the 1 yr$^{-1}$ frequency and matches observations to a root mean square error of 93 m yr$^{-1}$. Uncertainty analysis $r = 0.472 (p = 6 \times 10^{-33})$. A complete description of the sinusoid fit is explored and a full uncertainty analysis are provided in Appendix A.
3 Surface melt observations

Surface melt has been shown to affect the flow of grounded ice in Greenland when surface water drains through moulins or crevasses to the bed, where it alters basal water pressure and allows the overlying ice to accelerate (Zwally et al., 2002; Schoof, 2010; Bartholomew et al., 2010; Andrews et al., 2014). The seasonal velocity anomalies we observe at TIS are strongest near the floating ice front, so it is unlikely that the seasonal variability of TIS velocity is driven by subglacial hydrology on nearby grounded ice. However, the presence of englacial liquid water can weaken ice (Liu and Miller, 1979), and it is plausible that surface melt at TIS could percolate into the ice, weaken shear margins, and allow TIS to speed up as a result of reduced buttressing.

To assess the possible link between surface melt and TIS velocity anomalies, we used daily observations of surface melt from passive microwave radiometers (SMMR and SSM/I) gridded to 25 km resolution (Picard and Fily, 2006). We limited the period of analysis to 2000 through 2017 to roughly coincide with available MODIS image data. Figure 4 shows the spatial distribution of mean annual surface melt during this period. Using the masks developed by Mouginot et al. (2017b) with the Antarctic Mapping Tools for MATLAB dist2mask function (Greene et al., 2017b), we define three subdomains for surface melt analysis as

1. outer TIS: the floating portion of the ice shelf up to 50 km from the ice shelf front,

2. inner TIS: the floating portion of the ice shelf more than 50 km from the ice shelf front, and

3. grounded: all grounded ice within 50 km of the TIS grounding line.

Surface melt is most prevalent in the outer TIS, where in some locations surface melt is detected up to 16 days per year. Fewer surface melt days occur far from the ice front on the inner TIS, and surface melt is least common on the high-elevation grounded ice surrounding TIS. Figure 4 shows that although the number of annual surface melt days varies throughout the region, the timing of surface melt is roughly the same in all three subdomains, with the typical melt season lasting from December to February. For the outer TIS, the onset of surface melt typically occurs on December 23 (±1σ = 12–1σ = 12 days), with the earliest summer melt recorded on December 6, 2006. The mean final day of surface melt occurs on January 23 (±1σ = 9–1σ = 9 days), but has been observed as late as February 11 in 2005.

4 Modelled ice shelf basal melt

On interannual timescales, TIS is known to accelerate in response to prolonged periods of elevated basal melt rates (Roberts et al., 2017; Greene et al., 2017a), and a similar process has been observed at Pine Island Ice Shelf in West Antarctica (Christianson et al., 2016). For these laterally-bounded ice shelves restrained largely by shear stress at their margins, thinning reduces resistance to glacier flow and allows ice shelf acceleration.

To assess whether TIS may dynamically respond to basal melt anomalies at subannual timescales, we used the Regional Ocean Modeling System (ROMS; Shchepetkin and McWilliams, 2005) to simulate TIS-ocean interactions, then considered
the effects that subannual basal melt anomalies could have on TIS velocity. The model domain extended from 104.5°E–130°E in longitude and 60°S–68°S in latitude with a horizontal resolution of approximately 2 to 3 km. A terrain-following vertical coordinate provided enhanced resolution close to the seafloor and ice shelf interface. Modifications to the code allowed thermodynamic interaction between ocean and steady-state ice shelves, following Dinniman et al. (2003) and Galton-Fenzi et al. (2012). Seafloor bathymetry for the deep ocean and continental shelf was taken from the RTopo-1 dataset (Timmermann et al., 2010). As RTopo-1 does not contain the cavity of TIS, we inferred the cavity geometry. Ice basal draft for the TIS cavity was obtained from ICESat-derived surface elevations, assuming hydrostatic equilibrium and a mean ice density of 905 kg m$^{-3}$ (following Fricker et al., 2001). Water column thickness was obtained by linearly interpolating from 0 m thick along the grounding line to 300 m thick along the central flow line of the ice shelf (see Gwyther et al., 2014, for details). The model lateral and surface boundaries were forced over the hindcast period 1992–2012. Lateral forcing was derived from the
Figure 5. Modeled basal melt. (a) Mean melt rate distribution shows melt focused where the ice shelf base is deepest, exceeding 80 m yr\(^{-1}\) near the grounding line of the inner TIS. Gray contours show melt rate lag times in days relative to anomalies at the ice front, indicating melt anomalies propagate in a clockwise fashion around the cavity. (b) Two years of 1992–2012 climatological melt rates averaged within the subdomains in (a). (c) Ice thickness anomalies corresponding to the time integral of melt rate anomalies. (d) Ice velocity anomalies expected to result from seasonal variations in ice shelf thickness.

Estimating the Circulation and Climate of the Ocean Phase II (ECCO2) cube92 reanalysis solution (Menemenlis et al., 2008); surface forcing was ERA-interim wind stress (Dee et al., 2011), and heat and salt fluxes were derived from Special Sensor Microwave/Imager (SSM/I) algorithms for sea-ice production (Tamura et al., 2016). The model was spun-up for 21 model years using 1992–2012 forcing. After spin-up, the 1992–2012 forcing was repeated and we analyzed the mean seasonal cycle of melt from the second run. The mean spatial distribution and temporal variability of basal melt are shown in Fig. 5.

The distribution of basal melt at TIS mimics observations of other ice shelves, with the highest melt rates focused near the deep grounding line (e.g., Dutrieux et al., 2013). Seasonal variability is most significant in the inner TIS, where mean melt rates are highest, whereas the shallow ice base of the outer TIS experiences only a weak seasonal cycle superimposed on a low mean melt rate. Everywhere beneath the ice shelf, basal melt rate reaches a maximum in autumn and a minimum in the spring.

Ice shelf thinning tends to reduce buttressing and allow ice shelf acceleration. Using a simple model developed by Greene et al. (2017a) (adapted from Joughin et al., 2004) to estimate velocity anomalies resulting from seasonal changes in ice thickness, we find that on subannual timescales, basal melt anomalies should only affect TIS velocity on the order of 1 m yr\(^{-1}\) (Fig. 5). Note that velocity predictions are negatively correlated with ice thickness, which is calculated from the time integral.
of basal melt rate anomalies. Accordingly, velocity maxima related to basal melt do not correspond directly to basal melt rate maxima, but should occur at the end of the high-melt season in July, when ice thickness reaches a minimum.

Small perturbations in ice shelf thickness have the greatest influence on ice shelf buttressing where the ice shelf is thin. However, the thick ice of the inner TIS experiences much more seasonal melt variability than the thin ice of the outer TIS, so it is somewhat by coincidence that the large (> 8 m yr\(^{-1}\)) increase in melt rate from spring to autumn beneath the thick (> 2000 m) ice of the inner TIS, affects local ice velocity to approximately the same degree as the much smaller (∼ 3 m yr\(^{-1}\)) seasonal melt rate variability in the outer TIS, where ice is much thinner (Fig. 5).

The model we use to estimate melt-induced velocity anomalies assumes TIS velocity is limited only by lateral shear stress at the ice shelf margins and velocity anomalies are purely a function of local ice thickness. These assumptions vastly oversimplify the complex stress regime of the TIS, but are used to obtain an order-of-magnitude approximation of how the TIS should respond to seasonal variability of ice thickness driven by basal melt. From this simple model it is clear that the <1 m yr\(^{-1}\) variability expected to result from seasonal basal melt anomalies cannot explain the observed >100 m yr\(^{-1}\) seasonal variability of TIS velocity. Moreover, holding other factors are constant, the seasonal cycle of basal melt produces an ice shelf that grows throughout the summer and reaches a maximum thickness in February. Accordingly, basal melt anomalies should result in a summer slowdown and a velocity minimum in February, when observations show TIS nearing its velocity maximum. Thus, it is unlikely that the seasonal cycle of basal melt could explain the observed pattern of spring-to-fall acceleration of TIS.

5 Sea ice concentration and thickness

To assess whether the presence of sea ice may influence the flow of TIS, we analyzed observational data from microwave, thermal, and visual band satellite sensors, along with model data of sea ice thickness near the TIS front. We used daily observations of sea ice concentration (Cavalieri et al., 1996) and generated a time series given by the mean of three 25 km grid cells located close to the TIS front (shown in Fig. 6). In addition to ice concentration observations, we also analyzed daily effective sea ice thickness from ECCO v4-r3 for the period 2000–2015 (Fukumori et al., 2017). We focused on the time series of sea ice thickness for the grid cell centered on (66.47° S, 116.50° E), indicated by gold markers in Figs. 1 and 6. To fully understand the spatial and temporal variability of sea ice, we also inspected 315 cloud-free MODIS visual (band 2) and 164 thermal (band 32) images acquired throughout the year by the Aqua and Terra platforms between 2000 and 2017 (Scambos et al., 2001, updated 2018; Greene and Blankenship, 2018).

Figure 6 shows the seasonal cycle of sea ice growth and decay. The minimum ice concentration typically occurs at the TIS front in mid March, followed by increasing ice concentration throughout autumn as air temperatures decline (Dee et al., 2011). Inspection of visual and thermal band images reveals that sea ice consolidates and fastens to the western TIS in early to mid May. The rigid connection of landfast ice to the TIS front holds throughout the winter, with the exception of a small polynya abutting Law Dome that briefly opened in July 2016 (see Alley et al., 2016). Regardless of polynya activity, the majority of landfast ice remains connected to the TIS front, and each year the landfast connection breaks in October or early November, followed by a visible reduction in sea ice cover that occurs throughout November. In some years, sea ice concentration continues
Figure 6. Seasonal cycle of sea ice at the TIS front. MODIS visual band imagery (Scambos et al., 2001, updated 2018), remotely sensed ice concentration (Cavalieri et al., 1996), and ECCO v4-r3 effective sea ice thickness (Fukumori et al., 2017) reveal a seasonal cycle of sea ice growth and decay beginning around March 12 each year, when sea ice concentration is at a minimum. Ice concentration and thickness time series are shown with corresponding shading indicating daily values of ±1σ. Light blue background shading indicates the presence of fast ice at the TIS front, from about May 3 to October 8. During this winter period, remotely sensed ice concentration values remain relatively constant despite continued growth of sea ice. Average sea ice thickness steadily declines throughout the summer, while thin, unconsolidated sea ice often temporarily fills the area and is detected by remote sensors. Five example MODIS images (Scambos et al., 2001, updated 2018) show sea ice fastened to half of the TIS front in May and September, with dashed quadrangles indicating the region of ice concentration averaging and a gold marker denotes the location of the ECCO sea ice thickness time series (not shown for 2016, when no ECCO v4-r3 data are available).

... to decline throughout the summer, but more commonly, the region temporarily fills with unconsolidated ice, causing sea ice concentration to peak in January (Greene, 2017). From January to March, sea ice melts or is exported away from the TIS front until concentration reaches a minimum in mid March, then the cycle repeats. Sea ice thickness data are more well behaved, generally waxing and waning monotonically between a minimum in late March and maximum in late September. In this way, sea ice thickness follows the broader climatology reported by Fraser et al. (2012), who found that landfast ice between 90°E and 160°E grows from a minimum extent in March to a maximum in late September or early October.

We suspect that sea ice concentration is a poor measure of ice strength, because the simple fraction of a grid cell’s surface area covered by ice offers no indication of ice thickness or level of consolidation. This is seen not only in the summer melt season during which sea ice concentration often increases, but also in winter, when ice concentration observations remain constant while the ice grows steadily thicker. We posit that sea ice thickness is a better proxy for sea ice strength because the
Figure 7. Causes of springtime acceleration. Time series of MODIS-derived ice velocity ±2σ shaded, ECCO v4-r3 sea ice thickness ±2σ shaded, and melt probability repeated from Figs. 3, 4, and 6. Note the inverted axis of the sea ice thickness time series. Characteristic springtime acceleration near the TIS front begins with the breakup of landfast sea ice, and is possibly enhanced later in the year by shear margin weakening resulting from surface melt. Vertical shaded blue areas indicate typical times of landfast ice connection with TIS. Two years of the mean cycle are shown for visual continuity. Velocity anomalies predicted from basal melt are not shown here because the ±1 m yr⁻¹ amplitude would visually indiscernible at the observed scale of interannual velocity variability.

ECCO v4-r3 model was indirectly constrained by observations of sea ice concentration, but also accounts for winter sea ice growth.

6 Discussion

6.1 Causes of seasonal variability

Studies of floating and marine-terminating glaciers around the world have found a diverse set of causes of seasonal velocity variability, suggesting that local phenomena control glacier flow on subannual timescales, and there is no single dominating global cause of seasonal variability. In some regions of Greenland, neighboring glaciers behave differently based on the mechanisms that control them, and those mechanisms can change for a given glacier throughout the year (Howat et al., 2010; Moon et al., 2014).

On grounded ice, seasonal velocity variability often results from surface water draining to the bed, where it can temporarily pressurize an inefficient hydrological system, allowing the overlying ice to accelerate until an efficient drainage system forms or the water otherwise evacuates (Zwally et al., 2002; Parizek and Alley, 2004; Bartholomew et al., 2010). At Totten Glacier, we detect very little seasonal velocity variability on grounded ice, and the onset of acceleration we observe on the floating ice shelf begins well before surface water is detected anywhere in the region (Fig. 7). We therefore rule out the possibility that surface melt is responsible for initiating TIS acceleration each year.
On multi-year timescales, basal melt can lead to ice shelf acceleration as thinning reduces the internal buttressing strength of ice (Christianson et al., 2016; Greene et al., 2017a). However, neither the timing nor the amplitude of melt-induced thinning can account for the seasonal velocity variability we observe at TIS. We find a small seasonal cycle of ice shelf thickness due to variable basal melt throughout the year, but the corresponding velocity anomalies should be two orders of magnitude smaller than the observed velocity anomalies. Roberts et al. (2017) pointed out that at TIS, a mechanism exists that can amplify the effects of basal melt on ice velocity: Ice rumples in the middle of TIS may provide decreasing resistance to flow as the ice shelf thins, so it is possible that the simple model of ice shelf buttressing we employ in Sec. 4 underestimates the effects of basal melt on TIS velocity. Nonetheless, we find that basal melt is still incapable of causing the observed seasonal velocity cycle because ice shelf thickness maxima occur each year nearly coincident in time with observed velocity maxima.

In the GoLIVE dataset and in MODIS-derived velocities, we find that the outer TIS accelerates each year between spring and autumn. The spatial pattern of the annual acceleration suggests that the flow of ice is governed by processes at the ice front (Fig. 1), and timing of the acceleration implicates the annual breakup of landfast ice at the TIS front as an influencing factor. Figure 7 shows the relationship between sea ice thickness, surface melt, and TIS velocity. The temporal and spatial resolution of data available at TIS limit our ability to investigate the specific processes by which the presence of sea ice may slow the flow of TIS, but similar studies elsewhere have found that backstress from ice mélange (Walter et al., 2012; Todd and Christoffersen, 2014; Otero et al., 2017) can stabilize the ice front and reduce or entirely shut down calving over winter (Sohn et al., 1998; Reeh et al., 2001; Amundson et al., 2010; Moon et al., 2015; Robel, 2017), thus preserving internal stresses in the glacier and slowing its flow (Krug et al., 2015). The pattern of TIS acceleration we observe is similar to seasonal velocity anomalies observed at other marine-terminating glaciers and ice shelves, where the annual breakup of sea ice causes velocity anomalies that are seen up to tens of kilometers from the glacier terminus (Nakamura et al., 2010; Walter et al., 2012; Zhou et al., 2014).

We find that the outer TIS accelerates each year, likely in response to lost buttressing upon the breakup of rigid sea ice at the ice shelf terminus. The response we observe is consistent with other studies that have shown a seasonal pattern of ice front calving and glacier acceleration in response to the disintegration of rigid sea ice caused by warm sea surface temperatures (Howat et al., 2010; Cassotto et al., 2015; Luckman et al., 2015). Ice front processes are likely responsible for the onset of TIS acceleration each spring, but we cannot rule out the possibility that other factors may influence the flow of TIS in other parts of the year. It is possible that onset of acceleration begins with the breakup of sea ice at the TIS front, but surface melt could play a role later in the summer or autumn, if water percolates into the ice and weakens the shear margins.

In Figs. 3 and 7 we approximate the seasonal velocity variability of TIS as a sinusoid. The seasonal flow of TIS is likely more complex, and the timing and magnitude of spring-to-autumn speedup presumably vary from year to year. Nonetheless, we have shown that TIS responds to local forcing on subannual timescales, the response is observable, and it correlates with the breakup of sea ice at the glacier terminus each spring.
6.2 Impacts of seasonal variability on measurements of long-term change

We find that TIS accelerates each year from spring to autumn, and this seasonal variability has the potential to contaminate estimates of annual mass flux and interannual variability. Most common methods of measuring ice velocity rely upon subannual displacement measurements to characterize annual ice flux (e.g. Mouginot et al., 2017a), but where ice velocity varies throughout the year, short-term measurements can be aliased by the natural seasonal cycle and provide an inaccurate measure of annual ice flux. The seasonal variability we observe at TIS suggests that measurements acquired in the spring likely underestimate, and autumn measurements overestimate, the mean annual velocity of the ice shelf.

The most significant seasonal variability at TIS is found near the ice front, where spring and autumn velocities can differ by up to 10% percent. Although this represents a small modulation of the mean flow, it is on the order of interannual variability that has previously been attributed to interannual changes in ocean forcing, and the pattern of summer acceleration we show in Fig. 1 bears a notable resemblance to accelerations that have previously been reported as evidence of long-term change (Li et al., 2016). Although direct investigations of interannual change are beyond the scope of this study, we can consider how seasonal variability may have influenced previous studies of TIS velocity.

Velocity variability at TIS has been investigated in three recent papers that tracked ice accelerations and slowdowns over the past few decades, and each study found that on interannual timescales, TIS dynamically responds to ocean forcing from below. We do not find any evidence that contradicts the overall findings of the previous studies, but in some cases, velocities were measured over periods of less than one year, and may have been aliased by seasonal variability. Roberts et al. (2017) and Greene et al. (2017a) each measured displacements between images separated by near-integer multiples of years. By this method, it is unlikely that they inadvertently captured subannual variability, unless the timing of acceleration events occurred out of sync with the calendar year. Such is likely the case for the 2009 to 2010 acceleration observed by Li et al. (2016), who compared velocity measurements obtained between September and January of both years. Although the periods of observation were roughly the same in both years, the spring breakup of fast ice did not occur until after the start of observations in 2009, whereas the spring breakup was already underway when observations began in 2010, and the TIS had already begun to respond. The velocity difference between the 2009 and 2010 measurements shows acceleration focused at the ice shelf terminus, and this likely reflects a difference in timing of the seasonal cycle that may not be associated with any difference in mean annual velocities. The inconsistent timing of fast ice breakup each year suggests that assessments of interannual change made from short-term displacement measurements can be contaminated by seasonal effects, even if observations are taken at the same time each year.

Despite the seasonal variability we observe near the TIS front, mass balance of an ice sheet is more meaningfully measured at the grounding line, where ice begins to have an impact on sea level. Our results show little subannual velocity variability at the grounding line, thus supporting the grounding line flux estimates by Li et al. (2016).
6.3 Sea ice influence on ice sheet mass balance

The GoLIVE and MODIS velocity measurements show that TIS is sensitive to environmental forcing on subannual timescales, and its flow is primarily controlled by the presence of sea ice at the TIS front. This finding warrants consideration of how changes in sea ice could affect the stability of the TIS and the long-term mass balance of the Aurora Subglacial Basin. Elsewhere in Antarctica, loss of multiyear landfast ice has led to major calving events and glacier acceleration (Khazendar et al., 2009; Miles et al., 2017; Aoki, 2017). The landfast ice we observe is not multiyear ice, and is thus unlikely to be associated with any catastrophic events at TIS in the near future. However, calving front processes can have far-reaching effects on glacier thickness and velocity (Nick et al., 2009), and it is possible that long-term changes in winter sea ice cover (Bracegirdle et al., 2008) could have integrated effects on TIS buttressing: If the duration and thickness of winter sea ice control the total annual buttressing at the ice shelf front, long-term changes in sea ice cover could affect the annual flow of TIS, and potentially the mass balance of TIS and the Aurora Subglacial Basin.

7 Conclusions

We find that TIS has a characteristic seasonal velocity profile, which could lead to inaccurate estimates of the annual mass balance of TIS, and may have aliased some previous measurements of interannual variability. Annual ice velocity maps are now available covering most of Antarctica (Mouginot et al., 2017c), but interpreting such datasets at TIS and elsewhere requires understanding where ice velocity varies seasonally and by how much. Our results provide context for how and where such velocity mosaics may be used to interpret interannual change at Totten Glacier.

Previous studies have linked interannual velocity variability at TIS to periods of ice shelf thickening and thinning caused by sustained basal melt rate anomalies. On subannual timescales, however, the seasonal amplitude of basal melt variability is insufficient to produce enough thinning to elicit an observable velocity response. Furthermore, seasonal basal melt anomalies result in an ice shelf that is thinnest, weakest, and should flow the fastest in winter, when our observations show the TIS reaches its minimum velocity.

In accord with other studies of ice shelves and glaciers around Antarctica and Greenland, we find that the seasonal variability of TIS velocity is most closely linked to the presence of sea ice at the ice shelf front. Each spring when surface waters warm, rigid landfast ice breaks its connection to the TIS front, the calving rate increases, and the TIS responds by accelerating by nearly 10% close the ice shelf terminus. Velocity anomalies are most significant over floating ice, and spring acceleration precedes surface melt each year, together suggesting that subglacial hydrology does not cause the seasonal cycle of TIS velocity we observe.

We find that winter sea ice is a primary contributor to the seasonal variability of the outer TIS velocity. If the future brings long-term changes in the thickness or extent of winter sea ice, the integrated effects of changes in buttressing could manifest as long-term changes in the mass balance of TIS and the Aurora Subglacial Basin.
Figure A1. Bootstrap distributions of sinusoid parameters. Scattered data show the phase and amplitudes of sinusoids fit by least squares to 10,000 random samples of MODIS velocity measurements. Red color indicates local density of the scattered data. Histograms show the distributions of each parameter. Contour lines and axis tick marks indicate one-standard-deviation intervals for each parameter.

Code and data availability. GoLIVE data (Scambos et al., 2016; Fahnestock et al., 2016) is available at https://nsidc.org/data/NSIDC-0710. MODIS images (Scambos et al., 2001, updated 2018) used in this study were obtained from ftp://sidads.colorado.edu/pub/DATASETS/ICESHelves. The ImGRAFT template matching software (Messerli and Grinsted, 2015) is available at http://imgraft.glaciology.net. The melting-1979-2017-v2.nc surface melt data from Picard and Fily (2006) are available at http://pp.ige-grenoble.fr/pageperso/picardgh/melting/. ECCO v4-r3 sea ice effective thickness data can be found at ftp://ecco.jpl.nasa.gov/Version4/Release3/nctiles_daily/SIheff. Analysis was performed with Antarctic Mapping Tools for MATLAB (Greene et al., 2017b). The background image in Figs. 1, 4, and 5 is the MODIS Mosaic of Antarctica (Haran et al., 2014). Figs. 1, 4, 5, and 6 use cmocean colormaps (Thyng et al., 2016).

Appendix A: Uncertainty quantification

To verify the presence of seasonal variation in ice flow speed, we performed a p test using the null hypothesis that the amplitude of seasonal variability is 0 m yr\(^{-1}\). The 565 MODIS velocity measurements match the sinusoid fit by least squares with a Pearson correlation coefficient of \(r = 0.472\) and a corresponding \(t_{\text{statistic}} = 12.70\). The probability of the null hypothesis that there is no seasonal cycle is \(p = 6 \times 10^{-33}\), and thus we reject it in favor of the alternate hypothesis that cyclic seasonal behavior is present at TIS.

We used a bootstrapping technique to estimate uncertainty in the characteristic sinusoid fit to the MODIS velocity data. Figure A1 shows the phases and amplitudes of sinusoids fit by least squares to 10,000 random resamplings of the MODIS velocity dataset. The mean amplitude of the sinusoid is 106 m yr\(^{-1}\) with a 1-sigma uncertainty of 9.1 m yr\(^{-1}\). The phase of the sinusoid is characterized by a maximum velocity on March 21 (and corresponding minimum September 19) with a 1-sigma
uncertainty of 4.9 days. The root-mean-square of the measurement residuals is 92.6±2.9 yr⁻¹, and reflects a combination of measurement error, interannual variability in amplitude and timing of acceleration or slowdown, and the difference between true seasonal variability and the sinusoid approximation.

Author contributions. CAG conceived of this study, generated the figures, and wrote the manuscript. Analysis was conducted by CAG under the direction of DDB, with guidance from DAY. DEG and BKGF developed the ice/ocean model described in Section 4 and assisted in interpreting its results.

Competing interests. The authors declare that they have no conflicts of interest.

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