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3	Improved retrieval of land ice topography from CryoSat-2 data and its
4	impact for volume change estimation of the Greenland Ice Sheet
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8	Abstract
9	A new methodology for retrieval of glacier and ice sheet elevations and elevation changes from
<mark>10</mark>	CryoSat-2 data is presented. Surface elevations and elevation changes determined using this
<mark>11</mark>	approach show significant improvements over ESA's publically available Cryosat-2 elevation
<mark>12</mark>	product (L2 Baseline-B), compared to near-coincident airborne laser altimetry from NASA's
<mark>13</mark>	Operation IceBridge and seasonal height amplitudes from the Ice, Cloud, and Elevation Satellite
14	(ICESat).
15	Applying this methodology to CryoSat-2 data collected in Interferometric Synthetic
16	Aperture mode over the high relief regions of the Greenland ice sheet we find an improvement
17	in the root-mean-square-error (RMSE) of 27% and 40% compared to ESA's L2 product in the
18	derived elevation and elevation changes, respectively. In the interior part of the ice sheet, where
19	CryoSat-2 operates in Low Resolution Mode, we find an improvement in the RMSE of 68% and
20	55% in the derived elevation and elevation changes, respectively. There is also an $86%$
21	improvement in the magnitude of the seasonal amplitudes when compared to amplitudes
22	derived from ICESat data. These results indicate that the new methodology provides improved

24 tracking of the snow/ice surface with lower sensitivity to changes in near-surface dielectric

25 properties.

To demonstrate the utility of the new processing methodology we produce elevations, elevation changes and total volume changes from Cryosat-2 data for Greenland Ice Sheet during the period Jan-2011 to Jan-2015. We find that the Greenland Ice Sheet decreased in volume at rate of  $289 \pm 20$ , km<sup>3</sup> a<sup>-1</sup>, with high inter-annual variability and spatial heterogeneity in rates of loss. This rate is 65 km<sup>3</sup> a<sup>-1</sup> more negative than rates determined from ESA's L2 product, highlighting the importance of Cryosat-2 processing methodologies.

32 1 - Introduction

33 The European Space Agency (ESA) launched CryoSat-2 in April 2010 tasked with monitoring 34 the changes of the Earth's land and sea ice. CryoSat-2 carries a new type of Doppler/delay radar altimeter (Raney, 1998) referred to as SIRAL (SAR Interferometric Radar Altimeter). 35 36 SIRAL operates in two different modes over land ice. Over the interior part of the ice sheets it 37 operates as a conventional pulse limited radar system, referred to as the "Low Resolution 38 Mode" (LRM). In more complex high-sloping terrain the system uses a novel second antenna to 39 operate in "Interferometric Synthetic Aperture Radar" (SIN) mode. These new features allow the 40 satellite to monitor changes in complex terrain including ice caps, glaciers and the high relief 41 marginal areas of the ice sheets. Such areas are sensitive to changes in climate and contribute 42 greatly to current rates of sea level rise, (e.g., Gardner et al. (2013) and Shepherd et al. (2012)). 43 Ku-band radar altimeters are insensitive to cloud cover providing superior coverage to 44 laser altimeters (e.g., ICESat) but experience significant amounts of volume scattering, which is 45 controlled by the time-evolving dielectric properties of the near-surface snow, firn, and ice (Lacroix et al., 2008; Remy et al., 2012). These effects can have large implications for the 46 47 determination of mass change over a wide range of both spatial and temporal scales. Changing 48 snow conditions can introduce time-varying biases in the data that, in combination with the radar

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52	signals interaction with the surface, introduce large elevation biases (0.5 - 1 m) (Nilsson et al.,	
53	2015a). This, combined with other factors such as processing methodology and surface	
54	topography, makes it difficult to measure small changes for much of the word's ice covered	
55	regions (Arthern et al., 2001; Gray et al., 2015; Nilsson et al., 2015b).	
56	The mitigation of these effects in the processing of radar altimetry data is required for	
57	improved accuracy of derived temporal and spatial changes in surface elevation of glaciers and	
58	ice sheets. Several studies have proposed different approaches to assess these effects and	
59	improve the retrieval process of surface elevation and elevation changes from radar altimetry	
60	data. These include different approaches to waveform retracking (Davis, 1993, 1997; Gray et	
61	al., 2015; Helm et al., 2014) and empirical corrections to the estimated surface elevation	
62	changes (Davis and Ferguson, 2004; Flament and Rémy, 2012; Sørensen et al., 2015;	
63	Wingham et al., 2006b; Zwally et al., 2005, 2011). Relatively little work has been done to assess	
64	methods for, improving elevation and elevation changes derived from ESA's CryoSat-2 data	Johan Nilsson 9/18/2016 5:50 PM
65	Abulaitijiang et al., 2015; Gray et al., 2013, 2015; Helm et al., 2014).	Deleted: from
65 66	(Abulaitijiang et al., 2015; Gray et al., 2013, 2015; Helm et al., 2014). Here we conduct a thorough analysis of CryoSat-2 SIN and LRM waveform retracking	
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81 mapper and seasonal height amplitudes estimated from Ice, Cloud, and Elevation Satellite

82 (ICESat) data.

# 83 *2* - Surface elevations from CryoSat-2

84 2.1 - Low Resolution Mode (LRM)

85	The LRM mode is used over the interior parts of the ice sheet, which mostly consist of low
86	sloping terrain, Here, SIRAL operates as a conventional pulse limited radar system with a
87	transmission frequency of 13.6 GHz (Ku-band) and has Pulse-Limited Footprint (PLF) radius of
88	approximately 1.5 km and a beam-limited footprint (BLF) radius of approximately 7.5 km over
89	flat terrain (Bouzinac, 2014). The gentle terrain allows for accurate mapping of the surface
90	elevation of the ice sheet down to decimeter-level (Brenner et al., 2007). Within the LRM
91	waveform we define the location of the surface from the leading edge of the waveform, based
92	on a fraction of the maximum amplitude of the received power. This approach is commonly
93	referred to as a threshold retracker. Following Davis et al. (1997) we use 20% threshold to
94	define the location of the surface. Davis et al. (1997) argued that a 20% threshold represents
95	the best compromise between waveforms that are entirely dominated by either volume or
96	surface scattering, making it suitable for obtaining estimates of surface elevation for most parts
97	of the Greenland Ice Sheet.
98	The CryoSat-2 LRM radar waveforms suffer from measurement noise, in the form of
99	speckle noise. Furthermore, over the steeper parts of the LRM-area the range gate tracking-
100	loop can loose track of the surface, producing non-usable waveforms. To remove bad or loss of
101	track waveforms the radar waveform (20 Hz) is first filtered using a zero-phase low pass filter to
102	reduce speckle noise on a line-by-line basis. The signal-to-noise-ratio (SNR) of the waveform is

103 then estimated and if the SNR < 0.5 dB the waveform is rejected. The SNR threshold was

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106 empirically chosen to obtain a good trade-off between the quality of the measurements and

107 sampling.

108 Before the waveform can be retracked the first surface return (first major peak) is 109 identified within the range gate window. A copy of the waveform is heavily smoothed to remove 110 small-scale surface roughness signals, keeping the overall surface signal intact. The range gate 111 index of the first peak from the copy is then used to extract the leading edge of the original low 112 pass filtered waveform. Only leading edges with a peak index above 20 are used in the 113 retracking, as peaks before or after that indicate troublesome surface ranging. The extracted 114 leading edge is then oversampled by a factor of 100 (c.f. (Gray et al., 2013; Helm et al., 2014), 115 and the range R between the surface and satellite is determined based on the 20% threshold 116 computed according to Davis et al. (1997). The range is then corrected for several atmospheric 117 and geophysical effects relevant to land ice studies according to Bouzinac (2014). The surface 118 elevation H of the topography, relative to the WGS84 ellipsoid, is estimated as H = A - R, where 119 A is the altitude of the satellite. 120 The measured surface return over a sloping surface does not originate from the 121 satellites nadir location, but from the "Point Of Closest Approach" (POCA) to the spacecraft 122 (Brenner et al., 1983). These off-nadir returns can introduce a large range bias to the surface, 123 depending on the magnitude of surface slope, ranging from 0-120 m (Brenner et al., 1983) as 124 the measured surface height is mapped to an erroneous position (i.e. the nadir position). To 125 mitigate the effect of this error we correct the measured range and location to the POCA point 126 using an a-prior DEM, following the approach of Bamber (1994). In contrast to previous studies 127 we account also for the local surface curvature, as Remy et al. (1989) showed that accounting 128 for surface curvature in addition to surface slope significantly improve results. The surface 129 slope, aspect and curvature are estimated from an a priori DEM. The GIMP elevation model 130 (Howat et al., 2014) was used to derive surface parameters for the slope-induced error 131 correction in the LRM mode. The DEM was resampled to 2 km resolution, using bilinear

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133	interpolation, prior to parameter estimation, corresponding to the pulse-limited footprint of the	
134	LRM mode,	Johan Nilsson 9/25/2016 5:29 PM
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135	2.2 - Interferometric Synthetic Aperture Radar Mode (SIN)	
136	The SIN mode is used over the marginal areas of the ice sheets and other smaller glaciated	
137	areas. In these areas the SIRAL altimeter operates as a Doppler/Delay radar system (Raney,	Johan Nilsson 9/18/2016 6:10 PM
138	1998). The Doppler/Delay radar allows for higher along-track resolution compared to	Deleted: Keith
139	conventional altimetry, resulting in 350 m resolution in along track and 1500 m across track. In	Johan Nilsson 9/18/2016 6:17 PM
140	ordinary SAR operation only the amplitude of the radar echo is measured and the phase content	Deleted: . SIN modes allows for a
141	is discarded or ignored. With the inclusion of a second antenna on CryoSat-2, interferometric	
142	SAR can also be performed. Difference in the path length between the POCA and the individual	
143	antennas introduce a phase shift between the two retrieved signals that can be related to the	
144	angle of arrival (look angle). The look angle can in turn be used to resolve the across track	
145	(across antenna) location of the echo.	
146	Multi-look processing is applied to ESA's L1B waveform product (Bouzinac, 2014) to	
147	reduce the noise in the SIN waveform but it is still affected by speckle-noise, as is the case for	
148	the LRM waveforms. To mitigate this effect, and to help identify the leading edge of the first	
149	return, we apply speckle reduction filtering and leading edge extraction of the SIN waveforms in	
150	the same way as for the LRM processing with minor changes due to differences in range gate	
151	resolution. In this case, compared to the LRM retracking algorithm, only leading edges with a	
152	peak index in the range of 20-350 are used for retracking the radar waveform.	
153	The estimated coherence C of the multi-looked waveforms is then filtered in two stages;	
<mark>154</mark>	(i) all coherence measures larger than one is set to zeros (larger than one coherence exists in	
155	(the L1B product reason unknown). (ii) The coherence array, as a function of range is filtered	Johan Nilsson 9/18/2016 6:27 PM
156	using a 2D 5x5 Wiener filter to remove high frequency noise. The filtering of the waveform and	Deleted: the Johan Nilsson 9/18/2016 6:26 PM Deleted: range power image

162	the coherence is applied to remove noise in the recreation of the interferogram. This is further	
163	discussed later.	
164	The measured differential phase $\phi$ of the return signal is affected by phase ambiguities;	
165	a sudden shift of $2\pi$ in the measured phase. To reduce phase noise and aid the phase, an	
166	unwrapping of the radar interferogram <i>I</i> is performed according to Gray et al. (2013):	
167	$I = P \cdot C \cdot e^{-i\phi} \tag{1}$	
168	The interferogram is then filtered using a wavelet-based de-noising technique, where the real	
169	and imaginary parts of the interferogram are filtered separately. The unwrapping of the	
170	interferogram allows for indirect filtering of the phase, without being affected by the phase-	
171	ambiguities. Phase filtering is an important consideration as it has a direct affect on accuracy of	
172	the position of the ground echo. We selected a bi-orthogonal as the mother wavelet to produce	
173	the wavelet coefficients decomposed into three levels. Soft thresholding was applied to detail	
174	coefficients, using a heuristic threshold rule to remove noise at every level. This was done on a	
175	line-by-line basis. The final filtered differential phase was then recovered by:	
176	$\phi_f = Re\{I_f\} + Im\{I_f\} \tag{2}$	
177	To resolve the phase ambiguities the filtered phase measurements require unwrapping. The	
178	phase unwrapping is done on a line-by-line basis in two directions starting from the center of	
179	gravity of the waveform (Wingham et al., 1986).	
180	The return power distribution of a Doppler/Delay radar system shows an important	
181	distinction from those from conventional pulse-limited radar systems. Here, the point	
182	corresponding to the mean surface is not located at the half-power point on the leading edge,	
183	but rather closer to the maximum (Wingham et al., 2006a). Therefore a new retracker has been	
184	developed, closely related to the one used in Gray et al. (2013), to allow for adaptive retracking	
185	of the upper parts of the leading edge of the SAR waveform. The algorithm follows the main	
186	concept of the threshold retracker, developed by Davis, (1997), but instead of a pre-defined	

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188 threshold it tracks the maximum gradient of the leading edge of the waveform. We refer to this

189 approach as that "Leading-edge Maximum Gradient retracker" (LMG).

The surface returns are geolocated using the across track look-angle  $\theta$  estimated from the differential phase at the retrackig point according to (Wingham et al., 2006a). This, in combination with the viewing geometry, is used to define the location of the surface return on the ground using basic across track interferometric principles. We correct  $\theta$  for the interferometer surface slope error by applying the look-angle scaling factor estimated in (Galin et al., 2013).

196 The along-track differential phase estimate, interpolated to the retracking point, is 197 affected by phase ambiguities not corrected for during the phase unwrapping procedure. To 198 reduce residual phase ambiguities an a priori DEM (GIMP) is used to extract the DEM surface, 199 resampled to 500 m resolution (corresponding to the along-track sampling), elevations at the 200 nadir and echolocation using bilinear interpolation. Over a sloping surface the surface return 201 should always come from a position upslope from the nadir point. Therefore the following 202 relation must hold where  $(H_{echo} > H_{nadir})$  or for a more practical application  $(H_{echo} - H_{nadir}) > \varepsilon$ , 203 where  $\varepsilon$  is the uncertainty of the DEM used. If this relation is violated  $2\pi$  is added or subtracted 204 to the individual along-track phase estimate, depending on the sign. 205 A final step is applied to correct for any lingering phase ambiguities not corrected by the 206 a priori DEM. This step uses the assumption that the along-track phase should follow a 207 consistent pattern over most part of the satellite ground track. Hence, any large discrepancies

from the overall pattern of the along-track phase would indicate an ambiguity. The ambiguity is detected by computing the residuals of the along-track phase by removing a smoothed version. If any of the residuals have a magnitude larger than  $\pi$  it is considered ambiguous and thus corrected by adding or subtracting  $2\pi$ .

# 213 3.1 – Surface fit method

214	The surface-fitting method is based on fitting a linear model to the elevations as a function of		
215	time and space inside a search radius of 1 km (e.g., Howat et al., 2008; Moholdt et al., 2010;		
216	Sørensen et al., 2011; Wouters et al., 2015). The linear model consists of a time-invariant		
217	(static) bi-quadratic surface model to account for variable topography inside the search radius		
218	and time-variant part used to extract the temporal change in elevation. The model consists of a		
219	total of 7-parameter whereof six of the parameters (a-coefficients) describe the bi-quadratic		
220	surface modeling function, <i>dh/dt</i> the linear elevation change rate, t time in decimal years, to the		
221	mean time inside the footprint and $\varepsilon$ the residuals from the linear regression.		
222	$h_{\mathbf{y}}(x, y, t) = a_{0} + a_{1}x + a_{2}y + a_{3}xy + a_{4}x^{2} + a_{5}y^{2} + \frac{\partial h}{\partial t_{\mathbf{y}}}(t - t_{0})_{\mathbf{y}} + \varepsilon $ (3)		
223	The algorithm estimates the elevation change at every echolocation (or grid-node if desired) in		
224	the data set. In each solution the signal amplitude and phase are also estimated by fitting a		
225	seasonal signal model to the surface-fit elevation residuals, according to:		
225 226	seasonal signal model to the surface-fit elevation residuals, according to: $\Delta h(t) = s_0 \cos(wt) + s_{1v} \sin(wt) + \varepsilon \qquad (4)$		
226	$\Delta h(t) = s_0 \cos(wt) + s_{1v} \sin(wt) + \varepsilon $ (4)		
226 227	$\Delta h_{x}(t) = s_{0} \cos(wt) + s_{1x} \sin(wt) + \varepsilon $ (4) where $\Delta h_{x}$ is the elevation residuals estimated from the plan-fit model, $s_{0,1}$ are the model		
226 227 228	$\Delta h(t) = s_0 \cos(wt) + s_1 \sin(wt) + \varepsilon \qquad (4)$ where $\Delta h_i$ is the elevation residuals estimated from the plan-fit model, $\underline{s}_{0,1}$ are the model coefficients and t the time. The amplitude <i>A</i> is then defined as $A = \sqrt{s_0^2 + s_1^2}$ and the phase <i>P</i> as		
226 227 228 229	$\Delta h_{k}(t) = s_{0} \cos(wt) + s_{1w} \sin(wt) + \varepsilon $ (4) where $\Delta h_{k}$ is the elevation residuals estimated from the plan-fit model, $s_{0,1}$ are the model coefficients and t the time. The amplitude A is then defined as $A = \sqrt{s_{0}^{2} + s_{1}^{2}}$ and the phase P as $P = \tan^{-1}\left(\frac{s_{1}}{s_{0}}\right)$ .		
226 227 228 229 230	$\Delta h_{k}(t) = s_{0} \cos(wt) + s_{1w} \sin(wt) + \varepsilon \qquad (4)$ where $\Delta h_{k}$ is the elevation residuals estimated from the plan-fit model, $s_{0,1}$ are the model coefficients and t the time. The amplitude <i>A</i> is then defined as $A = \sqrt{s_{0}^{2} + s_{1}^{2}}$ and the phase <i>P</i> as $P = \tan^{-1}\left(\frac{s_{1}}{s_{0}}\right)$ . To remove outliers an iterative $3\sigma$ -filter is used in the full model solution, i.e. the		
226 227 228 229 230 231	$\Delta h_{k}(t) = s_{0} \cos(wt) + s_{1w} \sin(wt) + \varepsilon \qquad (4)$ where $\Delta h_{k}$ is the elevation residuals estimated from the plan-fit model, $s_{0,1}$ are the model coefficients and t the time. The amplitude <i>A</i> is then defined as $A = \sqrt{s_{0}^{2} + s_{1}^{2}}$ and the phase <i>P</i> as $P = \tan^{-1}\left(\frac{s_{1}}{s_{0}}\right)$ . To remove outliers an iterative $3\sigma$ -filter is used in the full model solution, i.e. the topography, trend and seasonal signal are removed, using a maximum of 5-iterations. For each		
226 227 228 229 230 231 232	$\Delta h_{k}(t) = s_{0} \cos(wt) + s_{1w} \sin(wt) + \varepsilon \qquad (4)$ where $\Delta h_{k}$ is the elevation residuals estimated from the plan-fit model, $s_{0,1}$ are the model coefficients and t the time. The amplitude <i>A</i> is then defined as $A = \sqrt{s_{0}^{2} + s_{1}^{2}}$ and the phase <i>P</i> as $P = \tan^{-1}\left(\frac{s_{1}}{s_{0}}\right)$ . To remove outliers an iterative $3\sigma$ -filter is used in the full model solution, i.e. the topography, trend and seasonal signal are removed, using a maximum of 5-iterations. For each iteration residuals (full-model) with an absolute value larger than 10 m are removed, as		

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(5)

$$W = \frac{1}{\left(1 + \left[\frac{d}{\rho}\right]^2\right)}$$

where *W* is the estimated weight, *d* the distance and  $\rho$  the correlation or resolution parameter set to 500 m. The weighting allows the solution to better reflect local signal dynamics at the prediction point.

253

256

Local elevation time-series are further computed from the elevation residuals and elevation trend from each solution, according to:

$$h(x, y, t) = (t - t_0) \cdot \frac{\partial h}{\partial t} + \varepsilon$$
(6)

where *t* is the time epochs inside the search cap,  $t_0$  is the mean time of *t*, *dh/dt* is the elevation change rate and  $\varepsilon(t)$  is the elevation residual at each time epoch.

The elevation changes estimated from the surface-fitting method are then culled to remove outliers before spatial gridding. Elevation changes with a regression error larger than 15 m  $a^{-1}$  are removed. The resulting surface elevations are binned at 5-km resolution for outlier editing purposes. For each cell the local spatial trend is modeled as a bilinear surface, and removed. The residuals are then edited using an iterative  $3\sigma$  filter until the RMS converges to 2%.

3.2 - Crossover method

The crossover method is used to derive surface elevation at the intersection point between an ascending and descending satellite ground track separated in time (Brenner et al., 2007; Khvorostovsky, 2012; Zwally et al., 1989). The surface elevations and times are then estimated at the crossover location for each track by linear interpolation of the two closest data points. The crossover height difference is then estimated by taking the height difference between the two tracks according to:

272	$\Delta h = h_2 - h_1 + \varepsilon \tag{7}$		
273	were $\underline{h}_1$ and $\underline{h}_2$ are the surface heights at the crossover location at time epoch $t_1$ and $t_2$ ,		
274	respectively, and <i>E</i> is the random measurement error, including orbital, range and retracking		
275	errors.		
276	This approach produces crossover height differences with scattered time-epochs		
277	ranging from 0-4 years. CryoSat-2 has a 369-day repeat orbit configuration with a 30-day sub-		
278	cycle meaning that each crossover location will be revisited every 369 days and surrounding		
279	area every 30 days. This produces annual and sub-annual crossover difference around each		
280	crossover location. This fact is used to produce elevation change rates by incorporating all		
281	multi-temporal crossover difference within a neighborhood of 2.5-km around each crossover		
282	location. The elevation change is then estimated using the same procedure described for the		
283	surface-fit method, except that a bilinear model is used to remove any spatial trends in the		
284	topography of the crossover elevations according to:		
285	$dh_{\mathbf{y}}(x,y,t) = \mathbf{a}_{1}\mathbf{x} + \mathbf{a}_{2}\mathbf{y} + \frac{\partial h}{\partial t}(t-t_{0})_{\mathbf{y}} $ (8)		
286	where $dh$ is the crossover height difference, $(t-t_0)$ the time difference, $a_{t_0}$ and $a_{t_0}$ the across and		
287	along-track slope and <u>dh/dt</u> the elevation change rate. This produces elevation changes		
288	comparable in time and in spatial coverage with the surface-fit method. The same outlier editing		
289	schemes is applied to the crossover elevation change rates as for the surface-fit method.		
290	$3.3_{-}$ - Gridding of sparse elevation and elevation change data		
291	The gridding is done in a polar-stereographic projection with a latitude of origin at 70°N, central		
292	longitude of 45°W and origin at the North Pole. The projection is referenced against the WGS-		
293	84 ellipsoid and the grid-resolution. The observations derived from the surface-fit are gridded at		
294	a resolution of 1x1-km, due to the high spatial sampling.		
295	The method of Least Squares Collocation (LSC), described in Herzfeld (1992) is used to		
296	grid the observations onto a regular grid. LSC is similar to Kriging and allows for optimal		

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(9)

310	interpolation and merging of data with different accuracies, using their inherent covariance	
311	structure. The LSC-algorithm uses the 25 closest data points in 8-quadrants surrounding the	
312	prediction point to reduce spatial biasing. The prediction equation consists of two terms where	
313	the first term is the actual prediction term and the second term accounts for the non-stationary	
314	part of the data, as described by:	
	$\hat{s} = C_{SZ}(C_{ZZ} + N)^{-1}Z + \left(1 - \sum_{v} (C_{SZ}(C_{ZZ} + N)^{-1})\right)m(z)$	
315	were $C_{sa}$ is the cross-covariance, $C_{za}$ is the auto-covariance, N the diagonal noise-matrix	
316	consisting of the a priori RMS-error and $m(z)$ is the median value of the observations inside the	
317	search neighborhood.	
318	The covariance of the data inside the local neighborhood is modeled as a function of	
319	distance away from the prediction point using a third-order Gauss-Markov model described	
320	below.	
321	$C(r) = C_0 \left( 1 + \frac{r}{\alpha} - \frac{r^2}{2\alpha^2} \right) e^{\left(-\frac{r}{\alpha}\right)} $ (10)	
322	where <i>r</i> is the separation distance, $C_0$ the local data variance and $\alpha$ is a scaling factor estimated	
323	from the correlation length.	
324	LSC interpolation provides a RMS-error for each prediction point estimated from the	
325	modeled covariance of the data according to:	
	$C_{\hat{s}} = C_0 - C_{sz} (C_{zz} + N)^{-1} C_{sz}^T $ (11)	
326	where the RMSE of the prediction equals to $\sigma_{s} = (C_{s})^{1/2}$ and where $C_{ss}^{T}$ is the transposed cross-	
327	covariance matrix.	
328	The elevation changes estimated from the surface-fit and crossover methods are	
329	interpolated to a regular grid using their a priori error estimated from the LSC scheme. To avoid	
330	unrealistically small errors, common in the regression errors estimated over flat terrain, a	
004		
331	minimum error threshold is applied. Error values smaller than a specific threshold are set to the	

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changes over flat terrain and is set to 0.2 m a<sup>-1</sup>. The data are then gridded using a <u>75, km</u>
correlation length determined from the comparison of CryoSat-2 elevation to airborne
measurements (Section 5).

350 The LSC algorithm is also used to generate a DEM based on the surface elevations 351 generated from the surface-fitting algorithm. The surface elevations generated from the surface-352 fit were used as input to the gridding-algorithm. The use of surface elevations from the surface-353 fit provides several advantages compared to the raw observations as they: provide an almost 354 equal number of observations as the raw data, have been screened for gross outliers, have 355 been low-pass filtered using the 1-km search radius, and are all reference to the same time 356 epoch. Further the RMSE error generated from the surface-fit estimated surface height can be 357 used as an a priori error for the LSC gridding procedure.

The DEM is generated using the same approach as for the surface elevation changes, as described previously in the section. Before the gridding procedure is applied elevations H < 0and H > 3350 m are removed from the data set. Further, elevations with a standard error larger than 30 m are also removed. The elevations are binned spatially into a resolution of 1000 m and inside each cell the local surface trend is removed by fitting of a planar surface, and an iterative 30 filter is applied to the residuals to remove outliers.

364 4 - Surface elevations and elevation changes from ICESat

To assess basin-scale patterns of elevation change we compare elevation changes from CryoSat-2 data to elevation changes derived from Ice, Cloud, and Elevation Satellite (ICESat) data. Here we use release 33 (GLA06) data collected over the 2003-2009 period. The ICESat surface heights were used to generate surface elevation changes and seasonal parameters according to method M3 in Sørensen et al. (2011). The derived elevation changes were corrected for the G-C offset (Borsa et al., 2014). Valid elevation retrievals were selected Johan Nilsson 9/21/2016 6:14 PM Deleted: 50

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372	according to Nilsson et al. (2015b). The ICESat elevation, seasonal amplitude and phase, are	
373	then used for comparison with CryoSat-2 and to build continuous time series using the surface	
374	fit method described in Section 3.1. For the purpose of this study no correction for the inter-	
375	campaign bias was applied, as this is still an active area of investigation.	
376	5 -Validation	Johan Nilsson 9/20/2016 10:03 AM Deleted:
377	Elevation and elevation change results were generated for the entire Greenland Ice Sheet using	
378	CryoSat-2 data collected between Jan-2011 and Jan-2015 using the methodology presented in	
379	(Sections 2-3) (JPL product) and by applying the methods of (Section 3) to ESA's CryoSat-2 L2	
380	elevation products (ESA product). Surface elevations and elevation changes were validated	
381	against airborne data sets obtained from NASA's Operation Ice-Bridge Airborne Topographic	
382	Mapper (ATM), obtained from the "National Snow & Ice Data Center" (NSIDC) in the form of the	
383	ILATM2 product. The generated elevation product has a resolution of 80 m, with a 40 m spacing	
384	along-track. This mission produces both elevation and elevation changes with reported vertical	
385	accuracy of ~10 cm and temporal accuracy in the cm-level (Krabill et al., 2002).	
386	The derived surface elevations from CryoSat-2 are differenced against ATM surface	Johan Nilsson 9/19/2016 1:06 PM Deleted:
387	elevations within 50 m of each ATM locations. One month of CryoSat-2 data consistent in time	
388	with the ATM elevations are used for the validation to avoid biases due to temporal sampling	
389	and to obtain sufficient sample size. A total of four years of campaign data are used for the	
390	validation of the surface elevations (2011-2014). The residuals are edited using an iterative 3 $\sigma$	
391	filter to remove outliers. The accuracy and precision is estimated as the mean and standard	
392	deviation of the differences, respectively. The residual distribution is further binned according to	
393	surface slope estimated from the GIMP DEM (Howat et al., 2014) resampled to 500 m. The	
394	sensitivity to surface slope (slope error) can be identified in the standard deviation of the binned	Johan Nilsson 9/22/2016 3:50 PM
395	residuals and can be used to judge the quality of the produced surface elevation and elevation	<b>Deleted:</b> (slope error) is then defined by t a 1 <sup>st</sup> order polynomial to the slope and hei residuals. The rate estimated from the polynomial provides an indication of the magnitude of the slope-induced error over entire slope interval.

404	changes, while the binned-average for the elevations can be used to determine radar-signal	
405	penetration.	
406	Surface elevation change rates estimated from three different time-periods (2012-2014,	
407	2011-2013 and 2011-2014) of overlapping ATM observations (Krabill, 2014) are used to validate	
408	the surface elevation changes estimated from the CryoSat-2 data. The same validation	
409	methodology applied to surface elevations is applied to surface elevation changes, with a few	
410	minor modifications. First the search radius is increased to 175 m to make it conform to the ATM	
411	elevation change resolution of 250 m, as this search radius encloses the entire ATM grid cell.	
412	Secondly the estimated mean and standard deviation are multiplied with the individual time-	
413	intervals of the validation data sets to make the errors comparable.	
414	For the surface-fit and crossover methods, near-coincident elevation change rates were	
415	compared with ATM rates (e.g., April-2011 to April-2014). This provided three validation data	
416	sets for the surface-fit method, due to its high spatial coverage. However, only the 2011-2014-	
417	validation data set could be used for the crossover method, due to the lower spatial sampling of	
418	the crossovers.	
419	The overall accuracy and precision for both the surface elevation and elevations	
420	changes are then estimated by taking the weighted mean, using the number of observations as	
421	weights, for each data set giving an average error for each measurement mode, as seen in	
422	Table-2. The weighted average errors for each mode and method have been summarized in	
423	Table-1 and Table-2 for both the ESA's and our solutions, where the values for the individual	
424	campaigns can be found in the Supplementary material.	
425	The estimated surface elevation changes from the $\underline{two}$ independent methods were	
426	validated separately using near-coincident ATM data. In general we find the same magnitude of	
427	improvement observed in the surface elevation validation analysis. The statistics of the	
428	elevation change validation have been summarized in Table-2 for each method independently	
429	for the two modes of instrument operation. We find the lowest RMSE errors for the surface-fit	

Johan Nilsson 9/20/2016 10:04 AM Deleted: The observational error for the DEMmethod is estimated similarly as for the surfacefit method, with some modifications. Here, the CryoSat-2 DEM is used to de-trend the corresponding months of CryoSat-2 data consistent with the months used to derive the ATM-elevation changes (e.g., April 2011 to April-2014). However, the monthly spatial CryoSat-2 coverage does not provide adequate number of comparison points. Thus the months of March and May were included in the analysis (i.e. March-May 2011 to March-May 2014) to increase the number of samples.

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method, followed by the crossover method. This differs from the findings of Moholdt et al. (2010) 444 who found lower intrinsic errors for the crossover method, compared to the surface-fit method 445 446 when applied to ICESat data. The larger search radius used for our application of the crossover 447 method most likely explains the difference in findings between studies. Further, we find that the 448 surface-fit method provides the largest reduction in RMSE for the JPL product, corresponding to 449 40% and 55% for the SIN and LRM-mode, respectively. 450 The correlation length used to derive the number of un-correlated grid-cells, which is 451 used to estimate the standard error, was determined from a semi-variogram analysis of the 452 elevation change residuals from CryoSat-2 minus ATM using the data from the surface-fit 453 method. The comparison was done for each mode separately for all the individual campaigns 454 and multiplied with the their individual time span. The semi-variogram was then computed from 455 all the time-invariant residuals, to maximize the spatial coverage, for each mode. Analysis of the 456 semi-variogram showed an approximate correlation lengths of 100, and 75, km for the SIN and 457 LRM-mode respectively. These correlation lengths are inside the range of the ones found by 458 Sørensen et al. (2011) for their analysis of ICESat data, which was found to be 50-150 km, 459 Although the main goal of this study is not to derive or compare different types of DEM's 460 they do play a critical part in removing the long-wavelength topography in order to derive the 461 monthly time-series of volume change from the DEM-method. To gain insight into the overall 462 quality of our CryoSat-2 derived DEM (referred to as JPL) we compare it to three other DEM's 463 derived from other data sets. Firstly, we compare it to a DEM derived from ESA CryoSat-2 L2 464 data (referred to as ESA) gridded in the same manner as our DEM (Section 3.3). Secondly we 465 compare it to a DEM from Helm et al. (2014), also based on CryoSat-2 data from 2011-2014 466 (referred to as AWI). Thirdly, we compare to a DEM from Howat et al. (2014) (which was used 467 to derive topographical parameters and corrections for the JPL CryoSat-2 data), based on 468 photogrammetry data from 1999-2002 co-registered to ICESat elevation data from 2003-2009

469 (referred to as GIMP),

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These data sets were then compared to IceBridge ATM elevations, spanning the four 481 482 different campaigns previously used for validation of the CryoSat-2 elevations. The DEM 483 elevation was estimated at each ATM location, using bilinear interpolation, and the elevation 484 difference computed as (DEM-ATM). No attempt was made to account for differences in DEM 485 and ATM epochs. The estimation of the errors of the DEM was determined in the same way as 486 for the individual CryoSat-2 surface heights. The results of the comparison have been 487 summarized in Table-3, as the weighted average of the different campaigns. The values from 488 each individual campaign can be found in the supplementary material.

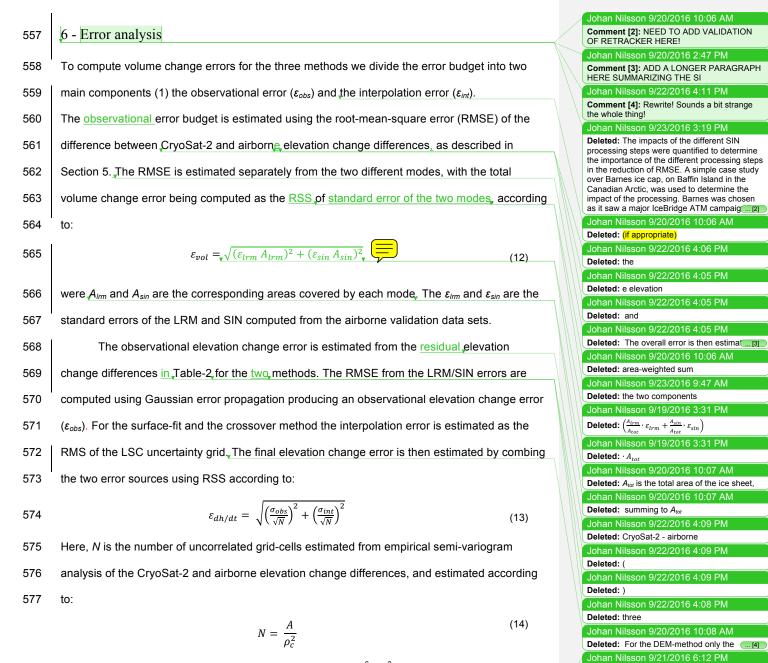
489 Analyzing the overall RMSE we find that the AWI produces the lowest RMSE, followed 490 by JPL, ESA and GIMP, due to AWI's lower standard deviation. However, the best accuracy is 491 obtained by the JPL DEM, which shows the lowest elevation bias of all DEM's. The ESA derived 492 DEM shows a slightly better standard deviation than the JPL DEM, which can be explained by 493 higher data density in the marginal areas for the ESA data. The difference in density is due to 494 the SNR rejection criterion applied in our elevation processing. This smoothing can explain the 495 lower standard deviations seen for the AWI product. The GIMP data set showed higher degrees 496 of impulse noise than the other products, explaining the higher observed standard deviation. 497 This impulse noise is attributed to that local elevation change rate, which was not accounted for 498 in the creation of the DEM (Howat et al., 2014). Overall we find that the JPL DEM provides a 499 suitable compromise between resolving of local detail and the minimization of bias. Further, 500 modification to the SNR filtering criteria will likely lead to additional improvements in the DEM. 501 To determine the effect of retracking on the accuracy and precision of the measured 502 surface heights from CryoSat-2 several tests was performed over different parts of Greenland 503 for both modes. Following the approach of Davis (1997) the accuracy (mean) and precision 504 (standard deviation) was computed as a function of leading edge threshold (in percent). This 505 computation was performed using a standard leading-edge threshold retracker, referred from 506 now on as LTH, for both the LRM and SIN mode independently. The validation was performed

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507	in the same manner as described in Section 5, where ATM elevations from 2013 was used as	
508	the surface reference.	
509	For the LRM mode data from April 2013 from the northern parts of Greenland, spanning	
510	the region 75-81°N and 54-44°W, was used to calculate height residuals for the different	
511	thresholds. This produced approximately 1000 comparison locations, which was used to	
512	calculate statistics. The same procedure was performed over Jakobshavn Isbræ, using the	
513	same time span, to calculate statistics for the SIN-mode providing roughly 2500 comparison	
514	locations.	
515	The results of this analysis, summarized in Figure-2, show that for the LRM-mode that	
516	the precision (as a function of threshold) follows the same behavior as observed by Davis	
517	(1997), with a decrease of precision following increasing retracking threshold. However, the	
518	most notable finding was the observed inverse relationship in precision for the SIN-mode	
519	compared to LRM. For LTH-algorithm, in the SIN-mode, we observe a clear increase in	
520	precision as the retracking threshold increases, seen in Figure-2, stabilizing around 30-40%.	
521	Analyzing the accuracy derived from the different thresholds a clear difference in apparent	
522	penetration depth of the radar signal can be observed for the two modes. For the SIN-mode,	
523	below 40%, a positive bias is observed indicating that retracker produces elevations larger than	
524	the corresponding airborne measured heights. For thresholds larger than 40% surface	
525	penetration of the signal is observed which are in general closer to the surface compared to the	
526	LRM-mode. We attribute this to differences in the near-surface density structure covered by the	
527	two modes.	
528	In general we conclude that for the LRM-mode that low retracking thresholds (0-30%)	Johan Nilsson 9/23/2016 3:20 PM
529	reduces the magnitude of the apparent surface penetration bias and provides higher precision	Formatted: Indent: First line: 0.5"
530	compared to higher thresholds. Therefore, a threshold of ~20% of the leading edge is	
531	suggested for retracking surface elevations for the LRM-mode, which was also previously	
532	suggested by Davis (1997) and Helm et al. (2014). However, for the SIN mode a threshold	

533	below 40% is not recommended, as this produces a clear positive elevation bias and poor		
534	precision, as seen in Figure-2. Analyzing the difference between the LTH and the adaptive LMG		
535	algorithm, used in the SIN-mode, we find that the LMG algorithm produces superior results in		
536	precision compared to the standard LTH-algorithm. Comparing the adaptive solution from LMG		
537	to the optimum threshold found by the LTH-algorithm, we find a comparable magnitude of the		
538	elevation bias and a 32% improvement in precision, with an overall 27% reduction in RMSE,		
539	using the LMG-retracker. Studying the results from this comparison between the two-retracker		
540	algorithms we recommend the use of the adaptive threshold approach, as it produces an		
541	elevation repeatability that exceeds that of the standard threshold retracker and provides a low		
542	penetration bias.		
543	A case study was also performed to determine the different processing steps affect on		
544	the quality of the retrieved observations. For this purpose the Barnes ice cap, on Baffin Island in		
545	the Canadian Arctic, was chosen due to its small size, excellent validation coverage and due to		
546	that it consist mostly of super-imposed ice (reducing radar signal penetration). The ice cap saw		
547	a major IceBridge ATM campaign in 2011 providing a large number of flight tracks (spanning in		
548	both North-South and East-West directions) suitable for validating CryoSat-2 data. The result of		
549	this case study, which is detailed in supplementary material (i.e. Table-S1) shows that the		
550	filtering of the differential phase has the highest impact on the overall accuracy of the		
551	observation, reducing the RMSE with 12%, followed by the ambiguity correction. This shows the		
552	importance of these steps, as they can have important implications for the overall quality of the		
553	retrieved elevations. This especially true in high relief areas where small changes in the look		
554	angle, or introduced phase ambiguity, can produce large elevation errors ranging from 0-100 m		
555	in elevation (Brenner et al., 1983).		

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were *A* is the total area of the Greenland Ice sheet (~1.7x10<sup>6</sup> km<sup>2</sup>) and the correlation length  $\rho_c$ of <u>75 and 100,km for the LRM and SIN mode respectively.</u>

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636	The measured surface elevations from the two CryoSat-2 products (JPL vs. ESA) showed large
637	differences in both accuracy and precision of the elevation measurements, as seen in Table-1.
638	The average accuracy and precision for the LRM-mode from the two products showed values of
639	0.00 $\pm$ 0.43 m and -1.06 $\pm$ 0.89 m for the JPL and ESA products respectively. This corresponds
640	to an average reduction in RMSE of 68% for the JPL product compared to the ESA LRM L2
641	data. Further, our product shows a lower residual slope error (seen in Figure-1c below ~0.5°),
642	indicating a lower sensitivity to the degradation of performance as the surface slope increases.
643	Surface elevations generated from the SIN-mode showed the same type of improvement
644	as for the LRM-mode. Here, an average accuracy and precision was found to be -0.52 $\pm$ 0.58 m
645	and -0.90 $\pm$ 1.05 m for the JPL and ESA SIN elevation products respectively. This further
646	corresponds to a reduction in the average RMSE of 27% for the JPL product compared to the
647	ESA product. For the SIN-mode the JPL processing produces a slightly lower residual slope
648	error, compared to the ESA processor (seen in Figure-1c above ~0.5°).
649	Larger improvements can be observed if separating the RMSE into its mean and
649 650	Larger improvements can be observed if separating the RMSE into its mean and standard deviation, corresponding to the accuracy and precision of the measurements. Using
650	standard deviation, corresponding to the accuracy and precision of the measurements. Using
650 651	standard deviation, corresponding to the accuracy and precision of the measurements. Using these definitions the analysis found that there is a 45% and 52% increase in precision for the
650 651 652	standard deviation, corresponding to the accuracy and precision of the measurements. Using these definitions the analysis found that there is a 45% and 52% increase in precision for the SIN and LRM mode respectively, compared to the ESA L2 product, and a 42% and 99%
650 651 652 653	standard deviation, corresponding to the accuracy and precision of the measurements. Using these definitions the analysis found that there is a 45% and 52% increase in precision for the SIN and LRM mode respectively, compared to the ESA L2 product, and a 42% and 99% improvement in accuracy for the respective modes.
650 651 652 653 654	standard deviation, corresponding to the accuracy and precision of the measurements. Using these definitions the analysis found that there is a 45% and 52% increase in precision for the SIN and LRM mode respectively, compared to the ESA L2 product, and a 42% and 99% improvement in accuracy for the respective modes. The estimated surface elevation changes generated from the surface-fit method also
650 651 652 653 654 655	standard deviation, corresponding to the accuracy and precision of the measurements. Using these definitions the analysis found that there is a 45% and 52% increase in precision for the SIN and LRM mode respectively, compared to the ESA L2 product, and a 42% and 99% improvement in accuracy for the respective modes. The estimated surface elevation changes generated from the surface-fit method also showed improvement in the estimated accuracy and precision, as seen in Table-2. Here, <b>a</b>
650 651 652 653 654 655 656	standard deviation, corresponding to the accuracy and precision of the measurements. Using these definitions the analysis found that there is a 45% and 52% increase in precision for the SIN and LRM mode respectively, compared to the ESA L2 product, and a 42% and 99% improvement in accuracy for the respective modes. The estimated surface elevation changes generated from the surface-fit method also showed improvement in the estimated accuracy and precision, as seen in Table-2. Here, <b>a</b> overall improvement in RMSE of 55% and 40% in the LRM and SIN mode, respectively, was

compared to 0.25 ± 1.51 m (LRM) and 0.34 ± 1.06 m (SIN) for the ESA derived changes. This 660 661 corresponds to an increase in elevation change accuracy of 56% and 12% for the LRM and 662 SIN-mode, respectively, for the JPL product compared to ESA L2 elevation changes. The 663 estimated elevation changes also show an increase in precisions for the JPL product of 56% 664 and 45% for the LRM and SIN-mode, respectively, compared to its ESA counterpart. 665 The implementation of the LMG SIN retracking algorithm was found to reduce noise in 666 the retrieved surface elevations compared to conventional threshold retracking Though roughly 667 comparable in accuracy, the LMG shows overall higher precision over all comparable leading 668 edge thresholds, The adaptive nature of the algorithm provides improved estimates of surface elevation and gives good trade-off between accuracy and precision. 669 670 The 20% threshold retracker implemented in the LRM-mode was also found to provide 671 improved estimates of surface elevation (both in accuracy and precision) compared to the model-based ESA-L2 retracker. Further, it also showed lower sensitivity to the 2012 melt event, 672 673 due to the lower threshold used on the leading edge of the waveform. 674 The estimated elevation changes of the Greenland Ice Sheet, excluding the peripheral 675 glaciers, over the period January 2011 to January 2015 show significant differences between 676 products (JPL and ESA) in both spatial patterns and the total magnitude (Figures 3, & 4). The 677 estimated volume change rate from the surface-fit method is -289 ± 20, km<sup>3</sup> a<sup>-1</sup> for the JPL-678 product and -224  $\pm$  38 km<sup>3</sup> a<sup>-1</sup> for the ESA-product with a mean difference of 65 km<sup>3</sup> a<sup>-1</sup>. The 679 surface-fit and crossover-method produced on the order of ~20 million and ~2.5 million usable 680 elevation changes, respectively, providing high spatial sampling. Due to the constraint put into 681 the JPL processor the ESA L2 data produced slightly more surface-fit observations (~10%), as 682 more surface elevations were accepted. 683 The ESA product produces a more positive elevation change pattern, which can be

684 attributed to the 2012 melt event that introduced a large positive bias with a magnitude of ~0.5 685 m (Nilsson et al., 2015). Larger differences in the marginal areas for the surface-fit methods are Johan Nilsson 9/27/2016 10:58 AM Deleted: the Johan Nilsson 9/27/2016 10:59 AM

Formatted: Not Highlight Johan Nilsson 9/27/2016 10:56 AM Deleted: This is exemplified over the Jakobshavn Isbræ area of the Greenland Ice Sheet (Figure ), where LMG and a leading edge threshold retracker were compared Johan Nilsson 9/27/2016 10:56 AM Deleted: . Johan Nilsson 9/27/2016 10:56 AM Deleted: has a 32% improvement in precision when, compared against elevations from airborne laser altimetry

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Nilsson et al., 2016

700	also observed. These are particularly noticeable in eastern Greenland (near 73.5 degrees in		
701	latitude Figure 3) where the ESA data shows marginal areas of rapid thinning that are not visible		
702	in the JPL solution. The positive signal detected in the interior of the ESA surface-fit-solution	Johan Nilsson Deleted: 2	
703	can also be found in the basin time series, correlating well with the timing of the summer of		
704	2012 melt event. These results are in agreement with earlier work demonstrating the sensitivity	Johan Nilsson Deleted: can al solution	
705	of the ESA retracker to the changes in the volume/surface scattering ratio (Nilsson et al., 2015).		
706	The two volume change methods produce consistent results from JPL derived elevation		
707	changes, with a difference of around, 1, km <sup>3</sup> a <sup>-1</sup> . The spread between volume change methods is	Johan Nilsson Deleted: three	
708	larger (50 km <sup>3</sup> a <sup>-1</sup> ) when using ESA L2 data. The larger discrepancy can mostly related to the	Johan Nilsson Deleted: less the	
709	sensitivity of the various methods to the melt event. The surface-fit method produces the most	Johan Nilsson Deleted: 7	
710	negative number (least affected by the melt event and the lowest estimated error) and is		
711	therefore taken as the most reliable estimate for both the JPL and ESA solution.		
712	Comparing the estimated volume change to other studies using CryoSat-2 we find that		
713	the JPL product is less negative than that estimated by Helm et al. (2014): -375 $\pm$ 24 km <sup>3</sup> a <sup>-1</sup> .		
714	This difference can be attributed to difference in processing methodology and to the different		
715	epoch of the data used by Helm et al. (2014) of January 2011 to January 2014. Using the		
716	corresponding epoch the JPL data gives a volume change estimate, based on the surface-fit		
717	method, of -353 $\pm$ 21 km <sup>3</sup> a <sup>-1</sup> , well within the stated uncertainty of Helm et al. (2014).		
718	To examine the regional behavior of volume change estimates of the Greenland Ice		
719	Sheet, gridded values from the three methods were divided into 8-drainage basins according to		
720	Zwally et al. (2012). When analyzing the volume change time-series at the basin scale clear		
721	differences can be observed in the annual and inter-annual behaviors (Figure 4). The northern		
722	and interior basins (1, 2, 7, 8) all exhibit large differences (Table 4: 0 - 30 km <sup>3</sup> a <sup>-1</sup> ) in the	Johan Nilsson Deleted: 3	
723	estimated volume change rates due to changes in the scattering regime resulting from the 2012		
724	melt event. In the majority of the southern basins (4, 5, 6, 7), located in areas with higher		

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732	precipitation, both products show good agreement in both trends and seasonal amplitude	
733	estimated from the surface-fit, method.	
734	The amplitude of the seasonal signal (Equation 4) estimated from the surface-fit (SF)	
735	method show large differences in both magnitude and spatial variability (Figure 5). For the	
736	surface-fit method a difference in amplitude of 54% is observed between the ESA and JPL	
737	products, corresponding to area-averaged amplitude of 0.17 m or the JPL product and of 0.37 m	
738	for ESA product. The comparison with ICESat derived amplitudes from 2003-2009 estimated in	
739	(Sasgen et al., 2012) using the same methodology as used here produced an area-averaged	
740	amplitude of 0.13 m, which is in good agreement with the JPL derived amplitude. This	
741	agreement is also spatially consistent, as seen in (Figure 5), indicating low sensitivity to	

742 seasonal changes in scattering regime of the upper snowpack. The observed difference in

743 amplitude bias, taking ICESat as the true surface amplitude while acknowledging that no inter-

744 campaign bias has been applied and the differences in epochs, is 0.03 ± 0.13 m for the JPL

745 product and 0.21 ± 0.27 m for the ESA product. The smallest differences are observed at high

746 altitudes above 2000 m a.s.l., where the three data sets show almost constant amplitude of 0.1

747 m (ICE/JPL) and 0.2 m (ESA), providing a factor of two larger amplitude for the ESA product.

748 Below 2000 m a.s.l., corresponding well to the equilibrium-line-altitude (ELA) of the Greenland

749 Ice Sheet, a rapid increase in amplitude is observed for all products. This is especially true for 750 the ESA product, which increases its magnitude by a factor of two.

751 Analyzing the amplitude patterns on a regional drainage basin level (Figure 5c) we find 752 good agreement between JPL CryoSat-2 and ICESat amplitude with ESA data producing 753 consistently larger amplitudes. Regionally, the highest amplitudes can be observed in the SE of 754

Greenland in basins (3,4,5) and are consistent with regional precipitation patterns that show

high average precipitation in these areas (Bales et al., 2009; Ettema et al., 2009).

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756 The seasonal phase of the peak in amplitude of the seasonal cycle is shown in (Figures

757 5b and 5c) and shows generally good agreement between the two products, providing the

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762	timing of the maximum of the accumulation signal, before the onset of melt, to the months of	
763	June/July for both JPL and ESA CryoSat-2 data sets. The ICESat derived seasonal phase	
764	shows a higher dependence on elevation where the maximum of the accumulation signal is	
765	found in late May below 2000 m and late July/August above 2000 m in elevation. The ICESat	
766	discrepancies from the CryoSat-2 data are found in specific basins. Disagreements between the	
767	retrieved phase of the peak amplitude from Cryosat-2 and ICESat data are due to differences in	
768	temporal sampling as discussed in more detail in Section 8.1.	
769	We used ICESat and CryoSat-2 derived surface heights to generate time series over	
770	three regions in Northeast area of Greenland (Zachariæ Isstrøm, Nioghalvfjerdsfjorden and	
771	Storstrømmen glaciers) for comparison purposes These areas have in recent time shown large	
772	and rapid changes, which has been noted by, e.g., Khan et al. (2014). The selected areas were	
773	defined using hydrological basins derived by Lewis and Smith (2009), seen in (Figure 6), and	
774	l were further divided into smaller areas around the termini to highlight performance for areas of	
775	rapid change. The ICESat and CryoSat-2 surface heights were then used to generate annual	
776	time-series from 2003-2015 using (Equation 6) in the surface fit method. The estimated 12 year	
777	time series show overall comparable elevation change rates over both time periods, especially	
778	in the terminus areas, providing confidence that CryoSat-2 can actually monitor changes in	
779	these areas.	
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### 786 8 - Discussion

787 The CryoSat-2 processing methodology presented here is found to produce accurate and 788 precise measurements of ice sheet elevation and elevation change. The main improvements 789 have been introduced in the SIN processor with the inclusion of a new type of land ice retracker 790 (LMG), advanced phase filtering and the inclusion of a phase ambiguity correction scheme. This 791 processing approach decreased the RMSE in the surface height retrieval by approximately 27% 792 (45% and 42% improvement in precision and accuracy). This improvement further propagated 793 into the quality of the estimated elevation changes for the SIN-mode, with the same magnitude 794 of improvement (Table-2). The described SIN-processing also generated surface elevations and 795 elevation changes with lower sensitivity to the local surface slope, indicating a higher degree of 796 accuracy in the geo-location and surface range estimation.

797 The SIN processing methodology further includes a phase filtering and phase ambiguity 798 correction scheme. Visual inspections of a large number of tracks have shown more coherent 799 estimation of the surface locations in our product and further the implementation of the phase-800 ambiguity correction greatly reduced the number of track offsets. It was also noted that a 801 relatively course DEM (~1 km) could be used to resolve phase ambiguities. The detection and 802 correction of phase ambiguities are relatively straightforward and rely mostly on the relative 803 accuracy of the DEM. The implementation of the phase ambiguity correction is particularly 804 important when monitoring smaller ice caps and outlet glaciers, where frequent and large track 805 offsets can bias the estimation of the underlying topography, 806 The new LRM processing methodology focused on optimal retrieval of surface 807 elevations over the interior parts of the ice sheet. Here the choice of retracking threshold has 808 proven to be the critical factor to acquire high quality surface elevations and elevation changes. 809 The choice of 20% leading edge threshold level reduced the sensitivity to changes in the

810 scattering regime for low slope, high elevation areas. The functional-based retracking algorithm

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Deleted: The statistical effects of these corrections have been analyzed over the Barnes Ice Cap on Ellesmere Island in the Canadian Arctic and are available in Table S1.

815	used in the ESA LRM processor corresponds roughly to a 50% threshold level (Wingham et al.,
816	2006a), which appears to suffer from a higher sensitivity to changes in the scattering properties
817	(volume scattering) of the near-surface firn, as the range is reference higher up (later in time) on
818	the leading edge of the waveform. This effect can be seen in (Figure 2a), and that the observed
819	l negative elevation bias (Table-1) for ESA-LRM (-1.0 m) fit well with the bias for the 50% LRM
820	threshold value shown in Figure 2a. This makes the algorithm more sensitive to annual and sub-
821	annual changes in snow-packs volume/surface scattering ratio, which can produce spurious
822	changes in elevation due to changes in the near surface dielectric properties. This is clearly
823	shown in patterns of ESA product derived elevation changes (Figure 3) where a large elevation
824	bias was introduced by the 2012-melt event (Nilsson et al., 2015a). The 20% threshold is less
825	sensitive to these types of changes (Table 1 & 2) and is in agreement with previous work that
826	has demonstrated that the 20% threshold best represents the mean surface inside the footprint
827	when exposed to a combination of surface and volume scattering (Davis, 1997).
828	Surface elevation changes, derived from multi-temporal radar altimetry observations, are
829	typically corrected for their correlation to changes in the radar waveform shape. This is to
830	reduce the effect of changes in the volume/surface scattering ratio of the ice sheets surface
831	(Davis, 2005; Flament and Rémy, 2012; Wingham et al., 2006b; Zwally et al., 2005). This
832	inherently adds to the complexity of the processing and analysis, introducing new biases and
833	error sources in the estimated parameters. For the processing approach presented here many
834	of these steps can be omitted or reduced, as they are an inherent part of the improved
835	waveform retracking. There have been attempts to remove spurious step-changes in elevation
836	resulting from sudden changes in surface scattering characteristics (caused by the 2012 melt
837	event) apparent in the ESA Baseline-B L2 data through post-processing strategies (Nilsson et
838	al., 2015c), but such approaches spread the bias over a longer period of time making the
839	"jumps" less noticeable in the time series by removing the step-change but introduces longer-

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١	Johan Nilsson 9/23/2016 3:38 PM
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Johan Nilsson 9/23/2016 3:25 PM Deleted: 2b 845 timescale bias of equal magnitude as the scattering layer is buried by less reflective snow and846 low-density firn.

847 The result of the validation procedure shows a larger slope dependent bias in the ESA 848 data, both in the elevation and elevation changes (Figure-1). This is especially true for the 849 surface elevations, which can be seen in the figures of precision and accuracy (Figure 1a and 850 1c), where both figures show clear linear slope for the ESA surface heights. In comparison, 851 estimated elevations from JPL-product show relatively stable statistics over the entire slope 852 range above 0.2°. The validation of the estimated surface elevation changes, seen in (Figure 853 1b) and (Figure 1d), shows the effect of the 2012 melt event on the ESA derived elevation 854 changes below 0.2°. Further, the accuracy of the ESA derived changes show a clear negative 855 trend as function of increased surface slope. The derived precision of the surface elevation 856 change increases dramatically above 0.5°, as more complex topography is measured. 857 The JPL CryoSat-2 processing methodology produces seasonal amplitudes that are in 858 good agreement with those derived from ICESat data, further indicating the processors abilities 859 to track real and physical changes of the ice sheets surface. The current ESA implementation 860 produces noisier estimates of elevation change, as indicated by the larger standard deviations 861 of the residuals in the ESA solutions for the surface-fit and crossover-method. Figure 5, further 862 shows an amplitude bias in the ESA data compared to the corresponding ICESat reference 863 amplitudes. The bias is constant above the Greenland ELA located around 2000 m in altitude 864 but increases linearly as elevations decrease below this. The linear increase in amplitude 865 seems to be connected to the higher and more variable precipitation in the ablation zone where 866 changes in the variable snow cover produces changes in apparent surface height. This is less 867 prominent for the JPL SIN and LRM retrackers. The estimated seasonal phase in Figure 5c and 868 5d show that both JPL and ESA CryoSat-2 elevation products can adequately resolve the 869 seasonal maximum of the accumulation signal. Both products provide a timing of the maximum 870 to the month of July over the entire ice sheet, independent of elevation. Assessing the CryoSat-

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875	2 derived maximum one does however notice a difference between CryoSat-2 and the	
876	reference ICESat dataset. This constitutes roughly a $\pm 1$ month difference depending on the	
877	7 elevation and the location. The cause of this difference can be attributed to the temporal	
878	sampling of the ICESat mission. During the mission, due to degraded laser lifespan, data was	
879	only collected in campaign mode during the spring and winter times corresponding to roughly	
880	two months of measurements for each period. When the CryoSat-2 data was resampled to	
881	1 coincide with the ICESat temporal sampling the same elevation and spatial pattern in the phase	
882	of the maximum seasonal amplitude was observed as determined from the ICESat data. No	
<mark>883</mark>	3 corresponding change in amplitude was observed. To mimic the temporal sampling of ICESat	
884	(the each year of the CryoSat-2 data was resampled using the total number of unique months in	
885	the ICESat campaign record. This as the specific months used in the ICESat sampling changes	
886	(with different campaigns.)	
887	The $\underline{two}_{r}$ independent methods used to estimate the volume change of the Greenland Ice	
888	Sheet produce consistent volume change estimates. This was especially true for volume	
889	changes derived from the JPL elevations, with a discrepancy of less than 1, km <sup>3</sup> a <sup>-1</sup> between	
890	methods. The two methods provided the same estimate of integrated volume change but the	
891	use of the surface-fit is recommended as it produces higher spatial sampling compared to the	$\overline{\ }$
892	crossover-method and lower errors. The good agreement between the methods further	
893	indicates a strong reliability in the estimated volume change rates of the Greenland Ice Sheet	
894	over the four-year period. It also shows the ability of CryoSat-2 to capture both small and large-	
895	scale spatial patterns in the rugged topography along the coastline and in the interior of	
896	Greenland. This is especially true in the major outlet glacier systems (e.g., Zachariæ Isstrøm,	
897	Nioghalvfjerdsfjorden and Storstrømmen).	
898	Studying the northern parts of the Greenland Ice Sheet we find that CryoSat-2 captures	
899	both intricate and complex behavior in the marginal areas of the ice sheet. This is exemplified in	
900	the NE regions of Greenland (Figure 6) near Zacahariæ Isstrøm, Nioghalvfjerdsfjorden and	

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Johan Nilsson 9/19/2016 1:22 PM Deleted: Zacahariae Isstrøm, Nioghalvfjerdsbrae (N79) and Storstrømmen Johan Nilsson 9/20/2016 11:01 AM Deleted: 6 Johan Nilsson 9/19/2016 1:23 PM Deleted: Nioghalvfjerdsbrae

910 Storstrømmen, which all show complex and localized patterns of elevation change. Here, 911 Nioghalvfjerdsbrae shows very small changes in elevation during the observational time-span, 912 while Zacahariae Isstrøm, its major neighbor shows large negative trends in elevation change. 913 The observed behavior agrees with the observations made in recent studies by Khan et al. 914 (2014) and Mouginot et al. (2015) who document rapid retreat and drawdown of the ice-front 915 position of the two systems beginning in 2012. Storstrømmen outlet glacier system also appears 916 to show signs of rapid thinning at low elevations near the ice-front position while a large positive 917 signal is observed roughly 100 km upstream of the terminus. This pattern has also been 918 observed by Joughin et al., (2010) and Thomas et al., (2009), using airborne altimetry and 919 surface velocity mapping. Rates of elevation change from ICESat and CryoSat-2 data show 920 good agreement in basin-scale trends (Figure 6b,c). 921 The observed volume change rates estimated from this study are within the range of 922 previous studies, ranging from -186 to -309 km<sup>3</sup> a<sup>-1</sup> for the time period 2003-2009, summarized 923 by Csatho et al. (2014). A more recent study by Helm et al. (2014: -375 ± 24 km<sup>3</sup> a<sup>-1</sup>) agrees 924 within uncertainties when differences in observation periods (2011 - 2014 vs. 2011 - 2015) are 925 taken into accounted. Assuming no changes in firn air content over respective study periods and 926 an ice density of 917 kg m<sup>-3</sup> we compare estimated changes with corresponding estimates of 927 mass change our estimated rate of Greenland glacier volume change. An assessment of 928 changes in firn air content is out of the scope of this paper. Velicogna et al. (2014) estimated 929 mass loss using the Gravity Recovery and Climate Experiment satellites (GRACE) over the 930 time-period 2003-2013 provided (converted from mass) a rate of -305  $\pm$  63 km<sup>3</sup> a<sup>-1</sup> for the 931 Greenland which is inclusive of changes in Ice Sheet and peripheral glacier ice mass (-41 ± 8 932 km<sup>3</sup> a<sup>-1</sup>, Gardner et al., 2013). The estimated volume change of -265 km<sup>3</sup> a<sup>-1</sup> from Csatho et al. (2014) and the estimated rate of -305 km<sup>3</sup> a<sup>-1</sup> from Velicogna et al. (2014) spans our estimated 933 934 rate of -289 km<sup>3</sup> a<sup>-1</sup>.

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940 9 – Summary and Conclusion

941 We conclude that the use of an adaptive retracker for the SIN-mode, based on the maximum 942 gradient method, and the use of 20% threshold retracker for the LRM-mode provide improved 943 performance to the retracker currently used for the ESA L2 elevation products. It is further 944 important, especially for the SIN-mode, to apply a leading edge discriminator to identify and 945 track the leading edge of the waveform. The functional model currently employed in the ESA 946 processor has, to the author's knowledge, no such discriminator currently implemented. This is 947 important in the SIN-mode, as it often contains multiple surface returns. The single-return model 948 applied in the ESA processor will here have issues fitting a waveform containing multiple 949 surface returns resulting in retrack jitter (Helm et al., 2014). 950 Using the new CryoSat-2 processing methodology for the LRM and SIN-mode we determine the volume change of the Greenland Ice Sheet to be  $-289 \pm 20$ , km<sup>3</sup> a<sup>-1</sup> during the 951 952 period January 2011 to January 2015. The validation against airborne ATM surface elevations 953 and elevation changes showed an average improvement in the RMSE of the measured 954 elevations of 68% and 27% for the LRM and SIN mode respectively compared to ESA Baseline-955 B L2 products. The new methodology also provide improved elevation changes with an 956 reduction in RMSE of 55% and 40% for the LRM and SIN mode respectively, compared to their 957 ESA L2 derived counterparts. 958 The methodology also showed less sensitivity to changes in near-surface scattering 959 properties than equivalent ESA products. The new processing methodology showed little effect 960 of slope-induced errors, providing better performance in the marginal areas of the ice sheets. 961 These improvements to the CryoSat-2 processing mitigate the need for post-processing to 962 correct correlations between changes in surface elevation and changes in the waveform shape 963 (i.e. backscatter and leading edge width etc.) that can introduce biases and add to the

964 complexity of the processing and analysis.

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966	The presented CryoSat-2 processing methodology provides a lower intrinsic error in the	
967	measured elevation, elevation change and volume change estimates, all of which will facilitate	
968	improved understanding of the geophysical process leading to changes in land ice elevation.	
969	Given the release of the ESA Baseline-C, which provides improved corrections and processing	
970	mainly for the L1B product, further improvements are expected in the near future.	Johan Nilsson 9/27/2016 1:14 PM Deleted: pending
971	The complete set of grids used in this study is available for the public from the main author	
972	(J.Nilsson) upon request and are provided in geotiff format.	
973	Acknowledgement	
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982	with NASA.	
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1153 Tables:

1154 Table 1: Accuracy (Mean), precision (SD) and the total RMS-error (RMSE) of surface elevation

1155 from CryoSat-2 observations compared to IceBridge ATM elevations. Here, the LRM mode

1156 represents the interior of the ice sheet and SIN the marginal high relief areas.

JPL	Mean (m)	SD (m)	RMSE (m)
LRM	0.00	0.43	0.45
SIN	-0.52	0.58	0.82
ESA	Mean (m)	SD (m)	RMSE (m)
LRM	-1.06	0.89	1.40
SIN	-0.90	1.05	1.13

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1160 Table 2: Accuracy (Mean), precision (SD) and the total RMS-error (RMSE) of surface elevation

1161 changes from CryoSat-2 derived from two\_independent methods [Surface Fit (SF) an the

1162 Crossover (XO) <u>method</u>, compared to IceBridge ATM data.

JPL - LRM	Mean (m)	SD (m)	RMSE (m)
SF	0.11	0.67	0.70
XO	0.24	0.72	0.78
ESA - LRM	Mean (m)	SD (m)	RMSE (m)
SF	0.25	1.51	1.57
хо	0.60	1.02	1.20
JPL - SIN	Mean (m)	SD (m)	RMSE (m)
SF	0.30	0.58	0.66
ХО	-0.60	1.26	1.26
ESA - SIN	Mean (m)	SD (m)	RMSE (m)
SF	0.34	1.06	1.11
ХО	-0.21	1.44	1.44

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- 1174 Table 3: Validation of four different DEMs, compared to IceBridge ATM elevation data. Based
- 1175 on the weighted (number of samples) average of the four different ATM campaigns from 2011 to
- 1176 2014. Elevation values at each ATM location were estimated by bilinear interpolation for each
- 1177 DEM product.

DEM	Mean (m)	SD (m)	RMSE (m)
AWI	-1.35	5.95	6.12
GIMP	-1.13	7.22	7.32
JPL	-0.87	6.31	6.39
ESA	-2.83	6.13	6.76

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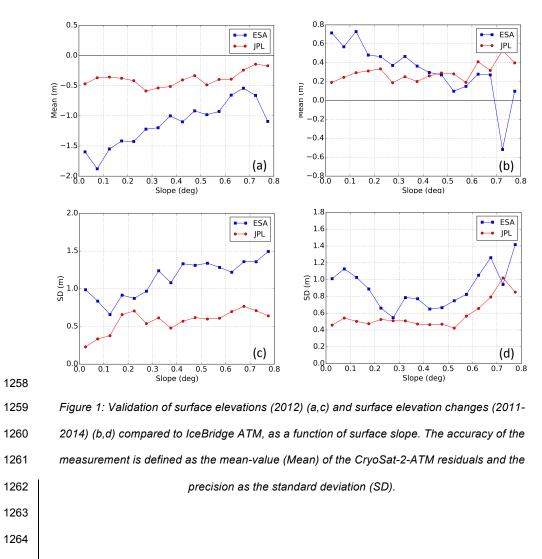
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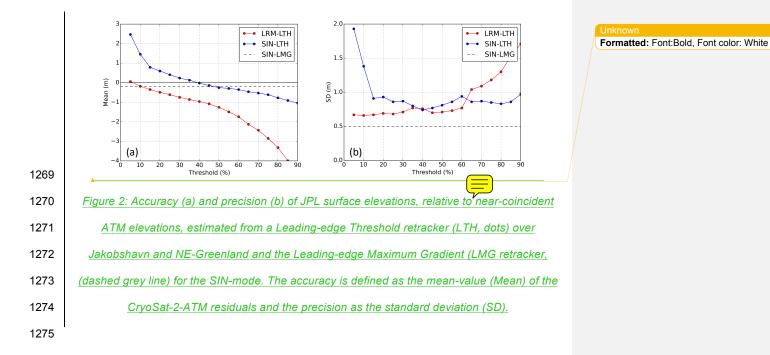
- 1180 Table 4: Individual basin volume changes (km<sup>3</sup>a<sup>-1</sup>) for the Surface-Fit (SF) and Crossover (XO)
- 1181 *method for the JPL and ESA product for the time period Jan-2011 to Jan-2015, with*
- 1182 corresponding volumetric error.

Basin	SF – JPL	XO – JPL	SF - ESA	XO - ESA	
1	-26 ± <u>8</u> ,	-23 ± <u>12</u> ,	-9 ± <u>14</u> ,	-11 ± 1 <u>5</u> ,	
2	5 ± <u>8</u> ,	0 ± <u>13</u> ,	31 ± <u>16</u> ,	30 ±1 <u>6</u> ,	
3	-38 ± <u>9</u> ,	-34 ± <u>19</u> ,	-46 ± <u>16</u> ,	-31 ± 2 <u>3</u> ,	
4	-36 ± 7,	-37 ± 1 <u>5</u> ,	-42 ± <u>12</u> ,	-16 ± 1 <mark>8</mark> ,	
5	-19 ± <u>4</u> ,	-27 ± <u>11</u> ,	-19 ± 7,	-6 ± <u>13</u> ,	
6	-72 ± 7,	-71 ± 1 <u>2</u> ,	-75 ± 1 <u>3</u> ,	-79 ± 1 <u>8</u> ,	
7	-56 ± 7,	-51 ± 10	-41 ± 14	-35 ± 1 <u>5</u> ,	
8	-48 ± <u>8</u> ,	-45 ± 1 <u>2</u> ,	-23 ± 1 <u>5</u> ,	-27 ± 1 <u>7</u> ,	
тот	-289 ± <u>20</u> ,	-288 ± <u>37</u>	-224± 3 <u>8</u> ,	-174 ± 4 <u>8</u> ,	

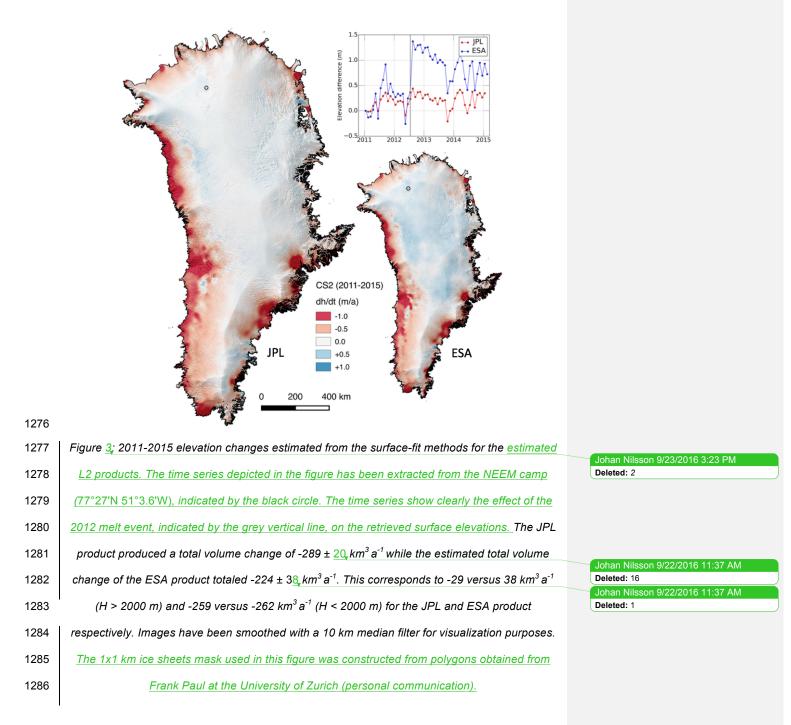




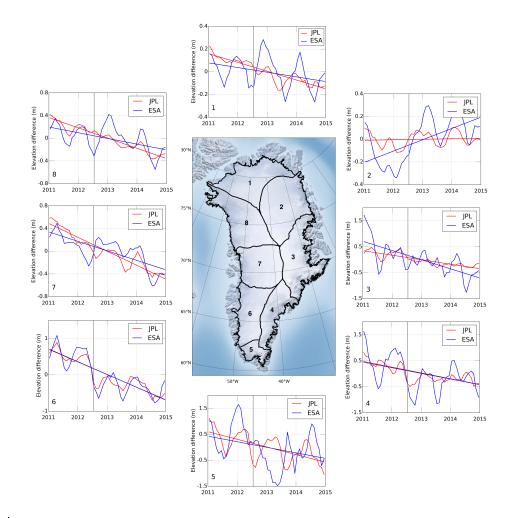




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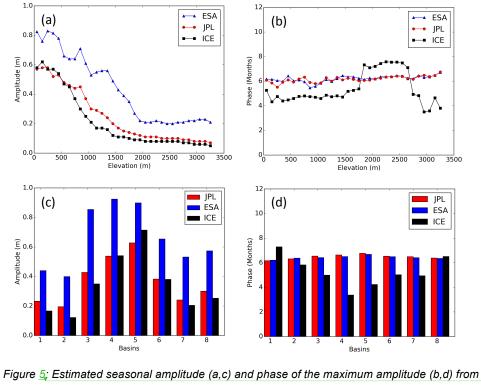
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1291Figure 4, Monthly elevation change time-series for 8 large drainage basins of the Greenland Ice1292Sheet. Time-series have been smoothed using a 3-month moving average for improved1293visualization. The grey vertical line indicates the timing of the 2012 melt event.

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the surface-fit method for CryoSat-2 [ESA (blue) and JPL (red)] compared to ICESat (ICE,
black)). Values are compared using a search radius of 500 m, using the closest point within this
distance, and the phase offset is referenced from 1<sup>st</sup> of January. The values of amplitude and
phase are then binned according to elevation using the median value within 100 m intervals.

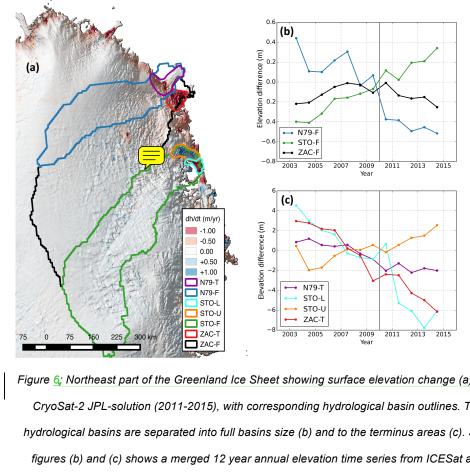
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1311	Figure <u>6</u> ; Northeast part of the Greenland Ice Sheet showing surface elevation change (a) from
1312	CryoSat-2 JPL-solution (2011-2015), with corresponding hydrological basin outlines. The
1313	hydrological basins are separated into full basins size (b) and to the terminus areas (c). Sub-
1314	figures (b) and (c) shows a merged 12 year annual elevation time series from ICESat and
1315	CryoSat-2 for each color-coded area in (a). The derived elevation time series was formed using
1316	the surface-fit method described in Section (3.1). The elevation change map is overlaid onto the
1317	CryoSat-2 hill shaded DEM based on surface heights from Jul-2010 to Feb-2015.
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### Response to reviewer #1 (L. Schröder)

We would first like to thank the reviewer for his constructive and insightful comments, which has greatly helped to improve this manuscript.

The reviewer's remarks are in bold font, while the author's response is in italic form below.

### **General Comments:**

The reviewer has in his analysis of the paper asked for more substantial proof of the sensitivity of the two retrackers used in this study (threshold versus functional-fit) to changes in snow-pack properties. To address this we have merged Figure (2 & 4) and expanded it to include a local time-series for the NEEM camp, which encompasses the time of the melt event. This figure clearly shows the effect of the change in snow-pack conditions at time of the 2012 melt event, and the subsequent introduction of a clear elevation bias in this case the ESA retracker, with a magnitude of  $\sim$ 1 m, which can be seen in Figure-2

In the case of the residual slope bias/error we have chosen to remove this metric. This was done to make the elevation and elevation change comparison more coherent. In the case of the elevation a clear linear relation, depended on topography and snow physics, can be found (penetration bias and precision in the retrieval of measurements) and easily quantified using a linear model. However, this does not hold for the elevation change over time, and further the slope-error is not easily characterized by a linear trend (or higher polynomials). This due to the 2012 melt event effect on the standard deviation in low sloping areas (0-0.3°) for the ESA product. So, we have chosen to remove this metric and guide the reader to Figure-1, where these different effects can clearly be seen and judged visually.

### I.12: Expr: This when compared

Changed the statement by merging it with the previous sentence.

# I.41: add brackets around "(e.g. Gardner et al., 2013; Shepherd et al., 2012)"

Brackets where added around the citation.

### I.44: Expr: "the characteristics of which is"

Removed the expression from the sentence

#### I.60: Expr: "methods from improving"

The sentence was changed where "from" was replaced by "for".

### I.77: I think 1° is no "low sloping terrain" anymore for radar altimetry. The switch to SARIn happens already at lower slopes (~0.5°).

[A] The reviewer has a good point and we have therefore removed the numbers to make the statement more generic.

## I.101: In Baseline-B LRM has only 128 bins so I think the interval should end some bins before.

Perfectly true! This statement belongs to the SIN retracker and was unfortunately overlooked. For the LRM-mode only peak indexes larger than 20 are used in the retracking procedure.

#### I.126: don't use the surname in the citation

This has been removed accordingly.

#### I.128: SIN mode allows... repeats more or less the last sentence

The sentence was re-written to remove any repetitive nature.

## I.142ff: This is not totally clear to me. Please explain a bit more in detail what the "coherence range power image" is.

This was changed to "Coherence as a function of range", e.g. across track coherence (an array of size Nobs\*512) for each measurement in the track.

#### I.205ff: Please use different letters for different variables (not again a0,a1).

This has also been pointed out by reviewer #2 and has been change accordingly in the entire manuscript.

I.323: I guess no ICESat campaign biases have been applied as in Nilsson et al., 2015b. Maybe the influence of those biases (~10 cm) on the seasonal amplitude and phase is not too big, but anyways this should be mentioned and discussed when taking ICESat as a reference for the "true surface amplitude" (I.551)

This is a good point and we have added a paragraph detailing this in the ICESat section. Further, a statement was also added in line 577 detailing that the correction has not been applied.

### I.407: Why has no attempt been made? Please explain!

Quantifying the accuracy of the different DEM's is inherently difficult, as they are based on both different types of datasets and acquired over different time spans, many not entirely consistent with the temporal coverage of the ATM data. For this case the interest was to compare them in a relative fashion to judge their quality. It's of course expected that older DEM's would show a larger statistical difference from the ATM data used in the comparison. However, as stated in the manuscript, we do not attempt to provide a full validation framework of the different DEM's, only a relative comparison.

### I.425: repetition: processing steps

Has been removed to improve reading

I.692: remove "by"

Removed

# In Tab.2 as it summarizes results of elevation change I guess the units shall be m\*a^-1.

The units on the table are correct, as we have chosen to multiply it with the timespan of each elevation change data set. This was done to make the statistics/errors more comparable between time periods, as the error is expected to decrease with time.

### **Response to reviewer #2 (Anonymous)**

Firstly we would like to thank the reviewer for taking the time to review our manuscript. We are thankful for the insightful and constructive comments that have been provided, which we feel will improve our manuscript.

The reviewer's remarks are in bold font, while the author's response is in italic form below.

### General remarks:

- (1) In response to the rational of using different elevation change methods we have included a sentence in the introduction describing the choice of different methods. Further, we have also chosen to remove the DEM-method from the manuscript, as we feel that this method does not bring any new insight to the study and can be reproduced by the SF-method (as seen with the ICESat/CryoSat time series). We hope that this will make the manuscript more concise.
- (2) To produce a manuscript of reasonable length we chose to keep the general description of the processing chain generic and highlight the details that make the processing scheme unique but agree that more details on the retracker in the main document would be beneficial. We have therefore moved the retracker comparison from the SI to the validation section of the main manuscript. For the curious or more technical reader we have supplied reference that provides a more technical description of the algorithms used.
- (3) We agree with the reviewer that the notations in the different equations have to be improved. This has been changed accordingly through out the manuscript. Please see track changes version of the revised manuscript.

### **Detailed remarks:**

Describing the error of the volume change estimations (lines 436-446) the authors treat the errors as systematic errors rather than random errors and thus overestimate the volume change errors.

We have changed Equation 12 to correct for this and we are grateful that this was pointed out to us. All volume change errors have been updated to encompass the corrected error propagation.

The error of the elevation change (lines 447-459) describes the error of the mean elevation change of the entire ice sheet rather than the error of a

single elevation change estimate. This error is not referred in the manuscript.

The single-observation uncertainty, or  $\sigma_{obs}$ , is estimated from the CryoSat-2 – ATM residuals as the RMSE. The derivation of this error source is described in the "Error Budget" section on line 583. We have rewritten this section to make it clearer how the single observation error and interpolation errors are defined.

Lines 327-333: what ATM products were used for the study and from where were those obtained? NSIDC distributes both individual ATM footprint locations and average ice sheet elevations for larger regions (ICESS). Ice sheet elevation accuracies are 0.071-0.085 m according to Krabill et al., 2002 – more like 0.1 m than cm level as quoted in the manuscript.

The ATM data obtain from the NSIDC was the ILATM2 product (IceBridge ATM L2 Icessn Elevation, Slope, and Roughness, Version 2), which contains the measured surface elevation, slope and roughness for each measurement averaged to 80 resolution with 40 m spacing. We have changed the manuscript to reflect this, where we have put in the source of the data and the accuracies.

## Lines 463-491: this section provides a verbal description of tables. Adding the percentage of improvement would be more informative.

If we understand correctly the reviewer asks for the percentage after each numbered value? This, as far as we believe, has been met, as the percentage values for RMSE is stated in the manuscript, which encompasses both the mean and standard deviation.

Lines 573-583 and later: please use the accepted names of these glaciers: Zachariæ Isstrøm, Nioghalvfjerdsfjorden and Storstrømmen glaciers.

This has been changed accordingly.

Lines 686-697: this seems to be a missed opportunity to emphasize the good spatial and temporal resolution of CryoSat-2 observations. The recovering surge of Storstrømmen glacier has been well documented, and additional references would improve the manuscript.

The reviewer has a good point here and we have added additional references documenting the recovery at Storstrømmen. We thank the reviewer for suggesting this improvement.

Table 2. Please include the period the elevation changes refer to Figure 2.What ice sheet mask was used to define the boundary of the ice sheet?

Figure 3. Were the monthly changes determined by the DEM method? Figure 4. There is no reference to this figure in the text. Can this figure be merged with Figure 2? Does not seem to include additional information. Figure 5. I assume that all the values here are average/mean values. If yes, this should be stated in the caption

We have added the periods to the figures showing time period of elevation change. The ice sheet mask was provided by personal communication with Frank Paul at University of Zurich. This has also been added to the manuscript.

In the case of the monthly time-series they where generated using the DEMmethod. However, as the DEM-method has been removed these have been replaced by the time-series from the surface-fit method.

Figure-2 and 4 has now been merged into one main figure to reduce the number of figures overall.

Figure-5 contains the median-values inside each 100 m elevation interval using. This has been added to the caption to make it clear