GRACE forward modeling of Gulf of Alaska water balances

ABSTRACT. ...

3 INTRODUCTION

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The GRACE mission has revolutionized our ability to monitor global water mass variability by mapping the Earths gravity field each month with a spatial resolution of 300-400 km. Applying GRACE solutions to highresolution studies, however, poses significant challenges. Due to large increases in the GRACE gravity errors at small spatial scales, the project spherical harmonic solutions must be filtered prior to their application to geophysical research. GRACE mascon estimation, initially developed by the gravity group at NASA Goddard Space Flight Center (GSFC) (Rowlands, 2005; Sabaka and others, 2010; Luthcke and others, 2013), is quickly becoming the preferred method for time-variable gravity estimation (Watkins and others, 2015). The mascon approach optimizes the solution signal to noise ratio by introducing a geophysically-11 based regularization matrix in the normal equations, which is applied across pre-defined "constraint" regions 12 exhibiting similar mass variations (e.g. oceans, major ice covered areas). This eliminates the need to design 13 and apply a post-processing filter to the GRACE solutions. However, the mascons still exhibit the same 14 fundamental spatial resolution as the spherical harmonics within a mascon constraint region. Attempts to 15 overcome the resolution limits of GRACE have been made with the estimation and application of 1x1 equal-16 angle gain factors using a combination of Terrestrial Water Storage (TWS) models and filtered GRACE 17 solutions (Landerer and Swenson, 2012). The main deficiency in this approach is the significant information 18 loss that occurs when filtering GRACE solutions to their fundamental resolution. Even though this scaled 19 GRACE TWS product is distributed at 11 degrees, it should only be analyzed after combining a sufficient 20 number of grid cells to form regions large enough to be resolved by GRACE. 21 These limitations in the spatial resolution of GRACE motivate the development of a forward modeling 22 approach. Forward modeling involves accounting for as much of the known mass variability in the Level 23 1B (i.e. the intersatellite range rate measurements) processing as possible, using independent models or 24 observations. Doing so minimizes the magnitude of the inter-satellite residuals and updates to the gravity 25 field, thereby reducing the well-known problem of temporal aliasing. Forward modeling also reduces the 26

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determined solutions. In the case of a perfect forward model, the Level 1B processing would produce 28 inter-satellite ranging measurement residuals of zero (ignoring the effects of noise and systematic errors), 29 and no updates to the gravity field would be made. It is important to note that though GRACE solutions 30 are limited in their ability to spatially resolve signals, the Level 1B inter-satellite measurements are in fact 31 32 sensitive to high temporal and spatial resolution variability, and unmodeled signals of sufficient magnitude at any resolution will manifest as non-zero updates to the mascons. 33 Since the beginning of the GRACE mission, GRACE processing centers have used forward modeling 34 to remove the effects of ocean tides, atmospheric, and non-tidal ocean mass variability. Sabaka et al., 35 2010 demonstrated the benefit of also including a TWS model for the further reduction of the inter-36 satellite measurement residuals and the mitigation of signal leakage into or out of regions of hydrologic 37 variability. Luthcke et al. [2013] expanded this idea further with the implementation of a fully iterated 38 time-variable mascon solution, where each iterative solution defines the forward model for the subsequent 39 iteration, demonstrating a simultaneous increase in signal and decrease in residual magnitude until solution 40 convergence occurred. In addition to the typical forward models listed above, the current NASA GSFC 41 global mascon solution also models TWS from Global Land Data Assimilation (GLDAS)/Noah [Rodell et 42 al., 2004], ICE-6G glacial isostatic adjustment (GIA) [Peltier et al., 2015], and the largest co-seismic events 43 [Han et al., 2013]. 44 Here we will extend the forward modeling concept by including independent estimates of hydrological 45 variations of the Gulf of Alaska (GOA) region in the GRACE Level 1B processing. We will then use the 46 GRACE residuals to adjust our estimates and conduct additional iterations. We select the GOA because 47 it has large amplitude hydrological variations minimize uncertainties in signal attribution, and because we 48 have an array of independent in situ and airborne observations with which to constrain our estimates.

portion of the estimated gravity field that is subject to the fundamental spatial resolution of the GRACE-

50 Previous Alaska GRACE studies

- Early GOA GRACE studies focused on recovering the mass trend for the entire region which was assumed to represent the contribution of GOA glaciers to rising sea level. (?) validated subsets of the NASA GSFC solution to independent altimetry Arendt and others (2013)
- The watersheds bordering the Gulf of Alaska (GOA) presently discharge approximately 850 km³ yr⁻¹ (Hill and others, 2015) of freshwater to the ocean, and as such are an excellent case study for developing new approaches for processing GRACE. Seasonal and multi-annual variations in these runoff volumes have

impacts on near shore ecosystems and ocean circulation patterns (O'Neel and others, 2015). Using a fully distributed energy balance model at 1 km resolution, Beamer and others (2016) have shown that the partitioning of GOA runoff between snow melt, glacier ice melt and rainfall during 2002-2014 was 63%, and 20% respectively.

Our proposal will leverage the Level 1B processing and global mascon estimation capabilities at NASA

61 (Lenaerts and others, 2013)

62 METHODS

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GSFC, resulting in monthly estimates of 41,168 11 arc-degree equal-area cells in terms of cm of equivalent 64 water height. As previously discussed, the true resolution of the solution (300 km) is lower than that of the 65 equal-area mascon grid (111 km), but the higher resolution mascon definition allows for a more accurate 66 representation of the land/ocean boundaries that define the constraint regions, across which mascons are 67 uncorrelated. 68 The forward model approach in the context of this study is as follows: GMELT datasets and models 69 of mass changes of the HMA (Figure 2) will define the high-resolution model in the HMA region while 70 the rest of the global model is the NASA GSFC GRACE mascon solution (all other previously described 71 components of the NASA GSFC forward model procedure are also included). Time series of mass changes 72 from GMELT will be combined into an ensemble product based on consensus across the HiMAT team. We 73 will explore the use of weighting factors to combine multiple overlapping estimates, or the testing of different 74 data/model combinations within our iterative framework. Next or best estimates will be expressed to at 75 least spherical harmonic degree and order 90, where the expansion size can be increased if it is determined 76 to be beneficial. Each new Level 1B processing produces a new global mascon solution, where the HMA 77 region mascons describe the long spatial wavelength GMELT error, which serves as feedback for the next 78 iterative construction of the GMELT water balances. Updated GMELT model output is iteratively applied 79 to new Level 1B forward model runs until the GRACE mascon updates are sufficiently close to zero, 80 indicating that the high-resolution GMELT model is in full agreement with the GRACE measurements. It may prove to be more effective and efficient to circumvent the formation and inversion of normal equations 82 by directly analyzing the range-rate or range-acceleration residuals generated from the Level 1B processing 83 as described in recent studies [Loomis et al., 2015; Eicker and Springer, 2016]. We note that our iterative 84 procedure enables adjustment of parameters on individual modules of GMELT, or on weighting factors 85 that combine the ensemble of mass change estimates. 86

NASA GSFC GRACE Level 1B processing applies a baseline orbit parameterization [Rowlands et al., 2002] that does not require the processing and reduction of GPS measurements, but instead uses the GPS-determined Level 1B navigation files to define the initial orbit that is adjusted simultaneously with the gravity from the inter-satellite measurements (also accounting for the Level 1B attitude quaternions and accelerometer measurements). This allows for the rapid formation and inversion of normal equations for the full duration of the GRACE mission. The ability to quickly process and invert new solutions is an important practical consideration for the work proposed here, where a number of different GMELT outputs will need to be analyzed and iterated.

DISCUSSION

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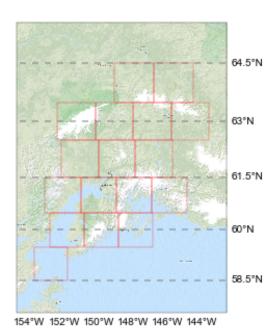


Fig. 1. Mascons of the western Gulf of Alaska.

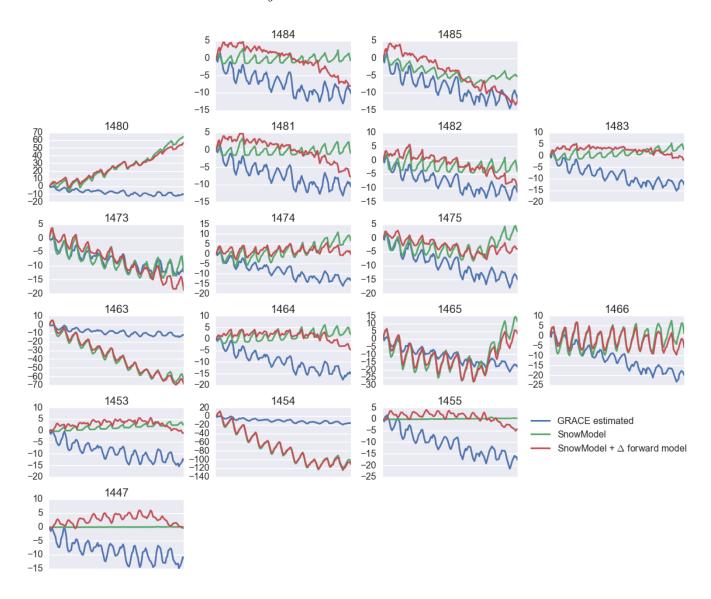


Fig. 2. Comparison of GRACE water balances prior to forward modeling (Luthcke and others, 2013), (blue); modeled water balances using SnowModel (Beamer and others, 2016) (green); and modeled water balance corrected by the forward modeling in this study (red). Subplots are arranged to approximately match the geographic layout of thier respective mascons in Fig. 1

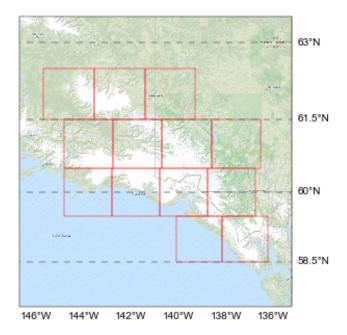


Fig. 3. Mascons of the western GOA

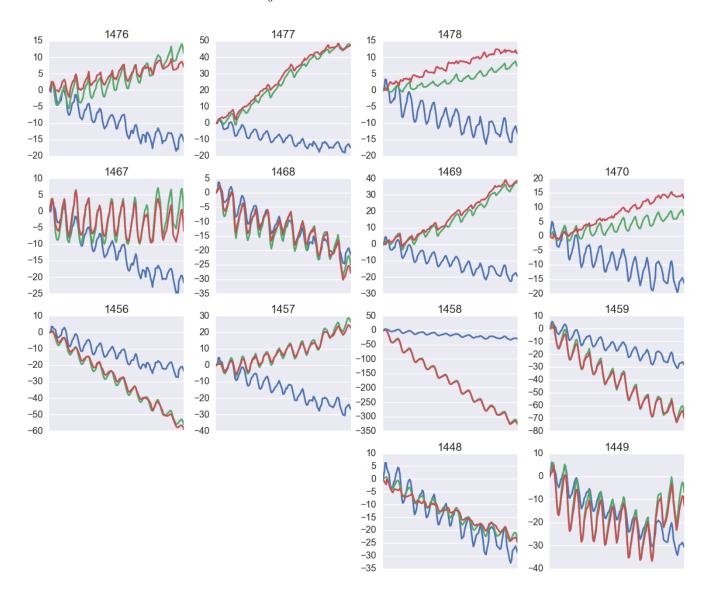


Fig. 4. Mascons of the western GOA

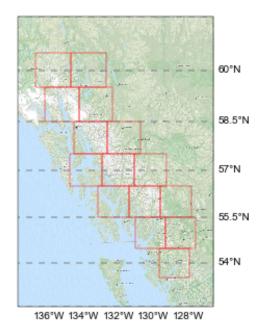


Fig. 5. Mascons of the western GOA

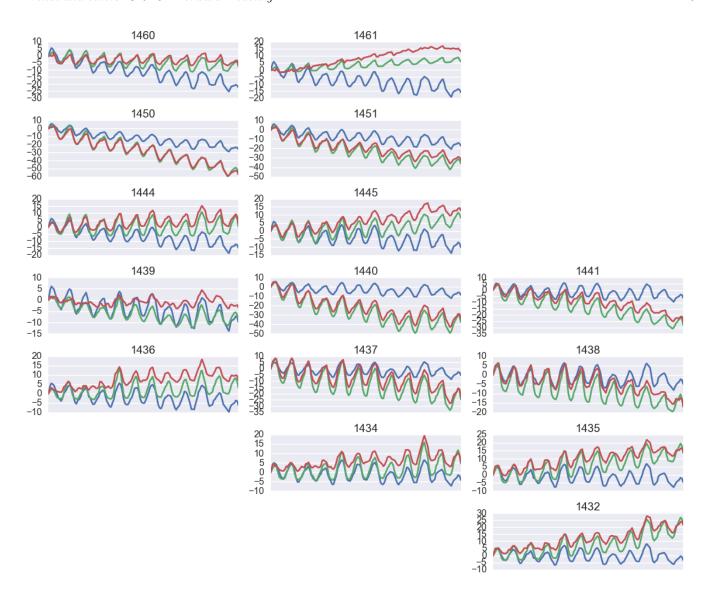


Fig. 6. Mascons of the western GOA