# GRACE forward modeling of Gulf of Alaska water balances

#### ABSTRACT. ...

#### 3 INTRODUCTION

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Data from the NASA/DLR Gravity Recovery and Climate Experiment (GRACE) have been fundamental tools for assessing mass variations of alpine and high latitude regions (Wouters and others, 2014). To date, much research has focused on the ice sheets, where the uniformity of land cover type minimizes the number of different geophysical signals that need to be partitioned in GRACE processing (Shepherd and others, 2012). In contrast, non-ice sheet (mountain glacier) regions usually exist within a dispersed mixture of land and ocean cover types that often include complex fjords, coastal temperate rainforests, lakes, rivers and seasonal snowpacks, making it difficult to partition the collection of hydrological and geophysical signals 10 that GRACE observes. The problem is compounded by the fact that many non-glacier surfaces have semi-11 annual periodicity in their mass fluctuation that mimic those of glaciers. Because of these complexities, 12 Most GRACE mountain glacier studies have focused on recovering the cumulative change in mass (Reager 13 and others, 2016) where glacier volume loss signals are assumed to be the dominant driver of changes in 14 hydrological storage. Nevertheless, there is a need for full utilization of the sub-annual components of the GRACE signal to inform studies of water resource availability, impacts of freshwater discharge on ecosystem 16 services, and predicting the likelihood of runoff-related hazards (O'Neel and others, 2015). 17 Nearly all GRACE studies aim to isolate a mass change signal of interest as a residual of GRACE 18 processing after all other systems have been accounted for through observations or models. In the case 19 of mountain glacier studies, models such as the Global Land Data Assimilation System (GLDAS, (Rodell 20 and others, 2004)) are used to remove all non-glacier sources of hydrological variations. Within GLDAS 21 are land surface hydrology models such as the Community Land Model (?) and the Variable Infiltration 22 Capacity model (?). At mid- to low latitudes, the accuracy of such models is generally good (e.g. Werth and 23 others, 2009). However at high elevation and latitude, these models are limited in their capacity to fully 24 account for changes in terrestrial snowpack, groundwater storage, lake level fluctuations and precipitation. 25 The primary problem is that such models do no have paramaterizations for glacier ice flow, and therefore 26

any solid precipitation that is not removed in a given season can accumulate indefinitely on high elevation 27 grid cells. To account for this, all GRACE studies to date have simply eliminated land surface hydrology 28 corrections from those grid cells containing glacier ice (e.g. Gardner and others, 2011; Jacob and others, 29 2012; Luthcke and others, 2013). Due to the relatively coarse resolution of the land surface hydrology 30 31 models (generally ¿25 km), this correction procedure removes significant non-glacier areas from the analysis, 32 leading to errors and mis-attribution of signal. An additional complication relates to the periodicity of mass change signals relative to the period of GRACE processing. Aliasing of daily, multi- and sub-daily cycling 33 in hydrological signals occurs when these signals are averaged over monthly time scales and then removed 34 from fully processed GRACE solutions. Additional computations are also necessary to properly compare 35 hydrological observations collected at discrete points in time with time-averaged monthly values acquired from multiple GRACE measurements over a particular region (Swenson and Wahr, 2006; Hill and others, 37 2015). 38

39 An alternative approach to partitioning GRACE signals involves

creating best estimates of the mass variations in the signal of interest are developed from independent 40 models (Beamer and others, 2016), datasets (Sasgen and others, 2012), or random algorithms (Gardner 41 and others, 2011; Colgan and others, 2013). The simulated signal is then forward modeled to a degree and 42 order that match GRACE resolution, and the observed and simulated GRACE signals are compared. Joint 43 inversion of GRACE together with GPS/ocean bottom pressure (Wu and others, 2010) and sea surface 44 height (Jensen and others, 2013) have been used to provide additional constraints that help minimize 45 uncertainty in attribution of the GRACE signal. All of these studies aim to better isolate the spatial and 46 temporal patterns in mass change of regions with complex hydrology. By forward modeling and using 47 inversion or Monte Carlo methods, they minimize uncertainties that result from smoothing and filtering 48 when removing signal from a fully processed GRACE solution. 49

Despite this progress in method development, there are several limitations to the above approaches. The
first is that independent datasets used to initialize forward modeling studies have focused on glacier volume
loss (e.g. Sasgen and others, 2012), i.e. the long-term trend in the GRACE signal. Seasonal changes have
been entirely attributed to glacier seasonal mass balances, with limited consideration of the role of terrestrial
snowpack and other annually-varying components of the system. The second is that the random generation
of a synthetic mass change curve makes it impossible to understand partitioning of mass changes among
various hydrological systems. While the full signal may be well captured by such approaches, it returns

a bulk measure of mass change with no knowledge about the individual source contributors. The third 57 problem is that the spatial resolution of existing inversion/forward modeling approaches is on the scale of 58 large watersheds, limiting our ability to use GRACE to understand processes at smaller spatial scales. 59 The watersheds bordering the Gulf of Alaska (GOA) presently discharge approximately 850 km<sup>3</sup> yr<sup>-1</sup> 60 (Hill and others, 2015) of freshwater to the ocean, and as such are an excellent case study for developing 61 62 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 63 distributed energy balance model at 1 km resolution, Beamer and others (2016) have shown that the 64 partitioning of GOA runoff between snow melt, glacier ice melt and rainfall during 2002-2014 was 63%, 65 17% and 20% respectively.

(Lenaerts and others, 2013)

#### 8 METHODS

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 70 to high-resolution studies, however, poses significant challenges. Due to large increases in the GRACE 71 gravity errors at small spatial scales, the project spherical harmonic solutions must be filtered prior to 72 their application to geophysical research. GRACE mascon estimation, initially developed by the gravity 73 group at NASA Goddard Space Flight Center (GSFC) [Rowlands, 2005; Sabaka et al., 2010; Luthcke et 74 al., 2013, is quickly becoming the preferred method for time-variable gravity estimation (e.g. [Watkins et 75 al., 2015). The mascon approach optimizes the solution signal to noise ratio by introducing a geophysical-76 based regularization matrix in the normal equations, providing a great benefit to researchers who no longer 77 need to design and apply a post-processing filter to the GRACE solutions. However, the mascons still 78 exhibit the same fundamental spatial resolution as the spherical harmonics within a mascon constraint 79 region. Attempts to overcome the resolution limits of GRACE have been made with the estimation and application of 1x1 equal-angle gain factors using a combination of Terrestrial Water Storage (TWS) models 81 and filtered GRACE solutions [Landerer and Swenson, 2012]. The main deficiency in this approach is the 82 significant information loss that occurs when filtering GRACE solutions to their fundamental resolution. 83 Even though this scaled GRACE TWS product is distributed at 11 (Figure 1), it should only be analyzed 84 after combining a sufficient number of grid cells to form regions large enough to be resolved by GRACE. 85

These limitations in the spatial resolution of GRACE motivated the development of a forward modeling 86 approach. The conceptual benefit of forward modeling is simple: by accounting for as much of the known 87 mass variability in the Level 1B (i.e. the intersatellite range rate measurements) processing as possible, the 88 magnitude of the inter-satellite residuals and updates to the gravity field are minimized. This approach 89 reduces the well-known problem of temporal aliasing and limits the portion of the estimated gravity field 90 91 that is subject to the fundamental spatial resolution of the GRACE-determined solutions. In the case of a perfect forward model, the Level 1B processing would produce inter-satellite ranging measurement 92 residuals of zero (ignoring the effects of noise and systematic errors), and no updates to the gravity field 93 would be made. It is important to note that though GRACE solutions are limited in their ability to spatially 94 resolve signals, the Level 1B inter-satellite measurements are in fact sensitive to high temporal and spatial 95 resolution variability, and unmodeled signals of sufficient magnitude at any resolution will manifest as 96 non-zero updates to the mascons. 97

Since the beginning of the mission, GRACE Level 1B processing centers have relied on forward modeling 98 to remove the effects of ocean tides, atmospheric, and non-tidal ocean mass variability. Sabaka et al., 99 2010 demonstrated the benefit of also including a TWS model for the further reduction of the inter-100 satellite measurement residuals and the mitigation of signal leakage into or out of regions of hydrologic 101 variability. Luthcke et al. [2013] expanded this idea further with the implementation of a fully iterated 102 time-variable mascon solution, where each iterative solution defines the forward model for the subsequent 103 iteration, demonstrating a simultaneous increase in signal and decrease in residual magnitude until solution 104 convergence occurred. In addition to the typical forward models listed above, the current NASA GSFC 105 global mascon solution also models TWS from Global Land Data Assimilation (GLDAS)/Noah [Rodell 106 et al., 2004, ICE-6G glacial isostatic adjustment (GIA) [Peltier et al., 2015], and the largest co-seismic 107 events [Han et al., 2013]. Our proposal will leverage the Level 1B processing and global mascon estimation 108 capabilities at NASA GSFC, resulting in monthly estimates of 41,168 11 arc-degree equal-area cells in 109 terms of cm of equivalent water height. As previously discussed, the true resolution of the solution (300) 110 km) is lower than that of the equal-area mascon grid (111 km), but the higher resolution mascon definition 111 allows for a more accurate representation of the land/ocean boundaries that define the constraint regions, 112 across which mascons are uncorrelated. 113

The forward model approach in the context of this study is as follows: GMELT datasets and models of mass changes of the HMA (Figure 2) will define the high-resolution model in the HMA region while

the rest of the global model is the NASA GSFC GRACE mascon solution (all other previously described 116 components of the NASA GSFC forward model procedure are also included). Time series of mass changes 117 from GMELT will be combined into an ensemble product based on consensus across the HiMAT team. We 118 will explore the use of weighting factors to combine multiple overlapping estimates, or the testing of different 119 data/model combinations within our iterative framework. Next or best estimates will be expressed to at 120 121 least spherical harmonic degree and order 90, where the expansion size can be increased if it is determined to be beneficial. Each new Level 1B processing produces a new global mascon solution, where the HMA 122 region mascons describe the long spatial wavelength GMELT error, which serves as feedback for the next 123 iterative construction of the GMELT water balances. Updated GMELT model output is iteratively applied 124 to new Level 1B forward model runs until the GRACE mascon updates are sufficiently close to zero, 125 indicating that the high-resolution GMELT model is in full agreement with the GRACE measurements. It 126 may prove to be more effective and efficient to circumvent the formation and inversion of normal equations 127 by directly analyzing the range-rate or range-acceleration residuals generated from the Level 1B processing 128 as described in recent studies [Loomis et al., 2015; Eicker and Springer, 2016]. We note that our iterative 129 procedure enables adjustment of parameters on individual modules of GMELT, or on weighting factors 130 that combine the ensemble of mass change estimates. 131 NASA GSFC GRACE Level 1B processing applies a baseline orbit parameterization [Rowlands et al., 132 2002 that does not require the processing and reduction of GPS measurements, but instead uses the 133 GPS-determined Level 1B navigation files to define the initial orbit that is adjusted simultaneously with 134 the gravity from the inter-satellite measurements (also accounting for the Level 1B attitude quaternions 135 and accelerometer measurements). This allows for the rapid formation and inversion of normal equations 136 for the full duration of the GRACE mission. The ability to quickly process and invert new solutions is an 137 important practical consideration for the work proposed here, where a number of different GMELT outputs 138

## 140 DISCUSSION

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#### ACKNOWLEDGEMENTS

will need to be analyzed and iterated.

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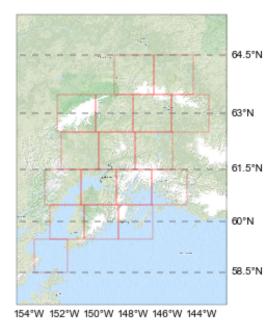


Fig. 1. Mascons of the western Gulf of Alaska.

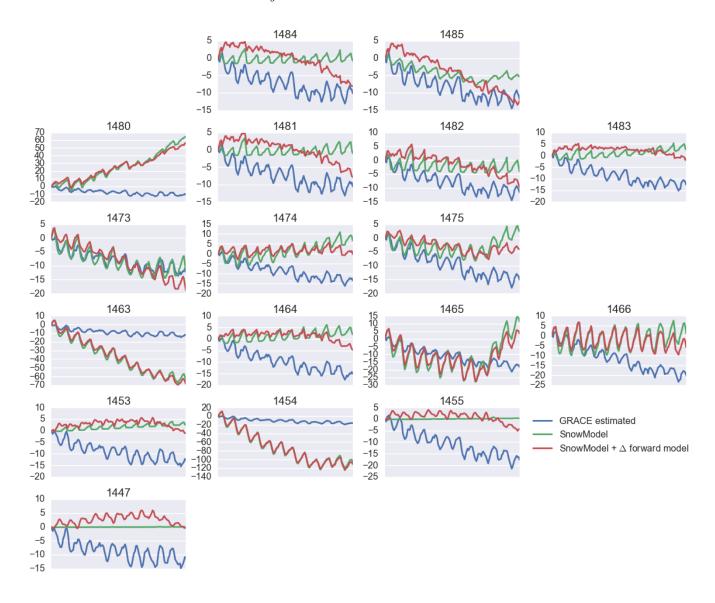


Fig. 2. Comparison of GRACE water balances prior to forward modeling (Luthcke and others, 2013), (blue); modeled water balances using SnowModel (Beamer and others, 2016) (green); and modeled water balance corrected by the forward modeling in this study (red). Subplots are arranged to approximately match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

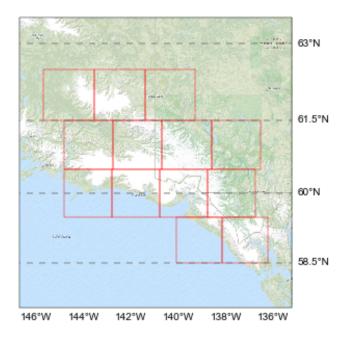


Fig. 3. Mascons of the western GOA

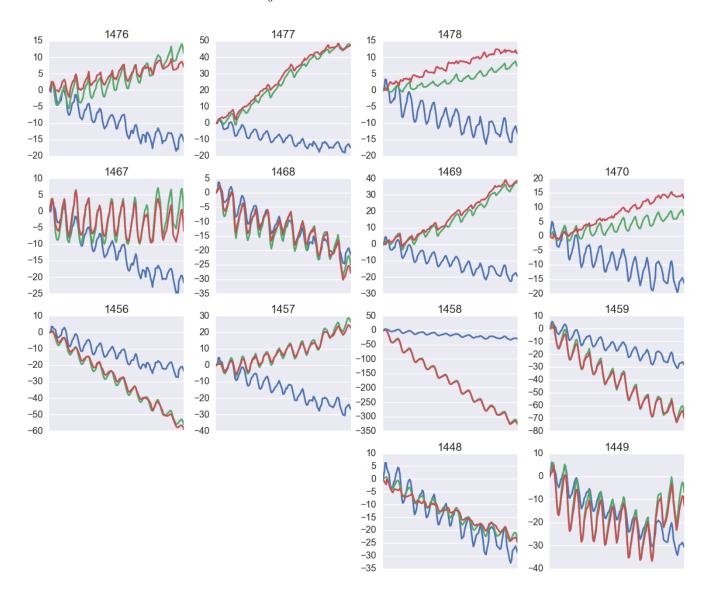


Fig. 4. Mascons of the western GOA

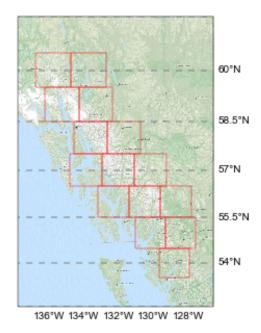


Fig. 5. Mascons of the western GOA

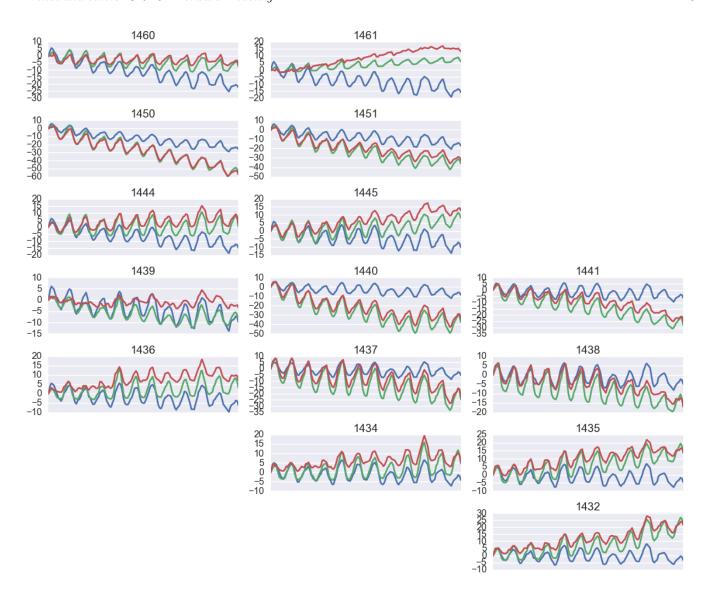


Fig. 6. Mascons of the western GOA