

Optimal filtering of mean dynamic topography models obtained using GRACE geoid models.

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ABSTRACT Least square

ABSTRACT Least squares collocation is an estimation technique where discretely located observations of different kinds can be integrated. The technique allows a rigrous description of the full covariance associated with signal, the errors as well as the sensitive of white the prival feed and the occan objective topography are analysed and described. The multi-disciplinary project "Good and Occan Circulation in the North Allantic (GOCINA) aims at enhancing the capacity in Earth observation using dafa from the European Space Agency missions ENVISA' and GOCE. Two examples of the where the standed GOCE gravity field models are used and one example where GOCE data are used detectly in least squares collocation to obtain an optimal product. In this study squares collocation to obtain an optimal product. In this study the techniques are applied to enhance the estimation of the Mean Dynamic Topography using the high resolution Mean Sea Surface KMS04 and the geoid model GGM02S from Sea Surface KMS04 and the geoid model GGM02S from GRACE. Especially, for modeling marine geodetic quantities with incomplete global coverage the methods have its advantages compared to a regular expansion into spherical harmonic functions.

1 Background

Lest squares collocation has been used widely for gravity field determination (e.g., Mortz, 1980). The method has mainly been used in regional computations ince the computational effort in inverting the equation system is quite big and depends on the number of observations. However, the technique has it strengths because different discrete data types may be integrated and ther this signal and error characteristics are taken rigorously into account. Tis therming (2004) has taken to be account in the account of the computed by the tools SST and SSO and demonstrated by salaw. In global analyses of the gravity field quantities are expanded into spherical harmonic functions. This is a well

expanded into spherical harmonic functions. This is a well known and efficient method to condense and synthesize the information from a large set of observations. The technique has its drawbacks because it is difficult to model high resolution fields where higher degree and order functions are required. The number of equations depends on the number of coefficients which depend on the harmonic degree and order Contracting which depend on the harmonic degree and order up to which the expansion is going. Hence, the computational effort in inverting the equation system becomes too big when a high resolution expansion is needed. In analyses of the ocean tides and ocean mean m becomes too big when a

In analyses of the ocean tides and ocean mean dynamic topography the use of global expansions into spherical harmonic functions have demonstrated is disadvaritages. The main problem occurs because the oceans only cover a subset of the sphere. Hereby, the nice properties of the spherical harmonic functions disagneer. They are no longer orthogonal and even for low degree expansions the equation systems become singular. Is to lost the least squares colocation and its regresentation of the mean dynamic ocean topography.

2 Least Squares Collocation and Representers

The least squares and minimum norm estimation techn called least squares collocation is used in this study. That is a method where an estimate of a quantity such as the geoid is obtained using the following expression

$$x = C_x^T (C + D)^{-1}$$

re C and D are cova iance matrices assoc ated with the signal and the errors of the observations y. x is the estimated quantity. The errors and error covariances were are using

 $\hat{c}_{xx} = c_{xx} - C_x^T (\mathbf{C} + \mathbf{D})^{-1} \mathbf{C}_x$ where cxx/ is the a-priori (signal) covariance between x and x/ (see e.g. Moritz, 1980).

By splitting up Eq.(1) and defining the vector **b** as the

 $b = (C + D)^{1} y$ we obtain $x=C_x^T b=\sum_{i=1}^N b_i K(P,Q_i)$

where the estimated quantity x in a point P now has been expressed by a sum of a series of the coefficients, or representers bi, multiplied by the reproducing kernel KP, QQ associating the estimate with the observations. The reproducing kernel is an expression for the covariance function.

Hunction. Hence, the estimates are expressed as a linear combination of a set of base functions which in this case are associated with the observations and not with some set of functions such as spherical harmonic functions. However, the spherical harmonic functions are fully included in the modeling of the reproducing kernel, as it is described in the next section.

3 Geoid covariance modeling

Using spherical harmonic functions signal and error covariances associated with the gravity field between points P and Q may be expressed as a sum of Legendre's polynomials multiplied by degree variances. That is

$$K(P, Q) = \sum_{i=2}^{\infty} \sigma_i^{TT} P_i(\cos \psi)$$

where are degree variances associated with the anomalous where are degree vanances associated with the anomalous gravity potential field and *y* is the spherical distance between P and Q. Expressions associated with geoid heights and gravity anomalies are obtained by applying the respective functionals on KP(P,Q), e.g. CMP-LN(LN(K(P,Q))) (more on collocation by Sansò, 1986, Tscherning, 1986).



(6)

(7)









The determination of the degree variances is essential to The determination of the degree variances is essential to obtain reliable and useful signal and error covariance bucklons. For the gravity field it has been accepted that the degree variances tend to zero somewhat faster than 3-2 and that the Tacheming-Rapp model (Tacheming & Rapp, 1974) may be used as a reliable model. When a spherical harmonic expansion of the gravity field up degree and order *N* has been used as a reference model and, hereby, been subtracted from the quantifies, then the associated error degree variances should enter the expression, eq. (5), up to harmonic degree N. That is

$$t_{i}^{TT} = \begin{cases} \mathcal{E}_{i} & i = 2,...,N \\ \frac{A}{(i-1)(i-2)(i+4)} \left(\frac{R_{i}^{2}}{R}\right)^{i+1} & i = N+1,.... \end{cases}$$

where A = 1544850 m4/s4, RB = R - 6.823 km were found in an adjustment so that agreement with empirical covariance values calculated from marine gravity data was obtained. The error degree variances, ti, are associated with the errors of the reference model from GRACE. The degree variances are shown in Figure 1.

Mean Dynamic Topography Covariance Modeling

To get reliable results of simulations and tests carried out using least squares methods it is important that both the signal and the error characteristics have been taken into account. In least error characteristics have been taken into account. In least squares collocation that means that the covariance function models should agree with empirically determined characteristics such as the variance and correlation tength. In analysis of errors formally estimated using eq.(7), it is very important that those quantities are reliable. That is also there case when MOT errors are analysed. Hence, a model describing the magnitude and the spectral characteristics of the MOT an needed. A kerned function associated with the MOT, may be expressed in a similar manner as the gravity fields as

 $C = \sum_{i=1}^{\infty} \sigma_i \mathcal{G} \mathbf{p}_i(\cos \psi)$

sion are assoc

where the degree variance in this expression are associated with the MDT, naturally. The degree variance model was constructed using 3rd degree Butterworth filters combined with an exponential factor (e.g. Knudsen, 1991). Hence, the spectrum of the MDT is assumed to have similar properties as the geoid spectrum; same type of smoothness and infinite. That is

$$\sum_{i}^{ss} = b \cdot \left(\frac{k_2^3}{k_2^3 + i^3} \cdot \frac{k_1^3}{k_1^3 + i^3} \right) \cdot s^{i+1}$$
(8)

where b, k1, k2, and s are determined so that the variance and the correlation length agree with empirically derived characteristics. This resulted in the model where b = 6.3 10-4 m2, k1 = 1, k2 = 90, s = ((R-5000.0)2/R2)2. The variance and correlation length are (0.20 m)2 and 1.3° respectively. The variance and correlation length of the current components (0.16 m/s)2 and 0.22° respectively. ents are

4 Results

The least squares collocation and its representation of the The least squares collocation and its representation or the determination of the mean dynamic topography. This is done using a set of observations that has been derived using the mean ess surface KNSO4 and a geoid computed from the GMM02S coefficients up to harmonic degree and order 60. The differences forming as so-called residual mean sea surface, or an estimate of the dynamic coveron the definition creases (see Fourier) This severation coveron the definition creases (see Fourier). This severation to severation the severation of the dynamic coveron the definition creases (see Fourier) This severation to severation the definition of the dynamic severation the severation to covering the global oceans (see Figure 2). This averaging eliminate most of the higher degree geoid signal without damaging the dynamic topography below harmonic degree 90 (see Figure 1).

Then the b-vector (eq. 3) was computed using proper covariance functions associated with the residual oid and the mean dynamic topography both averaged in 2 by 2 degrees cells (see Figure). Based on the representers in the b-vector and the

Based on the representers in the b-vector and same covariance functions the observations were reproduced in an estimation of the sum of the residual reproduced in an estimation of the sum of the residual good and the topograph versequed in Lo y despres color sum of e.g. 4. The RMS values of the observations, the sum of the and 0.13 m respectively. The got at sub hown in Figure and 0.13 m respectively. The got at sub hown in Figure 0.0 test as the underlying well known features of the topography.

90), Note also the underlying well known features of the topography. The using a proper cross covariance function the full mean dynamic topography was settimated using eq. 4 (see Figure). Note how well the geoid residuals have been eliminated. The estimated topography has been optimally fittered and give full resolution in a compromise with smoothness which the signal to noise relations an a rigorous manner balancing the least squares and the least norm oftenar.

5 Perspectives

The impact of the GOCE satellite mission on the recovery of the gravity field has previously been analysed for two simulated cases by Knudsen and Tscheming (2005). In the first case the GOCE Level 2 product is used where the the transmission of transmission of the transmission of the transmission of transmission of the transmission of t gravity field is approximated by spherical harmonic coefficients up to degree and order 200. In the second case synthetic Level 1B GOCE data are used directly in a gravity field determination using least squares collocation. In case two the full spectrum geoid error was improved from 31 cm to 15 cm and the resolution was doubled. The results are important for the future users of GOCE that need the extra accuracy

For the estimation and representation of the Mean Dynamic Topography the results from this study demonstrate that the collocation approach with its representers perform very well. Since only about half of the sphere is covered the number of averaged observations corresponds to the number of harmonic coefficients that provide the same resolution. Hence, the computational efforts are similar. Furthermore, the full signal and error characteristics are taken into account. This result in an optimal filtering of the data where aliasing caused by be truncation is avoided.

The result in an opening intering or the data where anison caused by be function is a voided. Finally, the major advantage is that full spectrum error covariances associated with the estimated topography are rigorously obtained. This is of crucial importance for the assimilation of altimetry into ocean circulation models.

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