A Drag free control based on Model Predictive Technics

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Abstract—This paper presents a solution for the drag free control problem of the European satellite GOCE based on Model Predictive technics. The approach followed during the orbit and attitude modelling yields to the decoupling of the plant’s six degrees of freedom into four linearized systems that include satellite’s dynamics, sensors and its corresponding noise profiles, actuators and its relative noise profiles and also, the characterization of the main exogenous disturbances (drag forces and torques). In the problem formulation are considered performance constrains in terms of the propulsion system activity, residual acceleration levels (linear and angular) and the spacecraft’s attitude angles. Also, it is presented the process to traduce the orbit and attitude performance requirements into specific mathematical constrains for the calculation of the optimal control law by means of predictive techniques. Simulation results using atmospheric drag profiles validated for this mission by Alenia Spazio (main mission contractor), are presented and analyzed in order to confirm the fulfillment of constrains and the performance requirements achieving.

I. INTRODUCTION

The Gravity Field and Ocean Circulation Explorer (GOCE) is a mission of the European Space Agency (ESA) oriented to the extremely high accurate measuring of the Earth’s gravity field (less than 10µm/s²) and the geoid modelling (accuracy of 1 – 2cm) at a spatial resolution better than 100km [1]. The mission will be developed with a single rigid octagonal spacecraft of approximately 5m long and 1m in diameter with fixed solar wings and no moving parts. In order to reach high accuracy in the determination of the lower harmonics of the gravity field, Precise Orbit Determination (POD) is implemented through a 12-channel GPS receiver with geodetic quality. Complementarily, the measurement of medium and high harmonics of the gravity field (0.005 Hz ≤ f ≤ 0.1 Hz) is developed through Satellite Gravity Gradiometry (SGG). This technic uses an ensemble of three pairs of three-axial electrostatic accelerometers (gravity gradiometers), each one containing a 320g proof mass electrostatically suspended and mechanically isolate from the spacecraft’s main body through a specially engineered cage. The specific role of the Drag Free Control (DFC) for the GOCE can be defined as an advanced drag compensation system that keeps the six proof masses in near ‘free fall motion’ and the average orbital altitude at about 250km. To develop these tasks, the forces that maintain each proof mass at the center of the cage are measured and the deviation from the nominal position within the allowed clearance band is used by the control system for commanding the propulsion system (two ion propulsion thruster and eight cold gas thrusters), that compensates the non-gravitational disturbance forces. The GOCE is thus forced to chase the proof masses, actively shielding it from the non-gravitational forces, the largest of which is the atmospheric drag. In this way, the complementary use of POD and SGG provide data sets of gravity gradient components with enhanced quality, respect to traditional techniques that have already reached their intrinsic limits [2].

Although, it is not a new problem, new realizations have been made in the last years integrating modern control techniques such H∞/H2 [3], [4] and Embedded Model Control [2], [5] as valid means for improving stability and performance robustness of the DFC control. Nevertheless, the H∞ formulation [3], [4] may not always achieve optimal performance, from the propulsion system point of view, when the worst-case plant (base assumption of these techniques) rarely or never occurs. On the other hand, in [5], is well described the difficult tradeoff between minimization of thrust level and achieving performance requirements at low frequency. In fact, not all the specifications for the GOCE Drag Free Control are satisfied in [5] at low frequencies as a payload for a low command activity.

In this frame, the aim of this paper is to solve the DFC problem for the European mission Gravity Field and Steady-State Ocean Circulation Explorer (GOCE) [1], showing that a Model Predictive approach allows the inclusion in the design of most restrictive constraints that conduct to a reduction of the propulsion system activity respect to other modern alternatives [3], [4]. Also, it is presented the process to traduce the orbit and attitude performance requirements into specific mathematical constrains for the calculation of the optimal control law by means of predictive techniques. Simulation results using atmospheric drag profiles validated for this mission by Alenia Spazio (main mission contractor), are presented and analyzed in order to confirm the fulfillment of constraints and the performance requirements achieving.
II. MATHEMATICAL MODEL

A. Reference frames

For establishing the mathematical model of GOCE’s motion, it is necessary to define a set of coordinate systems that allow a correct and easy integration of all the phenomena involved:

1) Inertial reference frame (IRF): For practical purposes, a geocentric coordinate system is a suitable IRF due to the almost circular and unaccelerated motion of the Earth around the sun (orbit period > 365 days) [6]. Based on this consideration, it is defined an IRF with origin at the Earth’s Center of Mass (CoM); the \( Z_{2000} \) axis, is the axis of Earth’s rotation that intersects the celestial sphere at the north pole; the \( X_{2000} \) axis completes the right-handed orthogonal coordinate system.

2) Local Orbit Reference Frame (LORF): It is a non-inertial reference frame used for describing the drag forces that perturbate the LEO. The origin is placed in the CoM of the spacecraft; the \( X \) axis is parallel to the instantaneous orbital velocity vector \( \mathbf{v} \); the \( Y \) axis is parallel to the instantaneous direction of the orbit angular momentum \( \mathbf{h} = \mathbf{r} \times \mathbf{v} \); the \( Z \) axis completes the right-handed orthogonal coordinate system.

3) Spacecraft Reference Frame (SCRF): It is a non-inertial reference frame used for describing the spacecraft attitude dynamics. The origin is placed in the CoM of the spacecraft; its \( \mathbf{i}_s \) points toward the motion direction; \( \mathbf{j}_s \) is orthogonal to the satellite earth face (positive direction is towards nadir); the \( \mathbf{k}_s \) axis completes the right-handed orthogonal coordinate system.

4) Natural Orbit Reference Frame (NORF): It is a non-inertial reference frame, suitable for the deduction of a space state model that describes the physical phenomena in terms of orbit plane (nominal case) and out-of-plane dimensions. The origin is placed in the Earth’s Center of Mass; the \( X \) axis points toward the perigee in the orbit plane; the \( Y \) axis is parallel to the instantaneous direction of the orbit angular momentum \( \mathbf{h} = \mathbf{r} \times \mathbf{v} \); the \( Z \) axis completes the right-handed orthogonal coordinate system.

The transformation from NORF to IRF is provided through the following rotation matrix:

\[
\begin{bmatrix}
X_L \\
Y_L \\
Z_L
\end{bmatrix} = \begin{bmatrix}
\mathbf{v} \\
\mathbf{v} \times \mathbf{r} \\
\mathbf{v} \times \mathbf{r} \times \mathbf{v}
\end{bmatrix} \begin{bmatrix}
X_N \\
Y_N \\
Z_N
\end{bmatrix}
\]

(2)

where, \( \mathbf{v} \) and \( \mathbf{r} \) are the instantaneous velocity and position vectors respectively. The transformation from the LORF to the SCRF, assuming an active attitude control, is provided through the following matrix:

\[
\begin{bmatrix}
X_s \\
Y_s \\
Z_s
\end{bmatrix} = \begin{bmatrix}
1 & \psi & -\theta \\
-\psi & 1 & \phi \\
\theta & -\phi & 1
\end{bmatrix} \begin{bmatrix}
X_L \\
Y_L \\
Z_L
\end{bmatrix}
\]

(3)

B. Orbit dynamics

Representing the position vector as \( \mathbf{r} = r \mathbf{e}_r \), then the linear velocity vector \( (\mathbf{v} = \dot{\mathbf{r}}) \) and the linear acceleration \( (\mathbf{a} = \ddot{\mathbf{r}}) \) in the NORF are:

\[
\mathbf{r} = r \mathbf{e}_r \\
\dot{\mathbf{r}} = \dot{r} \mathbf{e}_r + r \dot{\mathbf{e}}_r = \dot{r} \mathbf{e}_r + (r \dot{\theta} \cos \phi) \mathbf{e}_\theta + \dot{\theta} \dot{\phi} \mathbf{e}_\phi
\]

(4)

\[
\ddot{\mathbf{r}} = \ddot{r} \mathbf{e}_r + r \dddot{\mathbf{e}}_r + 2 \dot{r} \dot{\theta} \cos \phi \mathbf{e}_\theta + \dot{r} \dot{\phi} \sin \phi \mathbf{e}_\phi + r \dot{\theta} \dot{\phi} \mathbf{e}_\theta + (\dot{\theta} + \ddot{\phi} \sin \phi) \mathbf{e}_\phi
\]

(5)

In order to deduce the complete dynamic description of the system, the force model (4)-(6) is complemented with the vector of perturbation forces \( (\mathbf{d}_i) \), the residual terms (zonal and tesseral harmonics) of the gravitational model \( (q_i) \), and the vector of control variables \( (u_i) \) to be applied by the propulsion subsystem:

\[
\ddot{r} = -r \dot{\theta}^2 \cos^2 \phi - r \dot{\phi}^2 + \frac{\mu_o}{r^2} = \frac{d_r}{m} + \dot{u}_r + \frac{u_r}{m}
\]

(7)

\[
r \dot{\theta} \cos \phi + 2 \dot{r} \cos \phi - 2 \dot{\theta} \phi \sin \phi = \frac{d_\theta}{m} + \dot{u}_\theta + \frac{u_\theta}{m}
\]

(8)

\[
r \dot{\phi} + 2 \dot{r} \phi + \dot{r} \phi^2 \cos \phi = \frac{d_\phi}{m} + \dot{u}_\phi + \frac{u_\phi}{m}
\]

(9)

where, \( m \) represents the mass of the spacecraft. The orbit of the GOCE is characterized for: a near-polar trajectory (Figure 1(b)), a low eccentricity (< 0.005, Figure 1(c)), a high inclination angle (96.5°, Figure 1(d)), a mean altitude \( (r_o) \) of 250 km and a quasi constant angular velocity \( (\omega_o) \) of 1.17 mrad/s [1]. Considering these conditions as a nominal scenario and developing the corresponding linearization of the system of equations (7)-(9), the satellite’s orbit dynamics can be approximated through the following linearized
an important aspect to be assessed in a later treatment of the satellite’s propulsion system.

C. Attitude model

The Euler’s equation of motion, in the SCRF, for the GOCE can be expressed as:

\[ I_x \dot{\phi} - (I_y - I_z) \omega_y \omega_z = u_x + T_x \]  
\[ I_y \dot{\psi} - (I_z - I_x) \omega_z \omega_x = u_y + T_y \]  
\[ I_z \dot{\psi} - (I_x - I_y) \omega_x \omega_y = u_z + T_z \]

where, \( I_x, I_y, I_z \) are the principal moments of inertia, \( \omega_x, \omega_y, \omega_z \) are the SCRF components of the satellite’s absolute angular velocity, \( u_i \) are the control torques applied by the attitude controller and \( T_i \) are the exogenous torques due to non-gravitational forces (gravity gradient, Earth Magnetic Field, atmospheric drag). For small attitude deviation from the LORF orientation, the following linear relations can be applied:

\[ \omega_x = \dot{\phi} - \omega_z \psi \]  
\[ \omega_y = \dot{\theta} - \omega_x \psi \]  
\[ \omega_z = \dot{\psi} + \omega_x \phi \]

Using (14)-(16) into (11)-(13) and considering also the gravity gradient effect [6], the linear equations for rotational motion in the SCRF can be described as:

\[ I_x \ddot{\phi} - \omega_o (I_y - I_z) \psi + 4 \omega_o^2 (I_y - I_z) \phi = T_x + u_x \]  
\[ I_y \ddot{\theta} + 3 \omega_o^2 (I_x - I_z) \theta = T_y + u_y \]  
\[ I_z \ddot{\psi} + \omega_o (I_x - I_y + I_z) \phi + \omega_o^2 (I_y - I_z) \psi = T_z + u_z \]

An analysis of (17)-(19), shows that the pitch dynamics are decoupled from the roll/yaw dynamics, so the respective controllers can be synthesized separately.

D. Sensor modelling

The gradiometer bandwidth is about 1Hz, with a damping coefficient of 0.7 and a steady state gain asymptotically equal to 1. The measurement noise introduced in the control system by the gradiometer can be considered as a white noise with a spectral density of approximately 3.7810 – 11rad/s² [3].

E. Propulsion subsystem

The thrusters used to applied the forces required for orbit and attitude drag compensation are characterized for a second order behavior with a bandwidth of 20 Hz and damping coefficient of 0.7. The noise introduced into the control system by each thruster can be considered as a white noise with a spectral densities around 4.610 – 4rad/s² (u_x), 6.610 – 4rad/s² (u_y), 1.810 – 3rad/s² (u_z) [3]
III. MODEL PREDICTIVE CONTROL ALGORITHM

A. Control specifications

Once defined the DFC as a relative motion problem, the main performance requirements are established from the mission specifications. In particular, the techniques for gravity field determination on the GOCE mission (precise Orbit Determination, GPS/GLONASS orbit monitoring to cm-precision, and gravity gradiometry) determine that PSD of satellite incremental accelerations and attitude angles should be kept below the values specified in Table I (see [1], [3], [2]). Also, the propulsion technology considered (Ion thrusters and microthrusters), limits the amplitude of control forces to the values shown in Table II (see [2]).

B. Model Predictive Control (MPC) approach

As indicated in the model analysis of section II-B, four MIMO systems have to be controlled: two, related with the orbit modelling \((X_o-Y_o, Z_o)\) and the other two, related with the attitude characterization \((\phi - \psi, \theta)\). The MPC method applied is the same for all cases and consist of a fixed horizon algorithm with a quadratic cost function. The finite prediction horizon \((N_p)\) is a relevant concept into this problem because it allows to minimize the performance measure through an optimization problem with a finite number of decision variables and a finite number of constraints directly related with the orbit and performance requirements. From the control point of view, good disturbance rejection requires a high bandwidth but in the presence of sensor noise, higher controller bandwidth results in transmission of sensor noise into larger thrust usage. This stochastic balance of demands is satisfied optimally by a quadratic cost function. In this way, at each sampling time, starting at the current state, the optimal control problem is solved over a finite horizon. The optimal command signal is applied to the GOCE propulsion system only during the immediately following sampling interval. At the next time step a new optimal control problem based on new measurements of the state is solved over a shifted horizon. From a mathematical point of view, the problem is posed as to find the optimal control signal \((\hat{u}_t)\) or \((u_t)\) that minimizes the following finite horizon performance cost function:

\[
\min_U J(U, x(t|t), N_p, N_c) \tag{20}
\]

subject to: \(LU \leq K\) and \(TU \leq Z\)

where:

\[
J = \sum_{k=0}^{N_p-1} x^T_{t+k|t}Qx_{t+k|t} + \sum_{k=0}^{N_c-1} u^T_{t+k|t}Ru_{t+k|t}
\]

\(N_p\) is the prediction horizon

\(N_c\) is the control horizon \((N_c \leq N_p)\)

\[u_{t+k|t} = u_{t+N_c-1|k} \quad k = N_c, N_c + 1, \ldots, N_p - 1\]

\[U = \left[ u^T_{t|t}, u^T_{t+1|t}, \ldots, u^T_{t+N_c-1|t} \right]\]

\(Q = Q^T > 0\) and \(R = R^T > 0\) (the performance weights).

The variables \(x_{t+k|t}\) denote predicted states, given an input sequence \(u_{t+k|t}\), a state estimate \(x_{t|t}\) and considering a discrete space state model of the system.

C. Quadratic Programming Formulation of MPC

Due to the quadratic property of the performance objective \(J\) and in the presence of constraints on the control variables and the system states, the optimization problem in (20) can be transformed into a constrained quadratic programming problem [7] with consequent well known and efficient solution methods. To put the optimization problem in a form suitable for quadratic programming, there are introduced stacked vectors with future states and outputs:

\[
X = \left( x^T_{t|t}, x^T_{t+1|t}, \ldots, x^T_{t+N_p-1|t} \right)^T \tag{21}
\]
Using (23) the cost function becomes:
\[ U \]

Consequently the predicted states can be expressed as:
\[ X = \Omega x_{t|t} + \Gamma U \]

where
\[ C = \text{diag}(C, \cdots, C) \in \mathbb{R}^{N_p \times N_p - n} \quad \text{and} \quad \Omega, \Gamma \text{ are given by} \]
\[ \Omega = \begin{pmatrix} I \\ A_t^{2} \\ \vdots \\ A_{N_p - 1}^{N_p - 1} \end{pmatrix} \quad \Gamma = \begin{pmatrix} 0 & 0 & 0 & \cdots & 0 \\ B & 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ A_{N_p - 2}^{N_p - 2}B & \cdots & B & \cdots & 0 \end{pmatrix} \]

The cost function \( J \) can be articulated as:
\[ J(U, X) = X^T Q X + U^T R U \]

where,
\[ Q = \text{diag}(Q, \cdots, Q) \quad \mathcal{R} = \text{diag}(R, \cdots, R) \]

Using (23) the cost function becomes:
\[ J = \frac{1}{2} U^T H U + F^T U \]

where,
\[ H = 2 \left( \Gamma^T Q \Gamma + \mathcal{R} \right) \quad F = 2 x_{t|t}^T \Omega^T Q \Omega \]

**D. Constraint handling**

To ensure the stability of the control law obtained using MPC algorithm, it is impose the terminal constraint that the state vector \( x_{t|t+N_p|t} \):
\[ A_{N_p - 1}^{N_p - 1} + \sum_{i=0}^{N_p - 1} A_i^{N_p}Bu_{t+N_p - 1 - i|t} = 0 \]

which can be written as \( TU = Z \) where,
\[ Z = -A_{N_p}^{N_p - 1}x_{t|t} \]
\[ T = \left( A_{N_p - 1}^{N_p - 1}B, A_{N_p - 2}^{N_p - 2}B, \cdots, B \right) \]

The constraints imposed on the control variables when optimizing (20) will be the following:
\[ -u_{\text{max}} \leq u_{t+k|t} \leq u_{\text{max}} \quad k = 0, 1, \cdots, N_c - 1 \]

where \( U_{\text{max}} \) is a column vector containing the maximum values of control variables (Table II). The constraints imposed to the output variables are expressed as:
\[ -y_{\text{max}} \leq y_{t+k|t} \leq y_{\text{max}} \quad k = 0, 1, \cdots, N_p - 1 \]

where \( Y_{\text{max}} \) is a column vector containing the maximum values of output variables (Table II). Equations 27 and 28 can be written as linear constraints on \( U \) of the form \( LU \leq K \), where
\[ L = \begin{pmatrix} I_d \\ -I_d \end{pmatrix} \quad K = \begin{pmatrix} U_{\text{max}} \\ -U_{\text{max}} \end{pmatrix} \]

and \( I_d = \text{diag}(I, \cdots, I) \) is a \( N_c \cdot m \times N_c \cdot m \) matrix and \( I \) is the \( m \times m \) identity matrix (where \( m \) is the number of input variables). \( \Phi \) is a matrix with dimensions \( p \cdot m \cdot N_p \times N_c \cdot m \) where \( p \) is the number of system outputs) and is equal to:
\[ \Phi = \begin{pmatrix} C_B & 0 & 0 \[ C_{AB} & C_B & 0 \[ \vdots & \ddots & \ddots \[ C_{AN_p - 2} B & \cdots & C_{AN_p - 3} B & \cdots & C_{AB} \end{pmatrix} \]

The optimal solution can be numerically computed via the following optimization problem:
\[ U^{\text{opt}} = \arg \min_{LU \leq K} J(U) \]

This is a convex problem due to the quadratic cost and linear constraints [7].

**IV. Simulation results**

The atmospheric drag profiles used for testing the MPC 4 were provided by Alenia Spazio (main contractor of the GOCE) and correspond to a highly accurate simulations of the expected disturbances during the mission work time. The MPC algorithm is applied to the four respective discrete systems, using the mission sampling time \( T = 0.1s \) [5], the constraint values indicated in Tables I and II and the procedure described in section III-D. The \( Q \) and \( R \) matrices are selected considering the results of the controllability analysis (section II-B). The best values that satisfy a trade off between performance requirements (attenuation factor in the bandwidth) and propulsion system activity are:
\[ Q = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 10^8 \end{pmatrix} \quad R = 10^{-10} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \]

The prediction and control horizons are selected taking into account the nominal orbit requirements (period \( \approx 90\text{min} \)) and the sample time \( (0.1s) \). Specifically, with a \( N_p = N_c = 20 \), the number of parameters to optimize is...
consistent with feasible computation times for the GOCE mission. The quadratic constrained optimization problem was numerically solved using Matlab optimization ToolBox with final results as indicated below. The four controllers are designed with the above criteria and considering the dynamics and noise profiles for gradiometer and the spacecraft’s thrusters. The main results are showed in Figures 4(a) and 4(b), demonstrating that requirements are met with a comfortable margin of robustness. In fact, one of the most critical specification is the level of the PSD in the residual acceleration along the velocity vector ($Y_0$ in the ALORF), that the MPC maintains 90$dB$ under the established limit in the MBW. Comparatively, in Figure 5 are simultaneously presented the results obtained for the same acceleration variable ($Y_0$) with a $H_\infty$ [4] and a MP controllers. It is evident that the MPC reaches better performance with a control signal of the same level of that generated by the $H_\infty$ controller. The on-line computation difficulties related to the MPC law can be overcome by means of a "Fast" implementation of the MPC algorithm (FMPC) [8].

V. CONCLUSIONS

In this paper is proposed a Drag Free Control (DFC) based on predictive techniques for the GOCE mission of the ESA. The main contributions are related with the model set up as a linear decoupled plants and the use of predictive techniques for stabilizing the system and reaching attenuation values upper the performance requirements. Results demonstrate a better performance than traditional methods (state feedback) and modern frequency-weighted synthesis techniques.

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