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Towards New Controller Design Oriented Models of Propellant Sloshing in Observation Spacecraft

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The design of challenging and ambitious space missions entails a tightening of spacecraft pointing constraints. Among the many perturbations that have to be addressed, the sloshing of the on-board propellant is a complex issue. Recent developments in Computational Fluid Dynamics, supported by in-situ experiments like Fluidics or Sloshsat-FLEVO, open up ways for the characterization of liquids behavior inside tanks in microgravity. This knowledge can be applied in the context of spacecraft modeling and control. For this purpose, we present here a new approach to model the disruptive sloshing dynamics affecting spacecraft during attitude maneuvers. With the aim of mitigating slosh effects, this model will be used to design robust attitude controllers.

I. Introduction

FLUID dynamics defines sloshing as the movement of the free surface of a fluid inside tanks or containers. This complex nonlinear dynamics leads to disruptive effects which have to be addressed when a mass of fluid can move aboard vehicles. Sloshing then appears in tanker ships or trucks, but also in spacecraft. Indeed, orbital maneuvers (station-keeping, relocation and de-orbiting) require thrusters that consume liquid propellant. Since it is a strong constraint for the spacecraft lifespan, the mass of the on-board liquid propellant is a non-negligible part of the total spacecraft mass, making sloshing effects even more significant. Furthermore, sloshing is a low frequency phenomenon, it can compromise the system stability and it makes controller design more complex.

As a result of the coupled fluid-structure dynamics, the spacecraft undergoes disturbing efforts that alter both its attitude and orbit. In rare cases, propellant slosh dynamics can lead to serious consequences, from mission delay (NEAR, 1998 [1]) to mission loss (ATS-5, 1969 or Falcon 1, 2007 [2]). Several methods exist to deal with sloshing disruptive effects, passive and active ones. Hence, spacecraft tanks generally have baffles and membranes for slosh mitigation [3, 4]. In addition, time margins can be applied to let propellant come back at rest after aggressive maneuvers, or the desired angular velocity profiles can be smoothed to lessen the fluid excitation. Among active methods, frequency-based controllers are used to filter slosh dynamics [5].

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However these methods have some drawbacks :

- more complex tanks are also heavier
- time margins and smoothed profiles reduce mission availability
- frequency-based controllers can be rather hard to tune and sensitive to modeling errors [5]

The constraints on spacecraft pointing stability and pointing accuracy are becoming more stringent. Hence, it is necessary to develop a model of propellant sloshing that is representative enough to allow the design of a suitable attitude controller, which efficiently deal with sloshing and lighten the previously mentioned drawbacks.

Surface tension arises from the fluid molecular interactions, it creates and maintains the interface between two media. In microgravity, slosh dynamics becomes more complex as surface tension effects cannot be neglected in front of gravity [6]. Therefore many researches have been conducted in order to understand sloshing, particularly in space [7, 8]. It is difficult to reproduce microgravity conditions in laboratories (typically with zero-g flights or drop towers) and analytical descriptions of sloshing effects in such an environment are very complex. In-situ experiments dedicated to the study of sloshing were led, such as Sloshsat-FLEVO (ESA), Spheres (NASA) and Fluidics (ESA). The flight data have been used to adjust and validate CFD models. Due to their high computational cost, these CFD models are unsuitable to be used in a closed loop with the Attitude Control System (ACS), especially for the design of controllers. Thus, simplified Linear Time Invariant (LTI) models were developed [7, 8], but they do not take into account the dependency of the fluid response to the spacecraft maneuver.

In this paper we first state the problem, going through various kinds of models, some suitable for validation and others for design purposes. We then present our new modeling approach for propellant slosh dynamics in spacecraft, in which the fluid behavior is modeled by a nonlinear second order system with varying frequency and damping ratio. Such a model, exploiting similarities with any standard representation of LTI flexible modes [9], is expected to ease the design of attitude control systems with sloshing. The last part presents our identification procedure, applied on data computed with the CFD code DIVA (IMFT), and the subsequent results.

II. Problem statement

Attitude Control Systems have to ensure a pointing requirement after an attitude maneuver despite perturbations such as sloshing. To do so, controller-design and validation models are used during spacecraft ACS development.

The most representative and accurate models are those used in CFD. These models describe the coupled fluidstructure dynamics in microgravity with the Navier-Stokes equations extended by terms that depend on the propellant surface tension and the spacecraft inertial forces (linked to the angular velocity and acceleration). Recent developments led to the accurate computation of these equations. For instance, the CFD codes DIVA (Institut de Mécanique des Fluides de Toulouse) [10, 11] and COMFLO (University of Groningen) [12] are able to solve fluid-structure problems in microgravity. In particular, they can compute the forces and torques applied by the fluid on the tank. An interesting point is that both DIVA and COMFLO solvers were improved and validated, using respectively Fluidics and Sloshat-FLEVO experiments. However CFD solvers have a high computational cost, which makes difficult the closed-loop implementation with ACS (cf. figure 1, in which Γ_F is the fluid torque, Γ_C is the actuators torque and Γ_P is the torque from other perturbations, such as solar radiation pressure).



Fig. 1 System block-diagram with a closed-loop CFD model

A first solution is to use CFD in open loop as shown in figure 2. As there are no longer any interactions between the satellite dynamics and the solver, the implementation and the use of this model are easier. Yet it is only valid when the spacecraft angular velocity Ω is close to the desired one Ω_d . It requires the execution of many simulations to cover errors, still at high computational cost. Therefore, this model can be suitable for controllers validation but not for their design.



Fig. 2 System block-diagram with an open-loop CFD model

In order to design controllers, simplified mechanical-inspired models were developed. In these models, the fluid behavior is described by a mechanical system such as spring-mass, pendulum [13], free-mass or mass constrained on a surface [14, 15]. Equivalent Mechanical Models can be used to describe a complete satellite system including solar panels [16]. They are either based on a linearized fluid theory or on (semi-) empirical laws [17, 18]. These models, along with the passives methods mentioned in section I, have been successfully used for decades, for launchers and satellites [19]. Such models can be described with the standard representation of LTI flexible modes (cf. eq. 1 and 2).

$$\Gamma_F = L^I \ddot{\eta} \tag{1}$$

$$\ddot{\eta} + C_S \dot{\eta} + K_S \eta = -L \dot{\Omega} \tag{2}$$

In which $\eta \in \mathbb{R}^{n_S}$ are the n_S flexible modes, $L \in \mathbb{R}^{3 \times n_S}$ is the matrix of the modal participation of the flexible modes, $C_S = diag(2\xi_{S_i}\omega_{S_i})$ and $K_S = diag(\omega_{S_i}^2)$, where ω_{S_i} and ξ_{S_i} are respectively the natural frequency and the damping ratio of the *i*-th mode.

Equivalent Mechanical Models can be used in close-loop (cf. figure 3), providing a true interaction with the rigid spacecraft. Yet in these models the mechanical parameters (generally constant or possibly time-varying while considering propellant consumption [20]) do not evolve along with the spacecraft state.



Fig. 3 System block-diagram with a closed-loop LTI model

By using CFD solvers it is possible to characterize sloshing disruptive efforts depending on the spacecraft current maneuver. Such a study has been carried out with DIVA at IMFT [10], with a focus on the dependency of the liquid response to the centrifugal (eq. 3) and the impulse (eq. 4) Bond numbers.

$$B_O^{cent} = \frac{\rho \Omega^2 L D_0^2}{\sigma} \tag{3} \qquad B_O^{imp} = \frac{\rho \dot{\Omega} L D_0^2}{\sigma} \tag{4}$$

Where Ω is the spacecraft angular speed around the considered axis of rotation, *L* is the distance between the tank and the rotation axis (lever arm), ρ is the fluid density, σ is the fluid surface tension and D_0 is the pressurization gas bubble initial diameter (linked to the filling ratio).





Fig. 4 Example of DIVA results for a $4.72 \times 10^{-2} rad/s$ steady-state velocity

The figure 4 shows the time evolution of the sloshing torque along the rotation axis, for different values of the enforced angular acceleration. The torque discontinuities that appear at the start of each response are due to the square shape of the acceleration. As the system reaches a constant angular velocity, the torque goes to zero. The only remaining effort applied by the fluid on the tank comes from the centrifugal effect that pushes the fluid against the tank wall, which does not create a torque along the Z-axis. The data allow us to suppose that the torque could be approximated by a damped quasiperiodic function, with a frequency and a damping ratio that are functions of the spacecraft angular velocity and acceleration.

Usually, in Attitude Control Systems [21–23], the guidance loop provides an angular velocity profile in triangular shape or truncated triangular shape (cf. figure 5) for each satellite frame axis, similar to the profiles considered in the IMFT study. These profiles are computed to realize the attitude maneuver in near-minimal time given the spacecraft maximum angular velocity and acceleration constraints. The desired attitude profile is then obtained by integration of the speed guidance profile. By derivation, a bang-bang or bang-stop-bang torque input to be realized by the actuators is readily deduced. Consequently we will focus on these profiles.



Fig. 5 Angular velocity (dashed) and command torque (solid) profiles

III. Modeling of propellant Slosh Dynamics as a Nonlinear Second Order System

By studying the Navier-Stokes equations in microgravity, we can identify the parameters affecting sloshing :

- · tank filling ratio
- gravity vector orientation w.r.t. the spacecraft, linked to the attitude θ
- liquid properties, e.g. density, viscosity, surface tension
- tank geometry and position inside the spacecraft
- angular speed Ω and acceleration $\hat{\Omega}$

The system on which we apply our modeling method is the one that have been used in the IMFT study. We consider that thrusters are not used during attitude maneuvers, so the filling ratio is constant. This hypothesis is relevant since propellant is saved for orbit control, then reaction wheels and control moment gyros are preferred for both attitude control and attitude guidance maneuver. We also suppose that gravity effects can be neglected (microgravity) and that the propellant properties do not change. The tank position inside the spacecraft is fixed, and we consider a rigid tank. Hence the remaining variables are the angular speed and acceleration.

Let split the liquid-filled satellite in two subsystems: the dry satellite and the liquid propellant. By doing so, in order to study the satellite dynamics, we have to consider only the perturbations exerted by the liquid on the spacecraft, instead of the propellant behavior. From the preceding discussion, we propose to model the sloshing torque as the output of a second order system with a frequency and a damping ratio as functions of the satellite angular velocity and acceleration, thus we obtain the system shown in figure 6. Doing so, equations 1 and 2 become :

$$\Gamma_F = f(\eta, t) \text{ or } f(\ddot{\eta}, t)$$
(5)

$$\ddot{\boldsymbol{\eta}} + C_S(\boldsymbol{\Omega}, \dot{\boldsymbol{\Omega}}, t)\dot{\boldsymbol{\eta}} + K_S(\boldsymbol{\Omega}, \dot{\boldsymbol{\Omega}}, t)\boldsymbol{\eta} = -\boldsymbol{g}(\boldsymbol{\Omega}, \dot{\boldsymbol{\Omega}}, t)$$
(6)

Once the shape (e.g. linear or polynomial) of the functions f, g, C_S and K_S is chosen, the necessary parameters can be identified using CFD results. Note that we introduced a dependency to the time t, which has appeared to be necessary during the identification procedure development.

The fluid block can be easily added to an already existing satellite model as a feedback, making it *plug and play*. Even though the first mode energy contribution is the highest, the representation we use allows to add as many sloshing modes as necessary. Thus the model complexity, e.g. the state-vector size, can be adjusted at will.



Fig. 6 System block-diagram of the proposed closed-loop model

This approach can be considered as a generalization or an abstraction of Equivalent Mechanical Models, but it does not have to deal with the choice of equivalent mechanical models parameters (e.g. masses, stiffness coefficient, pendulum length and moment of inertia) which are generally taken constant and based on empirical or semi-analytical expression often only valid for axisymmetric tanks.

We will now describe our identification method and illustrate the precision of our identified model with respect to CFD solver results on a given example.

IV. Identification and Numerical Results

The CFD tool considered in this study to model propellant sloshing is the DIVA code (Dynamics of Interface for Vaporization and Atomization) developed in the Fluid Mechanics Institute of Toulouse. It is based on the level-set method of Osher and Sethian [24] to describe propellant and gas distributions within the computational domain. A transport equation is solved at each time step to capture the motion of the fluids throughout time. A second level-set function describes the static interface between the solid tank wall and the fluids in the tank referential. The incompressible Navier-Stokes equations for Newtonian two-phase flows are solved using the Ghost Fluid Conservative viscous Method with an Implicit scheme (GFCMI) of Lepilliez et al. [25]. This method uses the projection method of Chorin [26] adapted to two-phase flows by Sussman et al. [27]. The boundary conditions at the tank wall are considered within the computational domain and consist in a subcell Dirichlet boundary condition on the velocity and a subcell Neumann boundary condition on the pressure. The static contact angle is enforced at the tank wall by extending the level-set function within the solid domain [25]. The numerical solver has been used to compute propellant sloshing in spherical tanks in microgravity conditions for rotational manoeuvres at low Bond numbers [11] and exhibits a good agreement with data from the Fluidics experiment led in the International Space Station [28].

The data sets used in this study [11] correspond to a spherical tank with a diameter of 0.585 *m*, the lever arm is 0.4 *m*. The fluids that have been considered have properties close to the actual properties of fluids used in space applications. The propellant density is 1004 kg m^{-3} and its viscosity is $9.13 \times 10^{-4} kg m^{-1} s^{-1}$. The pressurization gas density is $2.41 kg m^{-3}$ and its viscosity is $1.99 \times 10^{-5} kg m^{-1} s^{-1}$. The surface tension is equal to $0.03325 N m^{-1}$.

The initial conditions are a tank at rest with the gas as a bubble located at the center of the tank. The static contact angle is set to zero. The maneuver is a rotation along the *Z*-axis following a bang-stop profile. We consider only the torque Γ_Z and the rotation $(\theta, \Omega, \dot{\Omega})$ along this axis. We have 42 data sets, each one corresponds to a bang-stop profile with an angular velocity in Ω_{max} (*rad/s*) and an acceleration in $\dot{\Omega}_{max}$ (*rad/s*²) (cf. eq. 7 and 8).

$$\Omega_{max} = \left[4.720 \times 10^{-3}, 6.680 \times 10^{-3}, 1.055 \times 10^{-2}, 1.493 \times 10^{-2}, 2.111 \times 10^{-2}, 3.338 \times 10^{-2}, 4.721 \times 10^{-2}\right] (7)$$

$$\dot{\Omega}_{max} = \left[4.500 \times 10^{-5}, 9.000 \times 10^{-5}, 2.230 \times 10^{-4}, 4.450 \times 10^{-4}, 8.920 \times 10^{-4}, 2.229 \times 10^{-3}\right]$$
(8)

We are in a black-box case, without any precise idea of the functions f, g, C_s and K_s (cf. eq. 5 and 6). Even if we describe the system with a nonlinear second order system, it is obvious that stronger nonlinearities occur. Yet, we can capture some of these nonlinearities by making the identification on several angular velocity and time intervals. We then get an evolution of the parameters. As we are not able to describe their behavior, we chose to average the parameters (and indirectly the nonlinearities) to find the best trade-off.



Fig. 7 System in equilibrium configuration (the colored part is the liquid

While examining the simulations results, we noted that the discontinuities in the sloshing torque, arising from the discontinuities in the acceleration profiles (cf. section II), were not well enough taken into account. Indeed it is a really strong nonlinear effect. To override this problem, we split the model on both sides of the discontinuity. By doing so we get two submodels, one *before the discontinuity (bang part)* and one *after the discontinuity (stop part)*. This kind of problem can be then addressed with switching control techniques.

In order to chose the functions f, g, C_s and K_s , we initially suppose that g and f are linear. As numerical integration is very sensitive to initial condition, we prefer to use numerical derivation. Thus the sloshing torque Γ_Z is directly the flexible state η . The resulting Linear Parameter Varying system to identify is then :

$$\Gamma_Z = \eta \tag{9}$$

$$\ddot{\eta} + C_S \dot{\eta} + K_S \eta = -A_S \Omega - B_S \dot{\Omega} \tag{10}$$

We consider ranges of interest for :

- the angular acceleration, $\dot{\Omega}_{rng} = [-\dot{\Omega}_N, \dots, -\dot{\Omega}_1, 0, \dot{\Omega}_1, \dots, \dot{\Omega}_N], N \in \mathbb{N}$
- the angular velocity, $\Omega_{rng} = [-\Omega_M, \dots, -\Omega_1, 0, \Omega_1, \dots, \Omega_M], M \in \mathbb{N}$
- the time, $T_{rng} = [0, T_1, \dots, T_P], P \in \mathbb{N}$

By using the data from each profiles, we then construct tensors *C*, *K*, *A* and *B*. For the model *before the discontinuity*, the (i, j)-th element of each tensor corresponds to the value of the associated function evaluated at $(\dot{\Omega}_i, \Omega_j)$. For the model *after the discontinuity*, the (i, j, p)-th element of each tensor corresponds to the value of the associated function evaluated at $(\|\dot{\Omega}_{0,j}\|, \Omega_j, T_p)$. $\dot{\Omega}_{0,j} \in \dot{\Omega}_{rng}$ is the acceleration that brought the tank to the velocity Ω_j . It is not possible to have $\Omega_j > 0$ and $\dot{\Omega}_{0,j} < 0$, and $\Omega_j < 0$ and $\dot{\Omega}_{0,j} > 0$, hence the use of the absolute value of the acceleration. The procedure is illustrated by figures 8 and 9.



(a) Before the discontinuity (time invariant)

(b) After the discontinuity (time varying) at time $T_p \in T_{rng}$

Fig. 8 Illustration of matrices construction and the used subscripts



Fig. 9 Splitting of bang-stop profiles

A. Identification in the case $\Omega > 0$ and $\dot{\Omega} > 0 \rightarrow$ Model before the discontinuity

This case allows us to fill the (L, R) blocks of each matrix (cf. figure 8) of the model *before the discontinuity*. It corresponds to the part of the profiles with a constant acceleration and a velocity ramp. We consider the profiles with the same angular acceleration $\dot{\Omega}_j \in \dot{\Omega}_{max}$. The number of velocity intervals N is chosen small enough to catch nonlinearities and to avoid over-averaging, and large enough to still have enough data to perform the identification. For this case, it is sufficient to use time invariant parameters.

On each velocity interval $[\Omega_i, \Omega_{i+1}]$, $i \in [1, ..., M]$, we solve a linear identification problem Y = PX. The measurement vector is given by $Y = \ddot{\Gamma}_Z(t)$, which is computed using central finite differences. The observation matrix is $X = [\dot{\Gamma}_Z(t), \Gamma_Z(t), \Omega(t), \dot{\Omega}(t)]$. The parameter vector is P = [C, K, A, B]. In order to ensure that the values of *C* and *K* make sense from a physics point of view, we bound them using maximal and minimal values of sloshing frequency and damping ratio. The sloshing frequency of each data set is computed using spectral analysis with a Fast Fourier Transform. We also add constraints to enforce both initial and final conditions on each interval. The resulting constrained optimization problem is then solved with a linear least square method by using the MATLAB *lsqlin* function. As it is a convex problem, we get a global result.

Each interval is common between different profiles with the same maximum angular acceleration. Hence the result for a given velocity interval is taken as the average of the values obtained for the same interval on different profiles. Finally we get the value of the parameters for the average velocity on the interval $\Omega_{i/2}$ and the acceleration $\dot{\Omega}_j$. To cover the whole ranges Ω_{rng} and $\dot{\Omega}_{rng}$, we interpolate using the MATLAB function *interp2* with a cubic-spline method. Note that even though the model *before the discontinuity* is time invariant, the parameters are not constant and change along the angular velocity ramp.

B. Identification in the case $\Omega > 0$ and $\dot{\Omega} = 0 \rightarrow$ Model after the discontinuity

This case allows us to fill the (L, R) blocks of each tensor (cf. figure 8) of the model *after the discontinuity*. It corresponds to the part of the profile that with a zero acceleration and a constant velocity. We consider profiles with the same steady-state velocity $\Omega_i \in \Omega_{max}$ and various accelerations $\dot{\Omega}_{0,j} \in ||\dot{\Omega}_{rng}||$. The methodology is almost the same as before. We define time intervals of a given length to catch nonlinearities without over-averaging, then we solve the constrained linear least square problem. We get the value of the parameters for the velocity Ω_i and the acceleration $\dot{\Omega}_{0,j}$ at any given time in T_{rng} . We then interpolate to cover the ranges Ω_{rng} and $\dot{\Omega}_{rng}$.

C. Identification in the case $\Omega = 0$ and $\dot{\Omega} > 0 \rightarrow$ Model before the discontinuity

This case allows us to fill the (L, M) blocks of each matrix (cf. figure 8) of the model *before the discontinuity*. Good results are obtained by extrapolating or duplicating values from (L, R) for the coefficients $C_{(L,M)}$, $K_{(L,M)}$ and $A_{(L,M)}$. $B_{(L,M)}$ is then computed by using the initial conditions, which are the same for each profile with the same angular acceleration. For this case it is also sufficient to use time invariant parameters.

D. Identification in other cases

Using the property that the system is symmetrical, we have (for instance for *C*):

- $C(\dot{\Omega}_i, \Omega_j, T_p) = C(-\dot{\Omega}_i, -\Omega_j, T_p)$
- $C(\dot{\Omega}_i, 0) = C(-\dot{\Omega}_i, 0)$
- $C(0, \Omega_i) = C(0, -\Omega_i)$

We can then fill the blocks (U, L), (U, M) and (L, L). If we suppose that the system is reversible, we flip the data along the time axis and use the identification procedure to fill the blocks (U, R) and (L, L). As we are able to fill the matrices blocks for which we do not have data, by using the system properties, our method requires less CFD simulations and thus reduce the duration of identification and validation steps.

In order to go from the matrices or tensors to the functions we can apply curve-fitting techniques, for instance a least square method applied on a polynomial function of the angular speed, acceleration and time. For validation purposes we can skip this step and use the SIMULINK *Look Up Table* block.

The whole identification code execution time is about two minute of time. It is fast enough to allow us to tune quickly the identification parameters (intervals lengths) to find the best fitting.

E. Results

For each data set, we compute the error ϵ between the torque given by our model Γ_Z^m and the data Γ_Z . In order to quantify the quality of our estimation, we get the relative error $||\epsilon||_2/||\Gamma_Z||_2$ using L^2 norm. The tables 1 and 2 show the results for both parts of the model, before and after the discontinuities.

	$\Omega_{max,1}$	$\Omega_{max,2}$	$\Omega_{max,3}$	$\Omega_{max,4}$	$\Omega_{max,5}$	$\Omega_{max,6}$	$\Omega_{max,7}$
$\dot{\Omega}_{max,1}$	0.5384	7.1728	5.0418	3.3715	3.1191	3.0924	3.9071
$\dot{\Omega}_{max,2}$	0.3856	0.3395	1.8239	1.6297	1.0071	0.8505	1.2408
$\dot{\Omega}_{max,3}$	0.0398	0.2123	0.0783	0.3534	0.1537	0.2346	0.4516
$\dot{\Omega}_{max,4}$	0.2333	0.5601	0.7921	0.7847	0.6367	0.5613	2.0376
$\dot{\Omega}_{max,5}$	0.1871	0.9303	1.1248	1.3472	1.5432	1.0736	1.2309
$\dot{\Omega}_{max,6}$	0.1138	0.5654	0.1559	0.3000	0.2963	0.4635	1.7179

Table 1Relative error (%) for each profile before the discontinuity

	$\Omega_{max,1}$	$\Omega_{max,2}$	$\Omega_{max,3}$	$\Omega_{max,4}$	$\Omega_{max,5}$	$\Omega_{max,6}$	$\Omega_{max,7}$
$\dot{\Omega}_{max,1}$	8.7077	3.9964	1.3782	0.8366	7.8829	52.3987	8.4349
$\dot{\Omega}_{max,2}$	11.8112	3.7038	2.0413	0.7709	2.5303	10.7775	5.7483
$\dot{\Omega}_{max,3}$	6.0635	2.8097	3.3712	2.1829	3.7826	3.9710	11.3312
$\dot{\Omega}_{max,4}$	11.7110	2.9681	7.4316	3.3607	4.4995	2.4865	1.9818
$\dot{\Omega}_{max,5}$	3.7777	4.2716	3.5500	8.1793	2.6146	2.8900	2.3253
$\dot{\Omega}_{max,6}$	5.5645	8.4743	4.0912	1.8475	4.0507	2.6784	4.3383

 Table 2
 Relative error (%) for each profile after the discontinuity

Results may change along with SIMULINK model configuration parameters, in particular the relative tolerance and the maximum time step. There is a trade-off between precision and computation speed.

For the part before the discontinuity the results are excellent, the mean relative error is 1.23%, with a maximum at 7.17% and a minimum at 0.04%. For the part after the discontinuity, the results are almost all under 10%. The mean relative error is 5.89%, with a maximum at 52.4% and a minimum at 0.77%. The second highest relative error is 11.8%. The worst case is presented in the figure 11a, we can see that it is still usable while considering robust control. Still, we are currently working on some improvements, mostly about tuning our algorithm parameters. Several other results are presented in figures 10 and 11.



Fig. 10 Example of the performances of the model "before the discontinuity"

This modeling approach for propellant sloshing in spacecraft is rather simple, yet it represents well the complex physics of the problem. As we introduce more physical knowledge into the model, the controller will be more adapted and will respond specifically to the sloshing dynamics. The main drawback of this method is the calculation time required to execute a CFD solver, but the computed data can also be used for validation purposes (cf. figure 2). However, as we split the system, the identification procedure has to be done one time for each kind of spacecraft tank. It can then be conducted in a spacecraft development life cycle. This method is particularly appropriate for satellite bus product lines, the identification and the controller design steps having to be done only once for several almost identical satellites. The model, along with the identification method, can be applied to various kind of tanks, even those for which we cannot use a symmetry hypothesis. In this case the model simply requires more CFD simulations. Multi-axis attitude maneuver cases are under study.

V. Conclusion and perspectives

In this paper, a new control design oriented modeling approach of sloshing dynamics in spacecraft has been presented. The fluid dynamics is efficiently approximated by a generalized second order system with varying frequency and damping ratio which are written as functions of the spacecraft angular speed and acceleration. Using results from the CFD solver DIVA of IMFT, we have been able to successfully identify our model parameters. By addressing the dependency of the fluid response on the spacecraft inertial forces, the proposed model includes more knowledge, which makes it more accurate. By doing so the highly nonlinear nature of sloshing can be reproduced by our model.



Fig. 11 Example of the performances of the model "after the discontinuity"

Yet, our approach is simple enough to design robust attitude controllers. The developed procedure also allows us to characterize identification uncertainties. These uncertainties, along with the ones on both the measurement and the modeling of the dry satellite, will be addressed using robust control techniques. The resulting controller will be able to deal with the errors and to ensure the spacecraft stability in spite of it.

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