Partitioned Strong Coupling Algorithms
for Fluid-Structure-Interaction

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Abstract
Numerical simulation of fluid-structure interaction is often attempted in the context of partitioned methods, where already existing solvers for fluid or structure alone are used jointly. Mostly this is done by exchanging information from time step to time step in an alternating fashion. These weak coupling methods are explicit and hence suffer from possible instabilities. Therefore often strong coupling — where equilibrium is satisfied jointly between fluid and structure in each time step — is desired; the simplest computational procedure is similar to the time stepping an alternating iteration. We show why also this approach may experience difficulties, and how they may be circumvented with block-Newton methods, still in the partitioned framework, by only using the solvers of the subproblems fluid and structure.

Keywords: fluid-structure-interaction, partitioned methods, strong coupling, block-Newton methods

1 Introduction
Fluid-Structure Interaction problems often show strong interplay between the fluid and the structure [1, 12], be it in the design of aircraft [2, 3], helicopters [4, 5], in the consideration of sloshing in tanks [6], or be it in biomechanical applications [7] — see also the examples in [14].

One possibility is to develop new software and solution methods for each of these coupled applications, as undoubtedly will happen in some areas. This is referred to as a monolithic approach [8], or sometimes as the direct method [16, 15, 17]. On the other hand, we shall assume here that the methods and software systems which have been developed for either fluid or structural applications will continue to be used. Therefore we consider partitioned methods [9, 10, 11] — also known as iterative methods [16, 15, 17] for fluid-structure interaction, i.e. separate solvers are used for the fluid and the structure [18, 24, 25].

For stability reasons, often a fully implicit formulation has to be used [15, 17, 6]. In this approach, we have to solve a large system of nonlinear equations with
the use of the (iterative) solvers for the subsystems. Commonly this is performed with block-Jacobi, block-Gauss-Seidel or related relaxation methods [19]. These simple methods do not always converge [20, 21, 22, 23, 24, 25]. We will introduce here a superior approximative block-Newton method [26, 22, 23, 24, 25].

2 Partitioned Formulation

We assume the fluid to be modeled adequately as an incompressible Newtonian fluid, satisfying the appropriate Navier-Stokes equation. As we want to couple this with a moving structure, we have to allow for a moving boundary. We take account of this by formulating the Navier-Stokes equation in an Arbitrary Lagrangean-Eulerian (ALE) [27] framework in the moving fluid domain \( \Omega_f \):

\[
\begin{align*}
\rho_f (\dot{v} + (v - \dot{x}) \cdot \nabla)v - \nabla p &= f, \\
2\sigma &= \nu(\nabla v + (\nabla v)^T), \\
\text{div} v &= 0.
\end{align*}
\]

The boundary \( \partial \Omega_f \) we assume to be divided into three disjoint parts \( \partial \Omega_f = \Gamma_v \cup \Gamma_q \cup \Gamma_c \) where on \( \Gamma_v \) the velocity is prescribed, on \( \Gamma_q \) the traction is given, and \( \Gamma_c \) is the coupling boundary, where the coupling conditions will be specified below. Here \( \rho_f \) is the fluid density, \( v \) the velocity, \( \sigma \) the viscous stress, the superimposed dot signifies the partial derivative w.r.t. time \( t \), \( x \) is the position of the reference ALE-coordinate system and \( \dot{x} \) its velocity. The fluid shear viscosity is denoted by \( \nu \), \( p \) is the pressure, \( g \) the body force in the fluid, and the differential operators are in the spatial frame.

The structure we assume to be modeled by a neo-Hookean material, hence the equilibrium equation takes the following form in a Lagrangean framework in a fixed domain \( \Omega_s \):

\[
\begin{align*}
\rho_s \ddot{u} - \text{DIV} (FS) &= f, \\
S &= \lambda(\text{tr} E)I + 2\mu E, \\
2E &= (C - I), \\
C &= F^TF
\end{align*}
\]

The boundary \( \partial \Omega_s \) we again assume divided into three disjoint parts \( \partial \Omega_s = \Gamma_u \cup \Gamma_t \cup \Gamma_c \), where on \( \Gamma_u \) the displacements \( u \) are prescribed, on \( \Gamma_t \) the tractions, and \( \Gamma_c \) is the coupling boundary with the fluid. The other quantities are the structure density \( \rho_s \), the displacement gradient \( F \), the second Piola-Kirchhoff stress \( S \), and the body load \( f \). The Lamé moduli are denoted by \( \lambda \) and \( \mu \), and \( E \) is the Lagrange-Green strain, derived from the Cauchy-Green tensor \( C \), and the differential operators are in the material frame.

On the coupling boundary \( \Gamma_c \) let a unique normal \( n \) be defined in the spatial frame. Then the coupling conditions may be expressed as requiring the velocities...
to coincide at the location $x(t) = x_0 + u(x_0, t)$

$$v(x(t), t) = \dot{u}(x_0, t),$$

and the tractions from fluid and structure to balance each other:

$$\langle \sigma - pI \rangle \cdot n = -\frac{1}{f} F S F^T \cdot n.$$

3 Discrete Formulation

We shall further assume that the fluid — described by Eq. (1) — and the structure — described by Eq. (3) — have been discretised; each one may use her favourite discretisation, e.g. finite elements, finite volumes etc., e.g. [13].

Up to now we have omitted to say how to move the reference coordinate system in the fluid domain. Several possibilities exist [28, 7, 29, 6], here we have followed [30] and modeled the connections between the nodes in the fluid domain as elastic springs. This fictitious inertia-free elastic body has displacement loading from the moving structure, on the outer boundaries it is fixed. The traction balance Eq. (6) introduces additional forces on the fluid, and the reaction force on the solid. Hence for the fluid the complete discrete equations of motion are [22]:

$$M_f \ddot{v} + N(v - \dot{x})v + K_f v + B_f p = g + T_f^T \tau$$  \hspace{1cm} (7)

$$B_f^T v = 0$$  \hspace{1cm} (8)

$$K_g x = Au.$$  \hspace{1cm} (9)

The terms in Eq. (7) are the discrete analogues of those in Eq. (1), the term $g$ includes the prescribed boundary stresses, and the additional term $T_f^T \tau$ comes from the interaction with the structure, where $\tau$ is the stress on the fluid-structure-interface. In Eq. (8) we see the discrete form of the incompressibility condition, and Eq. (9) describes the movement $x$ of the fluid domain, driven by the structure displacements $u$.

Similarly for the structure we obtain

$$M_s \ddot{u} + K_s(u)u = f - T_s^T \tau.$$  \hspace{1cm} (10)

To this we have to add the discrete coupling condition for the velocities,

$$T_f v = T_s \dot{u}.$$  \hspace{1cm} (11)

This combined set is a system of differential algebraic equations (DAEs) in the time variable $t$ of index 2, as can be easily verified. In order to allow an easier numerical treatment and to make them fit the more general formulation in section 4, they will be converted to index 1 by differentiation [22].
Differentiating Eq. (8) once gives $B^T_f \dot{v} = 0$. As $M_f$ is non-singular, we may solve for $\dot{v}$ from Eq. (7) and insert this into the above relation, giving

(12) \[ B^T_f M_f^{-1} (g + T^T_f \tau) - N_f (v - \dot{x}) v - K_f v - B_f p = 0. \]

Similarly, Eq. (9) for the grid movement has to be differentiated

(13) \[ K_g \dot{x} - A w = 0, \]

where we have introduced the new variable $w = \dot{u}$. With this definition the structural Eq. (10) can be reduced to first order:

(14) \[ M_s \ddot{w} + K_s (u) u = f - T^T_s \tau. \]

4 Abstract Coupling Algorithms

The development in the preceding section 3 enables us to write the coupled system in the form [21]:

(15) \[ \dot{y}_1 = f_1(y_1, y_2, z_1) \]
(16) \[ 0 = g_1(y_1, y_2, z_1, v), \quad \text{and} \]
(17) \[ \dot{y}_2 = f_2(y_1, y_2, z_2) \]
(18) \[ 0 = g_2(y_1, y_2, z_2, v), \quad \text{and the global coupling condition} \]
(19) \[ 0 = h(y_1, y_2, z_1, z_2). \]

The global DAE, as well as each local DAE, whether by itself or combined with the global coupling condition has to be an index 1 system. This means that the matrices

(20) \[ D_{z_i} g_i, \quad \begin{bmatrix} D_{z_i} g_i & D_{z_i} g_i \\ D_{z_i} h & 0 \end{bmatrix}, \quad i = 1, 2, \quad \begin{bmatrix} D_z g & D_z g \\ D_z h & 0 \end{bmatrix} \]

have to be regular, where $D_q$ is the partial derivative w.r.t. $q$.

For our concrete application, the abstract Eqs.(15–19) may be identified with the following terms [22]:

(21) \[ y_1 = \begin{bmatrix} v \\ x \end{bmatrix}, \quad z_1 = \begin{bmatrix} b \\ p \end{bmatrix}, \quad f_1 = \begin{bmatrix} b \\ \xi \end{bmatrix}, \quad v = \tau \]
(22) \[ g_1 = \begin{bmatrix} M_f b + N_f (v - \xi) v + K_f v + B_f p - g - T^T_f \tau \\ -B_f M_f^{-1} (-N_f (v - \xi) v - B_f p - K_f v + g + T^T_f \tau) \\ K_g \xi - A w \end{bmatrix} \]
for the fluid,
\begin{align}
y_2 &= \begin{bmatrix} u \\ w \end{bmatrix}, \quad z_2 = a, \quad f_2 = \begin{bmatrix} w \\ a \end{bmatrix}, \quad \nu = \tau \\
g_2 &= M_s a + K_s(u)u - f + T_s^T \tau
\end{align}
for the structure, and
\begin{equation}
h = T_f b - T_s a
\end{equation}
for the coupling condition.

This abstract setting allows us a simple formulation of different coupling algorithms. Let us remark that in the chosen formulation the differential equation only gives the definition of the various variables, whereas the equilibrium equations have gone into the algebraic conditions. What is still left is to verify the index 1 condition Eq. (20). This is a lengthy calculation [22], but the index 1 conditions are indeed satisfied as long as \( K_g, M_f, \) and \( M_s \) are regular — in our case they even can be assumed to be symmetric positive definite.

Let us assume that we have integrators for each system separately. We express this by saying that given the state at time \( t_n \), we have for each system separately a function \( \Phi_j \), \( j = 1, 2 \) which will advance the state — or an approximation thereof — to the end of the time increment at time \( t_{n+1} \). For the sake of simplicity we shall confine ourselves to the case where the time step in each subsystem is the same. Hence, for subsystem 1, we may say that given the values \( (y_1^{(n)}, z_1^{(n)}) \) and the functions \( (y_2, z_2, \nu) \) over the interval \( [t_n, t_{n+1}] \), we obtain
\begin{equation}
(y_1^{(n+1)}, z_1^{(n+1)}) = \Phi_1(y_1^{(n+1)}, z_1^{(n+1)}, y_1^{(n)}, z_1^{(n)}, y_2, z_2, \nu).
\end{equation}
We have written this equation in fixed point form for \( (y_1^{(n+1)}, z_1^{(n+1)}) \), assuming that this is also the iterative method to solve for the new values in case the single system method is implicit.

Similarly, for subsystem 2 we have an analogous function, we only have to exchange indices in Eq. (26):\begin{equation}
(y_2^{(n+1)}, z_2^{(n+1)}) = \Phi_2(y_2^{(n+1)}, z_2^{(n+1)}, y_2^{(n)}, z_2^{(n)}, y_1, z_1, \nu).
\end{equation}

From these single system integrators, different combinations may be built in order to integrate the global system.

5 Partitioned Solution Strategies

Probably the simplest and the best known are the staggered schemes, also referred to as weak or loose coupling. We first have to decide where we want to put the
global constraint Eq. (25). Usually this is put onto the “fluid-side”, i.e. we transfer the structural displacements, velocities, and accelerations onto the fluid domain, and not vice versa. On the other hand we then apply the fluid stress to the structure. We shall retain the denotation $\Phi_1$ also for this extended solution function, which now solves

$$(28) \quad (y_1^{(n+1)}, z_1^{(n+1)}, \nu^{(n+1)}) = \Phi_1(y_1^{(n+1)}, z_1^{(n+1)}, \nu^{(n+1)}, y_1^{(n)}, z_1^{(n)}, \nu^{(n)}, y_2, z_2).$$

The simplest staggered scheme then is for subsystem 1 to assume the time evolution of the state variables $(y_2, z_2)$ to be constant over the interval $[t_n, t_{n+1}]$, i.e. identically equal to $(y_2^{(n)}, z_2^{(n)})$, and then solve Eq. (28). Similarly for subsystem 2 we assume $(y_1, z_1, \nu)$ over the time increment to be constant and identically equal to the initial values $(y_1^{(n)}, z_1^{(n)}, \nu^{(n)})$, and with this we can solve Eq. (27). To sum up, information is only exchanged between the subsystems after the increment has been performed; otherwise inside the increment, each subsystem is solved in parallel independently of the other. It is quite easy to see that this scheme will only give first order accuracy in time, no matter how accurate the single system integrators are. This could be improved by a higher order extrapolation, but this will have a negative impact on the stability characteristics.

There are many variations on this staggering theme [20, 31], but they all share the same characteristic in that they are explicit when looked at as time-integrators although the integrators for the individual systems my be implicit. They also suffer from the same limitations as other explicit schemes in that they are not unconditionally stable, although some remedies exist for the linear stability. Often the time step limitations imposed by this explicit part are so severe that one wants to use an implicit method on the global level. Additionally, if global constraints have to be satisfied exactly, explicit methods are of no great use; global invariants such as energy conservation will almost certainly be violated by explicit schemes. All this makes the use of implicit methods desirable [20, 23, 24]. They offer the possibility to achieve the same results as monolithic methods [8], but with partitioned software. Satisfying an implicit condition globally is often referred to as strong or tight coupling. Additionally, this strong coupling allows better stability through energy conservation [10, 6, 32]. Another motivation for the use of implicit methods on the global level is that we are solving a DAE, and if we want to ensure that the constraints are satisfied, than it is advisable to use implicit methods.

As implicit methods will usually be iterative, we first need a starting value for the iteration through a process of extrapolating the solution into the new time step. Assuming that we know the values $(y_1^{(n+1)}, z_1^{(n+1)}, y_2^{(n+1)}, z_2^{(n+1)})$ and all the proceeding ones, we may define some kind of interpolation for the time interval $[t_n, t_{n+1}]$. The simplest possibilities are the constant values
(y_{1}^{(n+1)}, z_{1}^{(n+1)}, y_{2}^{(n+1)}, z_{2}^{(n+1)}), or the linear interpolation between these and the values at $t_n$. With this kind of extrapolation, we may again enter a process similar to Eqs.(27, 28). But as the values at $t_{n+1}$ were merely assumed and not known, this must now be regarded as an iterative process; the simplest again being certainly that one where the simple staggering idea is now performed iteratively inside one time increment.

That means that given some — assumed and approximate — values at time $t_{n+1}$, we perform one of the afore mentioned interpolation procedures, and then the process on Eqs.(27, 28) is carried out, i.e. each subsystem is solved independently, in parallel — of course, carrying this out in parallel is merely an option and not a requirement. This gives new, approximate values for the variables at $t_{n+1}$, and the whole process may now be performed in an iterative fashion until convergence.

This simple way of solving the global coupled system is easily recognised as some kind of nonlinear block-Jacobi method. We remember that there is no guarantee that this method will converge, even for starting values arbitrary close to the solution. Some kind of damping or under-relaxation or some other more sophisticated forms of pre-conditioning may be introduced in order to achieve convergence. This immediately brings to mind that the corresponding Gauss-Seidel process often converges faster [19]. It is easily constructed with the same ingredients. The only difference is that now we pick one subsystem — say subsystem 1 — to solve for first, then with these newly computed values we solve for subsystem 2, and this is repeated iteratively until convergence. As with the block-Jacobi process, we know that this process may also not converge, even for starting values arbitrarily close to the exact solution, and that the convergence behaviour depends on the order in which the subsystems are solved for. With only two coupled systems the ordering seems not so important and its influence could easily be tested, but for more than two coupled systems this is not feasible. It has been shown that it is possible to under-relax or dampen this process such that it is convergent for simple model problems [6, 32], and in [6, 11] it is discussed how to choose the relaxation parameter.

In this situation it is useful to note the following result [21, 33]:

Let

$$\alpha = \max_{t \in [0,T]} \| (D_{z_2} g_2)^{-1} D_{v_1} g_2 \left( D_{z_1} h \left( D_{z_2} g_1 \right)^{-1} D_{v_1} g_1 \right)^{-1} D_{z_2} h \|,$$

and let $L$ be the Lipschitz-constant of the extrapolation. Assume that $\alpha < 1$, and that at least $k$ iterations of the block-Gauss-Seidel scheme are performed, such that $L \alpha^k < 1$, and that $\Delta t = t_{n+1} - t_n$ is small enough. Then the global block-Gauss-Seidel method converges, and the error

$$\delta^{(n+1)} = \| y^{(n+1)} - y(t_{n+1}) \| + \| z^{(n+1)} - z(t_{n+1}) \| + \| v^{(n+1)} - v(t_{n+1}) \|$$
is bounded by

\[
\delta^{(n+1)} \leq C(\mu^{\max(0,k-2)}\delta^{(n)} y + \mu^{k-1}\delta^{(n)} z) + \varepsilon_1^{(n+1)} + \varepsilon_2^{(n+1)}.
\]

Here \(\varepsilon_1^{(n+1)}\) and \(\varepsilon_2^{(n+1)}\) are the errors incurred by the single system integrators Eqs. (27, 28) and \(\delta^{(n)} y\) and \(\delta^{(n)} z\) are the extrapolation errors, and \(\mu = \alpha + O(\Delta t)\).

As long as \(\Delta t\) is small enough such that \(\mu < 1\), the iteration will converge. We see that the contractivity constant \(\alpha\) is crucial, and without \(\alpha < 1\) the block-Gauss-Seidel method will not converge, no matter what the time step is; and the convergence depends strongly on the ordering of the subsystems in the block-Gauss-Seidel solution strategy.

Observe that if we perform enough iterations, essentially only the error components from the single system integrators remain. If we have \(\varepsilon_1^{(n+1)} = O((\Delta t)^p)\) and \(\varepsilon_2^{(n+1)} = O((\Delta t)^q)\) as convergence orders for the single system integrators, we obtain \(\delta^{(n+1)} = O((\Delta t)^{\min(p,q)})\), in contrast to the staggering scheme where we only have \(O(\Delta t)\).

In our view all this calls for strong or tight coupling methods which will converge unconditionally provided the time step is small enough; this then would roughly be a situation similar for the single system implicit integrator. With block-Gauss-Seidel methods there is no easy way of achieving this, although there are possibilities of preconditioning [21].

6 Block-Newton Methods

To have a more compact notation, we shall only consider one time step and introduce the fluid variables \(\xi = (y_1^{(n+1)}, z_1^{(n+1)}, u^{(n+1)})\) and the structure variables \(\zeta = (y_2^{(n+1)}, z_2^{(n+1)})\). Then the above Eqs.(27, 28) may be concisely written as

\[
(30) \quad \xi = \Phi_1(\xi, \zeta)
\]
\[
(31) \quad \zeta = \Phi_2(\xi, \zeta).
\]

It was observed in [26] that it is numerically advantageous to consider the system Eqs.(30, 31), i.e. the iteration equations, rather than the equilibrium conditions; and to consider the use of Newton’s method for the solution [26, 34, 35, 24, 25].

One step of Newton’s method for this combined system Eqs.(30, 31) entails the solution of the following linear system at each iteration, with \(\Delta \xi_k := \xi_{k+1} - \xi_k\) and \(\Delta \zeta_k := \zeta_{k+1} - \zeta_k\) and the iteration counter \(k\):

\[
(32) \quad \begin{bmatrix} I - D_{\xi} \Phi_1 & D_{\zeta} \Phi_1 \\ D_{\zeta} \Phi_2 & I - D_{\zeta} \Phi_2 \end{bmatrix} \begin{bmatrix} \Delta \xi_k \\ \Delta \zeta_k \end{bmatrix} = - \begin{bmatrix} \xi_k - \Phi_1(\xi_k, \zeta_k) \\ \zeta_k - \Phi_2(\xi_k, \zeta_k) \end{bmatrix}.
\]
We only want to use the existing solvers, i.e. the iteration mappings $\Phi_1$ and $\Phi_2$. In particular, we do not have direct access to the partial derivatives in Eq. (32). But if we solve the system Eq. (32) by an iterative method [36], all we need is a way to compute the product of the Jacobian matrix in Eq. (32) by an arbitrary vector.

We now want to consider a single iteration, and we will drop the iteration indices for ease of writing, and only look at computing the vector $[\Delta \xi, \Delta \zeta]^T$ from the right hand side. To start, we use — symbolically — block-Gauss elimination on the system Eq. (32):

\begin{equation}
\Delta \xi = -(I - D_{\xi} \Phi_1)^{-1}(\xi - \Phi_1(\xi, \zeta)) - C \Delta \zeta,
\end{equation}

with

\begin{equation}
C = (I - D_{\xi} \Phi_1)^{-1}[D_{\xi} \Phi_1].
\end{equation}

By inserting this into the second equation, we obtain

\begin{equation}
S \Delta \zeta = -r
\end{equation}
with the Schur complement

\begin{equation}
S = I - [D\zeta\Phi_2] - [D\zeta\Phi_2]C,
\end{equation}

and

\begin{equation}
r = (\zeta - \Phi_2(\xi, \zeta)) + [D\zeta\Phi_2]q,
\end{equation}

with

\begin{equation}
q = -(I - D\zeta\Phi_1)^{-1}(\xi - \Phi_1(\xi, \zeta)).
\end{equation}

In this way we can solve the second equation for \(\Delta\zeta\), and with this then solve

\begin{figure}
\centering
\includegraphics[width=\textwidth]{fig2}
\caption{Displacement Response for Strong Coupling}
\end{figure}

the first equation for \(\Delta\xi\). One step of this Newton-Raphson iteration may now be formulated as follows:

1. Solve \((I - D\zeta\Phi_1)q = \Phi_1(\xi, \zeta) - \xi\) for \(q\).
2. With \( r = \zeta - \Phi_2(\xi, \zeta) + [D\xi \Phi_2]q, \)

3. solve \( S \Delta \zeta = -r \) for \( \Delta \zeta. \)

4. Compute \( \Delta \xi = q - C \Delta \zeta. \)

It remains to specify how the Jacobians are computed, and how the terms with matrices are handled. To solve the first equation, we use the iterative solver for the fluid, \( \Phi_1. \) One iteration there can also be seen as one Newton-Raphson step for the solution \( q \) of the equation \( \xi - \Phi_1(\xi + q, \zeta) = 0 \) when \( \xi \) and \( \zeta \) are fixed. So with the iterative solver \( \Phi_1 \) we obtain \( q \approx z_m - \xi, \) where

\[
z_{j+1} = \Phi_1(z_j, \zeta), \quad j = 0, 1, \ldots, m - 1, \quad m > 1 \text{ with } z_0 = \xi.
\]

To apply \( [D\xi \Phi_2] \) in the second step we use finite differences:

\[
r = \zeta - \Phi_2(\xi, \zeta) + [D\xi \Phi_2]q \approx \zeta - \Phi_2(\xi + q, \zeta).
\]

In the third step an iterative Krylov method — here BiCGStab — is used to solve the system with the Schur complement matrix \( S; \) and we only need the action of \( S \) on some other vector \( w, \) again approximated via finite differences with some (small) step-size \( h: \)

\[
Sw = \frac{1}{h}S(hw) = \frac{1}{h}((I - D\xi \Phi_2)(hw) - [D\xi \Phi_2]C(hw)) \\
\approx \frac{1}{h}(hw + \Phi_2(\xi - C(hw), \zeta + hw) - \Phi_2(\xi, \zeta))
\]

Here we need \( C(hw) =: v, \) the solution of

\[
(I - D\xi \Phi_1)v = [D\xi \Phi_1](hw),
\]

computed as in the first step, where additionally finite differencing is used for \( [D\xi \Phi_1](hw): \)

\[
[D\xi \Phi_1](hw) \approx \Phi_1(\xi, \zeta + hw) - \Phi_1(\xi, \zeta).
\]

In the fourth step we know \( q, \) and again we need the action of \( C \) on some vector, this time \( \Delta \zeta, \) known from the third step. It is again done as in the third step for \( C(hw). \)

Convergence of these approximative Newton methods has been investigated in [37, 35, 36, 34], and quadratic convergence has been shown under some assumptions on the subsystem solvers.
7 Numerical Results

The numerical methods presented so far will be demonstrated [22] on a little two-dimensional example shown in Fig. 1, which is nevertheless quite challenging, and was introduced in [38]. It is a rigid block with a thin elastic appendage, in an incompressible but viscous flow from the left. The bluff shape of the square block induces vortices, which develop along the elastic appendage. Although the setup is symmetric, the vortices start to develop unsymmetrically, alternating from the top and from the bottom. As there are strong pressure variations in the vortices, this interacts with the elastic appendage, which in turn starts oscillating quite strongly. This is the situation shown in Fig. 1, together with the pressure field in the flow.

In Fig. 2 we show the response of the tip of the elastic appendage. This was computed with the strong coupling algorithm introduced in section 5 and section 6. If we solve this problem with the same time step but with an explicit staggering or so called weak coupling method, we observe the response in Fig. 3,

Figure 3: Displacement Response for Weak Coupling
and one immediately recognises the instability of the numerical scheme.

![Graph showing iteration count for block-Gauss-Seidel and block-Newton methods.](image)

**Figure 4: Iteration Count for block-Gauss-Seidel and block-Newton**

In another test, we compare the block-Gauss-Seidel method as described in section 5 with the block-Newton method from section 6. For the same configuration as before, we compare the number of iterations of either method in each time step in Fig. 4. The superior convergence characteristic of the block-Newton method is obvious. But as the block-Newton method is more expensive to perform per iteration, this does not necessarily translate into an overall performance gain — of course apart from those cases where the block-Newton converges and the block-Gauss-Seidel method does not. Therefore, as a measure of numerical expenditure, we count the number of times the subsystem solvers have to be called per time step. This is shown in Fig. 5, and again the advantage is on the block-Newton side.
8 Conclusion

For the problem of fluid-structure interaction, we have proposed a partitioned method to compute the transient response. We have proposed a general framework to formulate this coupling problem as a differential algebraic equation (DAE). For the sake of stability, implicit time stepping methods are considered.

The global system to be solved in each time step is approached with the solvers for the individual subsystems. The nonlinear block-Jacobi and nonlinear block-Gauss-Seidel methods come very naturally. We have discussed the problems which arise with this approach. We propose to solve the global implicit equations with an approximate block-Newton method. The linear system which arises in each iteration is solved via the subsystem solvers and a Krylov iterative method.

This iterative method only requires the possibility to use the subproblem solvers in an iteration-by-iteration fashion. The computation of the derivatives is approximated by the subproblem solvers and by finite differencing. These methods are faster both in number of iterations and in numerical effort; additionally
they have superior convergence characteristics. Needless to say that the general ideas about formulating and solving coupling problems are not specific to fluid-structure interaction and can be used in other circumstances as well.

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