A finite element solver and energy stable coupling for 3D and 1D fluid models

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Abstract

The paper develops a solver based on a conforming finite element method for a 3D–1D coupled incompressible flow problem. New coupling conditions are introduced to ensure a suitable bound for the cumulative energy of the model. We study the stability and accuracy of the discretization method, and the performance of some state-of-the-art linear algebraic solvers for such flow configurations. Motivated by the simulation of the flow over inferior vena cava (IVC) filter, we consider the coupling of a 1D fluid model and a 3D fluid model posed in a domain with anisotropic inclusions. The relevance of our approach to realistic cardiovascular simulations is demonstrated by computing a blood flow over a model IVC filter.

1. Introduction

Coupling a 3D fluid flow model and a system of hyperbolic equations posed on a 1D graph is a well established approach for numerical simulations of blood flows in a system of vessels [32]. Such a geometric multiscale strategy is particularly efficient, when the attention to local flow details and the qualitative assessment of global flow statistics are both important. The relevance to cardiovascular simulations and challenging mathematical problems of coupling parabolic 3D and hyperbolic 1D equations put 3D–1D flow problems in the focus of intensive research. Thus, the coupling of a 3D fluid/structure interaction problem with a reduced 1D model merged to outflow boundary, which acts as an absorbing device, was studied in [15]. The coupling of a 3D fluid problem with multiple downstream 1D models in the context of a finite element method was considered in [38]. In [33], a system of a 3D fluid/structure interaction problem and a 1D finite element method model of the whole arterial tree was implemented to model the carotid artery blood flow; and in [6] a unified variational formulation for multidimensional models was introduced. A splitting method, extending the pressure-correction scheme to 3D–1D coupled systems, was studied in [26].

In most of these studies, the 3D model was a generic fluid-elasticity or rigid fluid model, while numerical validations were commonly done for cylindrical type 3D domains (with rigid or elastic walls); several authors considered geometries with bifurcation [38,33] or constrained geometries (modeling a stenosed artery) [6]. More complicated geometries occur in simulations of blood flows, if one is interested in modeling the effect of endovascular implants, such as inferior vena cava (IVC) filters. In numerical simulations, a part of a vessel with an intravenous filter leads to the computational 3D domain with strongly anisotropic inclusions. A downstream flow behind the implant may exhibit a complex structure with traveling vortices, swirls, and recirculation regions (the latter may occur if plaque is captured by the filter). Moreover, the IVC blood flow is strongly influenced by the contraction of the heart, and both forward (towards the heart) and reverse (from the heart) flows occur within one cardiac cycle. Downstream coupling conditions for such flows may be a delicate issue. Thus, the flow over an IVC filter is an interesting and challenging problem for a 3D–1D flow numerical solver.

The coupling conditions of 3D and 1D fluid models and their properties were studied by several authors. Coupling conditions and algorithms based on subdomain iterations were introduced in [15], and the stability properties of each subproblem were analyzed separately. The first analysis of two models together was done in [16]. In that paper, it was noted that if the Navier–Stokes equations are taken in the rotation form and suitably coupled with a 1D downstream flow model, then one can show a bound for the joint energy of the system. It is, however, well known that using a
finite element method for the rotation form of the Navier–Stokes equations needs special care [22], and setting appropriate outflow boundary conditions can be an issue. In the present paper, we introduce an energy consistent coupling with a 1D model for the convection form of the Navier–Stokes equations. The joint energy of a coupled 3D–1D model is appropriately balanced and dissipates for viscous flows.

Handling highly anisotropic structures is a well-known challenge in numerical flow simulations and analysis. There are only a few computational studies addressing the dynamics of blood flows in vessels with implanted filters. Recently, Vassilevski et al. [34] numerically approached the problem of intravenous filter optimization using a finite-difference method on octree cartesian meshes to resolve the geometry of implants. In that paper, it was also discussed how the effect of an implant can be accounted in a 1D model through a modification of a vessel wall state equation (see also [36,37] for the development of this method for atherosclerotic blood vessels). In the present paper, we take another approach and locally resolve the full 3D model, while keeping the state equation unchanged. We report on a finite element method both by considering the accuracy of the finite element method for the rotation form of the Navier–Stokes equations. The joint energy of a coupled 3D–1D model is appropriately balanced and dissipates convection form of the Navier–Stokes equations. The joint energy needs special care [22], and setting appropriate outflow boundary conditions are homogeneous, the following energy balance holds:

\[ \frac{\rho}{2} \int_{\Omega} |\nabla \mathbf{u}|^2 + \frac{1}{2} |\nabla \mathbf{u}|^2 + \int_{\Gamma_{\text{out}}} \left( \frac{\rho}{2} |\nabla \mathbf{u} - \mathbf{u}| \cdot \mathbf{n} \right) \cdot \mathbf{u} ds = 0. \]

Here \( \mathbf{n} \) is the outward normal vector to \( \partial \Omega \). The system is also supplemented with initial condition \( \mathbf{u} = \mathbf{u}_0 \) (\( \text{div} \mathbf{u}_0 = 0 \)) for \( t = 0 \) in \( \Omega \).

We remark that the notion of ‘inflow’ and ‘outflow’ boundary is used here and further in the text conventionally, since the inequalities \( \mathbf{u} \cdot \mathbf{n} < 0 \) or \( \mathbf{u} \cdot \mathbf{n} > 0 \) are not necessarily pointwise satisfied on \( \Gamma_{\text{in}} \) or \( \Gamma_{\text{out}} \), respectively. In applications we consider, the mean flux, \( \mathbf{u} \cdot \mathbf{n} \) averaged in space and in time, is expected to be negative at \( \Gamma_{\text{in}} \) and positive at \( \Gamma_{\text{out}} \). However, for certain \( t \in [0, T] \), the flux \( \int_{\Gamma_{\text{in}}} \mathbf{u} \cdot \mathbf{n} ds \) may take positive values at \( \Gamma_{\text{in}} \) (negative at \( \Gamma_{\text{out}} \)).

If the solution to (1) is sufficiently smooth and the inflow boundary conditions are homogeneous, then solutions to (1) satisfy the following inequality holds:

\[ \int_{\Gamma_{\text{out}}} |\nabla \mathbf{u}|^2 \cdot \mathbf{u} ds \geq 0 \quad \forall t \in [0, T], \]

then solutions to (1) satisfy the a priori energy inequality and this opens possibilities for showing partial well-posedness results. Even though, the assumption (2) was used for the purpose of analysis in the literature on 3D–1D blood flow models (see [15,16]), it is hard to verify (2) for practical flows. Moreover, the inequality (2) no longer holds if reverse flows occur, as, for example, happens in IVC [27,39].

2. The 1D model

A one-dimensional model can be derived from the Navier–Stokes equations posed in a long axisymmetric elastic pipe by integrating over cross section, making some simplifying assumptions and considering integral average quantities as unknowns, see, e.g., [1,32]. Let \( \omega(t,x) \) be the cross section of the pipe normal to \( x, S(t,x) \) is the area of \( \omega(t,x) \) and \( u(t,x) \) is the axial velocity. Introduce the averaged variables: the mean axial velocity \( \bar{u} \) and the mean pressure:

\[ \bar{u}(t,x) = S^{-1}(t,x) \int_{\omega(t,x)} u(t,x) ds, \quad \bar{p}(t,x) = S^{-1}(t,x) \int_{\omega(t,x)} p(t,x) ds. \]

We consider the model given by the following system of equations for unknowns \( u, p, S \):

\[
\begin{align*}
\frac{\partial \bar{S}}{\partial t} + \frac{\partial (\bar{S} \bar{u})}{\partial x} &= \varphi(t,x,S,\bar{u}) \\
\frac{\partial \bar{u}}{\partial t} + \frac{\bar{S}^2 \bar{u} \varphi(t,x,S,\bar{u})}{\partial x} &= \psi(t,x,S,\bar{u}) \quad \text{for } x \in [0,1] \\
\bar{p} - \bar{p}_{\text{ext}} &= \rho \bar{S}^2 \psi(t,x,S,\bar{u})
\end{align*}
\]

For initial conditions, the mean velocity and the cross section area are prescribed, \( u|_{t=0} = u_0, S|_{t=0} = S_0 \). Here \( p_{\text{ext}} \) is the external pressure, \( \varphi(t,x,S,\bar{u}) \) is a function modeling the source or sink of the fluid, as may be required in hemodynamic simulations, if a blood loss happens in a vessel. Further, we assume \( \varphi = 0 \) and \( p_{\text{ext}} = 0 \), so from now \( p \) has the meaning of the difference between the fluid pressure and the external pressure. The term \( \psi(t,x,S,\bar{u}) \) accounts for external forces, such as gravity or friction. Following [1,28], we set

\[ \psi = -16 \mu v \bar{S}^2 |\bar{S}'| S^{-1}, \quad \bar{S} = \bar{S}^{-1} S. \]

Here \( v \) is the viscosity coefficient, \( d \) is the pipe diameter, \( \bar{S} \) is the reference area (in the hemodynamic applications \( \bar{S} \) is the cross section area of a vessel at rest) and

\[ \eta(\bar{S}) = \left\{ \begin{array}{ll} 2, & \text{for } \bar{S} > 1, \\ \bar{S}^{-1}, & \text{for } \bar{S} \leq 1. \end{array} \right. \]
The last equation in (3) relates the pressure to the cross section area. The function $f$ is defined by the elasticity model of the pipe walls, $c_0$ is the elasticity model parameter. We use the one from [28]:

$$f(S) = \begin{cases} 
\exp((S^{-1} - 1) - 1), & \text{for } S > \bar{S}, \\
\ln(S^{-1}), & \text{for } S \leq \bar{S}.
\end{cases}$$  \hfill (5)

Other algebraic defining relations linking the mean pressure and the cross section area are known from the literature, see, e.g., [32]. They are equally well suited for the purpose of this paper.

In [15], the authors derived the energy equality for the one-dimensional fluid model in $Q$, $S$, $p$ variables, where $Q = S\bar{u}$ is the flux. Up to possibly different choices of $\psi(t, x, S, \bar{u})$ and $f(S)$, the formulation in [15] is equivalent to (3). Written in terms of $\bar{u}$, $S$, and $p$, the energy equality for (3) is (recall that we assumed $\varphi = 0$):

$$\frac{d}{dt} \mathcal{E}_{1D}(t) - \rho \int_0^1 \mathcal{S}\psi(t, x, \bar{S}, \bar{u}) \, d\bar{u} = -S\bar{u}(\bar{p} + \frac{1}{2} \bar{u}^2)|_0^1,$$  \hfill (6)

with the energy functional

$$\mathcal{E}_{1D}(t) = \frac{\rho}{2} \int_0^1 \bar{u}^2 \, dx + \rho c^2 \int_0^1 \int_{\bar{S}}^{\bar{S}} f(s) \, ds \, dx.$$ 

For $f(S)$ given in (5), the second term in the definition of $\mathcal{E}_{1D}(t)$ is always positive, making $\mathcal{E}_{1D}(t)$ positive for all $t > 0$. The choice of $\psi(t, x, S, \bar{u})$ in (4) ensures that the second term on the left-hand side of (6) is positive as well. Thus, for the homogeneous boundary conditions the energy of the 1D model dissipates: $\frac{d}{dt} \mathcal{E}_{1D}(t) < 0$.

System (3) is hyperbolic and can be integrated along characteristics. To see this, we write (3) in the divergence form:

$$\frac{\partial V}{\partial t} + \frac{\partial F(V)}{\partial x} = g,$$

with $V = (S, \bar{u}, F = \{S\bar{u}, \frac{\bar{u}^2}{2} + \frac{\bar{p}}{\bar{S}}\}$, $g = \{0, \psi\}$. Denote by $\mathbf{w}_L$ and $\lambda_i$ ($i = 1, 2$) the left eigenvectors and eigenvalues of the Jacobian $A = \frac{\partial F}{\partial V}$. One finds

$$\lambda_i = \begin{cases} 
\bar{u} + (-1)^i c_0 \sqrt{\bar{S}/S} \exp((S^{-1} - 1) - 1), & \text{if } S > \bar{S}, \\
\bar{u} + (-1)^i c_0, & \text{if } S \leq \bar{S},
\end{cases}$$

and

$$\mathbf{w}_L = \begin{cases} 
c_0 \sqrt{(1/\bar{S}) \exp((S^{-1} - 1) \sqrt{\bar{S}})} \cdot \left((-1)^i \sqrt{\bar{S}}, 1ight)^T, & \text{if } S > \bar{S}, \\
c_0 \cdot \left((-1)^i \sqrt{\bar{S}}, 1ight)^T, & \text{if } S \leq \bar{S},
\end{cases}$$

Thus, system (3) can be written in the characteristic form:

$$\mathbf{w}_L \left(\frac{\partial V}{\partial t} + \lambda_i \frac{\partial V}{\partial x}\right) = \mathbf{w} g, \quad i = 1, 2.$$  \hfill (7)

Under physiological conditions in hemodynamics, it holds $\bar{u} < c_0$, implying that the eigenvalues $\lambda_i$ have opposite signs. Therefore, $\mathbf{w}_L$ is the incoming characteristic from point $x = 1$ and $\mathbf{w}_L$ is the incoming characteristic from point $x = 0$. Two boundary conditions, one at $x = 0$ and one at $x = 1$, are enough to close the system.

2.3. The coupling of 1D and 3D models

We now consider two domains $\Omega_{1D}$ and $\Omega_{3D}^{\text{down}}$, as shown in Fig. 1. In $\Omega_{3D}^{\text{down}}$ we pose the 1D fluid model and in $\Omega_{1D}$ we pose the full three-dimensional Navier–Stokes equations. The one-dimensional domain $\Omega_{1D}^{\text{down}}$ is coupled to the downstream boundary of $\Omega_{1D}$.

There are several options to define the coupling conditions of 1D and 3D models, as discussed, for example, in [15]. One can ask for the continuity of the mean pressure, the mean axial velocity, the flux, the normal cross section area, the averaged normal stress, or the entering characteristic. In general, the continuity of all these quantities cannot be satisfied simultaneously. A choice has to be made.

One common choice, see, e.g., [33,6], is to impose the continuity of the normal stress and the flux:

$$\left(-\frac{\partial \bar{u}}{\partial n} + \bar{p} n\right)_{\Gamma_{out}} = \bar{p}|_{x=1} n,$$  \hfill (8)

$$\int_{\Gamma_{out}} \mathbf{u} \cdot d\mathbf{s} = S\bar{u}|_{x=1} n.$$  \hfill (9)

This choice is, however, known to be energy inconsistent in the following sense. Assume the homogeneous boundary conditions on $\Gamma_{in}$ and the downstream end of $\Gamma_{down}^{\text{down}}$. Then the cumulative energy balance of the coupled model is

$$\frac{d}{dt} \left(\mathcal{E}_{1D}(t) + \mathcal{E}_{3D}(t)\right) + \nabla \cdot \mathbf{u}^2 + \int_{\Omega_{3D}^{\text{down}}} K(S) \bar{u}^2 \, dx$$

$$\quad = \int_{\Gamma_{out}} \left(\frac{\partial \mathbf{u}}{\partial n} \cdot (\mathbf{p} + \frac{\rho}{2} \mathbf{u}^2) \cdot n - S\bar{u}(\mathbf{p} + \frac{\rho}{2} \mathbf{u}^2)\right)|_{x=1} n,$$  \hfill (10)

with $\mathcal{E}_{3D}(t) = \frac{1}{2} |\mathbf{u}|^2$, $K(S)$ is a positive coefficient defined from (4). Easy to see that for coupling conditions (8)–(9) the right hand side of (10) reduces to

$$\frac{\rho}{2} \left(S\bar{u}^3_{|x=1} - \int_{\Gamma_{out}} |\mathbf{u}|^2 (\mathbf{u} \cdot n) \, ds\right).$$

In general, it is not clear if this quantity is non-positive and thus if the cumulative energy is properly dissipated, as it holds for the full 3D Navier–Stokes equations with the Dirichlet homogeneous boundary conditions. To circumvent this inconsistency, it was suggested in [16] to replace condition (8) by the continuity of so-called total stress:

$$-\frac{\partial \mathbf{u}}{\partial n} \cdot (\mathbf{p} + \frac{\rho}{2} |\mathbf{u}|^2) |_{x=1} n = (\mathbf{p} + \frac{\rho}{2} |\mathbf{u}|^2) |_{x=1} n $$  \hfill (11)

Condition (11) together with the continuity of the flux, i.e. condition (9), makes the right hand side of (10) equal to zero. Hence, the cumulative energy dissipates. However, setting the total stress equal to a constant is not a consistent outflow condition for the simplest Poiseuille flow. Moreover, (11) is the natural boundary condition for the rotation form of the Navier–Stokes equations. Using it with the common convection or conservation forms leads to non-linear coupling conditions and often requires iterative treatment [7]. Although the rotation form is an interesting alternative to the standard convection form, it is still not a standard option in the existing software and its use requires certain care [22,23].

The condition for the normal stress, as in (8), is the natural boundary condition for the commonly used convection and conservation forms of the Navier–Stokes equations. Such a condition has been shown to be surprisingly useful as outflow boundary condition [19]. Thus, instead of keeping (9) and changing (8) to the total stress condition, we retain (8) and instead of (9) assume the conti-
nuity of the linear combination of the fluid flux and the energy flux:
\[ p_d \left( \int_{\Gamma_{up}} \mathbf{u} \cdot \mathbf{n} \, ds + \frac{\rho}{2} \int_{\Gamma_{up}} |\mathbf{u}|^2 (\mathbf{u} \cdot \mathbf{n}) \, ds = (\rho \mathbf{S} \mathbf{u} + \frac{\rho}{2} \mathbf{S} \mathbf{u}^3) |_{x=d}. \right. \] (12)

One easily checks that the combination of (8) and (12) ensures the right-hand side of (10) equal to zero and thus the correct energy balance and energy inequality are valid as formulated in the following theorem.

**Theorem 2.1.** Consider the coupled 1D–3D fluid problem given by (1), (3), (4), (5), and coupling conditions (8), (12). Assume the homogeneous boundary conditions \( u_{in} = 0 \) and \( u_{in} = 0 \). Then a sufficiently smooth solution satisfies the following energy decay property:

\[
E_{3D}(t) + E_{1D}(t) + \frac{1}{\rho} \int_0^t \|
abla \mathbf{u}\|^2 \, dt' + \int_0^t \int_{\Omega_{1D}} K_{in}(S) \mathbf{u}^2 d\mathbf{r}'
= E_{3D}(0) + E_{1D}(0)
\]

for any \( t \in [0, T] \). If \( \psi \) is defined in (4), then \( K_{in}(S) = 16 \nu \psi(S) S \gamma - 2 \).

**Remark 2.1.** Since \( \tilde{p} \) has the meaning of the difference between fluid and external pressures, it can be negative. In this case, more than one value of \( \tilde{u} \) may satisfy (12). To ensure that the coupled model is not defective, one has to prescribe a particular rule for choosing the root of the cubic Eq. (12). In our numerical experiments, we take \( \tilde{u} \) which is the closest to \( |\Gamma_{out}|^{-1} \int_{\Gamma_{out}} \mathbf{u} \cdot \mathbf{n} \).

**Remark 2.2.** The boundary condition (12) is the combination of fluid and energy fluxes and so it does not guarantee to conserve the ‘mass’ of the entire coupled system. Although, we do not observe any perceptible generation or loss of mass in our numerical experiments, it does not necessarily mean that for all problems this effect should be negligible. Actually, one may consider any other linear combination of fluid and energy fluxes coupling on 3D–1D boundary to compromise between energy stability and mass conservation.

In practice, one may also be interested in coupling the 1D fluid model to the upstream boundary of the 3D domain. Hence, we now consider three domains \( \Omega_{up}^{3D}, \Omega_{down}, \) and \( \Omega_{down}^{3D} \), as shown in Fig. 2. In \( \Omega_{up}^{3D} \) and \( \Omega_{down}^{3D} \), the simplified 1D model is posed and in \( \Omega_{down} \), the full three-dimensional Navier–Stokes equations are solved. The domain \( \Omega_{up}^{3D} \) is coupled to the inflow (upstream) boundary of \( \Omega_{1D} \) and \( \Omega_{down}^{3D} \) is coupled to the outflow (downstream) boundary of \( \Omega_{1D} \). The downstream coupling is described above. In the literature, it is common not to distinguish between upstream and downstream boundary conditions. For example, in [6,33] conditions (8), (9) are assumed both on upstream and downstream boundaries. Following this paradigm, one may consider (8), (9) or energy consistent conditions (8), (12) as the coupling conditions on \( \Gamma_{in} \) between 1D model posed in \( \Omega_{up}^{3D} \) and 3D model posed in \( \Omega_{1D} \). Note that in entirely 3D fluid flows simulations, inflow and outflow boundary conditions usually differ. If a numerical approach to 1D–3D problem is based on subdomains splitting (see the next section for an example), then it is appropriate to distinguish between upstream and downstream coupling conditions. Thus, we impose the upstream coupling conditions in such a way that the 3D problem is supplied with the Dirichlet inflow boundary conditions. This is a standard choice for incompressible viscous fluid flows solvers and is especially convenient if third parties or legacy codes are separately used to compute 3D and 1D solutions, and they communicate only through coupling conditions.

For the upstream boundary, \( \Gamma_{in} \), we introduce a reference velocity profile \( \mathbf{g}(\mathbf{x}), \mathbf{x} \in \Gamma_{in} \), such that \( \int_{\Gamma_{in}} \mathbf{g} \cdot \mathbf{n} \, ds = 1 \). Then the boundary condition on \( \Gamma_{in} \) is Dirichlet, given by \( \mathbf{u}_{in} = -2 \mathbf{g} \) on \( \Gamma_{in} \).

Setting \( x = S \mu_{in} \) ensures the continuity of the flux (9) on \( \Gamma_{in} \). If \( x \) is found to satisfy the equation

\[ p_0(x) + \frac{\rho}{2} \int_{\Gamma_{in}} |\mathbf{g}|^2 (\mathbf{g} \cdot \mathbf{n}) \, ds = (\rho \mathbf{S} \mathbf{u} + \frac{\rho}{2} \mathbf{S} \mathbf{u}^3) |_{x=b}, \]

then the coupling condition (12) is valid on \( \Gamma_{in} \). Two more scalar boundary conditions are required for the 1D model in \( \Omega_{up}^{3D} \). We assume that \( \tilde{u} \) or \( \tilde{p} \) are given in \( x = a \) and in \( x = b \) an absorbing condition is prescribed: in computations we set \( (\mu_{in}) = 0 \) in \( x = b \); another reasonable absorbing condition would be setting the incoming characteristic equals zero. On the downstream end of \( \Omega_{down}^{3D} \), we also assume an absorbing boundary condition.

We summarize the properties of the 3D–1D coupling introduced in this section:

- It ensures the energy balance, as stated in Theorem 2.1.
- The inequality (2) is not assumed.
- It can be easy decoupled with splitting methods into the separate 1D problems and the 3D problem with usual inflow-outflow boundary conditions on every time step.

### 3. Discretization and algebraic solver

In this section, we introduce a splitting numerical time-integration algorithm for subdomain splitting. Further, we consider a fully discrete problem and review one state-of-the-art algebraic solver.

Denote by \( \tilde{u}^n, \tilde{p}^n, S^n, \mathbf{u}^n \), and \( p^n \) approximations to the corresponding unknown variables at time \( t = t_n \). Given these approximations, we compute \( u^{n+1}, p^{n+1}, S^{n+1}, \mathbf{u}^{n+1}, \) and \( p^{n+1} \) for \( t = t_{n+1} \) (\( \Delta t = t_{n+1} - t_n \)) in three steps:

**Step 1.** Integrate (7) for \( t \in [t_n, t_{n+1}] \), with given \( u(t_{n+1}) \) on the upstream end of the interval \( \Omega_{up}^{3D} \) and the absorbing downstream condition at \( x = b \).

**Step 2.** Set \( u_0 \) according to (14), with \( \mu(p, S) = (\mu^{n+1}, p^{n+1}, S^{n+1}) \), and compute \( p \) and \( S \) as the linear extrapolations of \( \mu_{in} \) and \( S_{in} \) from times \( t_n \) and \( t_{n+1} \). Solve the linearized Navier–Stokes problem in \( \Omega_{3D} \) for \( u^{n+1}, p^{n+1} \):

\[
\begin{align*}
&\frac{1}{\rho} \left( 3u^{n+1} - 4u^n + u^{n-1} \right) + (2u^n - u^{n-1}) \cdot \nabla u^{n+1} \\
&- \nabla u^{n+1} + \nabla p^{n+1} = f^{n+1}, \\
&\text{div} u^{n+1} = 0, \\
&u^{n+1}|_{in} = u_{in}, \quad u^{n+1}|_{b} = 0, \\
&\left( -S^{n+1} \frac{\partial u^{n+1}}{\partial n} + p^{n+1} \right) |_{out} = \tilde{p} n.
\end{align*}
\]

**Step 3.** Find \( u^{n+1} \) from

\[
(\tilde{p} S^{n+1} + \frac{\rho}{2} S \mu_{out}^{n+1}) |_{x=d} = p(d) \int_{\Gamma_{out}} u^{n+1} \cdot \mathbf{n} \, ds + \frac{\rho}{2} \int_{\Gamma_{up}} u^{n+1} \cdot \mathbf{n} \, ds.
\]
Now, using \( \mathbf{u}^{n+1} \) for boundary condition in \( x = d \) and the absorbing boundary condition in \( x = e \), we integrate (7) for \( t \in [I_n, I_{n+1}] \) to find \( \mathbf{u}^{n+1}, p^{n+1}, S^{n+1} \) in \( \Omega_{2D}^{\text{down}} \).

For the numerical integration of the 1D model equations, we use a first order monotone difference scheme applied to the characteristic form (7), see [35]. To handle the 3D model, one has to solve on every time step the linearized Navier–Stokes equations, also known as the Oseen problem:

\[
\begin{cases}
\begin{align*}
\mathbf{f} & = -\nu \mathbf{u} + (\mathbf{w} \cdot \nabla) \mathbf{u} + \nabla p = \mathbf{f} \quad \text{in} \quad \Omega_{2D}, \\
\text{div} \mathbf{u} &= 0, \\
\mathbf{u}|_{\Gamma_1 \cup \Gamma_2} &= \mathbf{g}, \\
(\text{div} \mathbf{u} - p \mathbf{n})|_{\Gamma_{\text{out}}} &= 0,
\end{align*}
\end{cases}
\]

where \( \mathbf{u} = \mathbf{u}^{n+1}, \mathbf{w} = \mathbf{w}^{n+1} - \mathbf{w}^n, \) the body forces term and the advection velocity field depend on previous time velocity approximations, \( \mathbf{f} = (2\Delta t)^{-1}(4\mathbf{u}^n - \mathbf{u}^{n+1}), \mathbf{w} = (2\mathbf{u}^n - \mathbf{u}^{n+1}), \) and \( \beta = 3(2\Delta t)^{-1} \). Here and in the remainder of this section, we dropped out the time-step dependence \((n+1)^{\text{index}}\) for unknown velocity and pressure.

To discretize the Oseen problem (15), we consider a conforming finite element method. Denote the finite element velocity field \( \mathbf{u}^h \) with \( \mathbf{u}^h \rightarrow \mathbf{u} \) as the optimal boundar y condition setup both for the conforming discretization of the pressure Poisson problem with Neumann boundary conditions:

\[
(\mathbf{L}^p)_{ij} = (\nabla \psi_i, \nabla \psi_j).
\]

Likewise, Neumann boundary conditions are conventionally used to define the pressure convection–diffusion problem on \( \Gamma_h \). However, the optimal boundary conditions setup both for \( \mathbf{L}^p \) and \( \mathbf{A}_p \) depends on the type of the boundary and flow regime, see [14,25].

We use a modified PCD preconditioner defined below. This modification partially obviates the issue of setting pressure boundary conditions and is consistent with the Cahouet–Chabard preconditioner [10], if the inertia terms are neglected. The Cahouet–Chabard preconditioner is the standard choice for the time-dependent Stokes problem and enjoys the solid mathematical analysis in this case [24]. To define the preconditioner, we introduce the discrete advection matrix for continuous pressure approximations as

\[
(N_p)_{ij} = (\mathbf{w} \cdot \nabla \psi_i, \psi_j).
\]

Then the modified pressure convection–diffusion preconditioner (mPCD) is (compare to (22)):

\[
\tilde{S}^{-1} := v\tilde{M}_p^{-1} + (\beta I + N_p)(BM_c^{-1}B^T)^{-1},
\]

where \( M_c \) is a diagonal approximation to the velocity mass matrix.

Regarding the numerical analysis of the algebraic solver used here, we note the following. The eigenvalues bounds of the preconditioned Schur complement:

\[
0 < c_1 \leq |\lambda(\tilde{S}^{-1})| \leq C_1,
\]

were proved for \( \beta = 0 \) and the LBB stable finite elements in [13] and for a more general case in [25]. The constants \( c_1, C_1 \) are independent of the meshsize \( h \), but may depend on the ellipticity, continuity and stability constants in (17)–(19), and thus may depend on the problem parameters. In particular, the pressure stability constant \( \gamma_2 \) and so \( c_1 \) from (23), depends on the geometry of the domain \( \Omega \) [11] (tending to zero for long or narrow domains) and for certain FE pairs \( \gamma_1 \) depends on the anisotropy ratio of a triangulation [2]. Both of these dependencies require certain care in using the approach for com-
puting flows in 3D elongated domains with thin and anisotropic inclusions (prototypical for simulating a flow over an IVC filter). Characterizing the rate of convergence of nonsymmetric preconditioned GMRES [18]. Nevertheless, experience shows that for many linear systems arising in practice, a well-clustered spectrum (away from zero) usually results in rapid convergence of the preconditioned iteration. This said, we should mention that a rigorous proof of the GMRES convergence applied to (20), with block-triangular preconditioner (21), is not available in the literature (except the special case, when $\bar{S}$ is symmetric [21,5]). Thus, the numerical assessment of the approach is of practical interest.

### 4. Accuracy and efficiency assessment

In this section, we validate the accuracy and stability of the solver for the 3D–1D coupled fluid model by (i) comparing the computed discrete solutions against an analytical solution for a problem with simple geometry; (ii) computing the drag coefficient and the pressure drop value for the flow around the 3D circular cylinder. The Taylor–Hood $P2$–$P1$ elements were used for the velocity–pressure approximation. The resulting linear algebraic systems (20) were solved by the preconditioned BiCGstab method. The initial guess in the BiCGstab method was zero on the first time step and equal to $(\mathbf{u}^0, p^0)$ for the subsequent time steps. The stopping criteria was the $10^{-6}$ decrease of the Euclidean norm of the residual.

#### 4.1. Test with analytical solution

First we consider an example with analytical solution. The 3D domain is $\Omega_{3D} = \{ x \in \mathbb{R}^3 \mid x \in (-1, 1), y^2 + z^2 < 1 \}$. The circular cross sections are the inflow and outflow boundaries. Domains $\Omega_{3D}^{up}$ and $\Omega_{3D}^{down}$ are two intervals of length 5. The analytical solution is given by

$$S = \cos(2\pi t) + \bar{S} - 1, \quad u = 1 - \cos(2\pi t), \quad p = c^2 f(S),$$

in $\Omega_{3D}^{up} \cup \Omega_{3D}^{down}$,

$$\mathbf{u} = \left( \frac{1}{2} (1 - \cos(2\pi t))(1 - y^2 - z^2), 0, 0 \right)^T, \quad p = 10(1 - x) + \bar{p},$$

in $\Omega_{3D}$.

### Table 1

Errors to analytical solution in $\Omega_{3D}$ on the sequence of refined meshes. Reduction orders are given in brackets.

| Mesh     | $\max_{x \in \Omega} \|u - u_h\|_{L^2}$ | $(\int_0^T ||\nabla (u - u_h)||^2 dt)^{1/2}$ | $(\int_0^T ||u - u_h||^2 dt)^{1/2}$ | $(\int_0^T ||p - p_h||^2 dt)^{1/2}$ |
|----------|-----------------------------------------|------------------------------------------|----------------------------------|----------------------------------|
| Mesh 1   | 0.37                                    | 0.30                                     | 0.30                            | 8.10E-2                          |
| Mesh 2   | 8.60E-2 (2.11)                          | 6.13E-2 (2.29)                           | 5.97E-2 (2.33)                  | 4.04E-2 (1.00)                   |
| Mesh 3   | 2.66E-2 (1.69)                          | 1.68E-2 (1.87)                           | 1.62E-2 (1.88)                 | 2.02E-2 (1.00)                   |

### Table 2

Errors to analytical solution in $\Omega_{3D}^{up}$ and $\Omega_{3D}^{down}$ on the sequence of refined meshes. Reduction orders are given in brackets.

| Mesh     | $(\int_0^T ||u - u_h||^2 dt)^{1/2}$ | $(\int_0^T ||S - S_h||^2 dt)^{1/2}$ | $(\int_0^T ||u - u_h||^2 dt)^{1/2}$ | $(\int_0^T ||p - p_h||^2 dt)^{1/2}$ |
|----------|----------------------------------|---------------------------------|----------------------------------|----------------------------------|
| Mesh 1   | 0.25                             | 8.84E-4                         | 0.19                             | 7.02E-4                          |
| Mesh 2   | 5.78E-2 (2.11)                   | 2.03E-4 (2.12)                  | 3.67E-2 (2.37)                  | 1.00E-4 (2.81)                   |
| Mesh 3   | 1.43E-2 (2.02)                   | 5.01E-5 (2.02)                  | 5.64E-3 (2.70)                  | 2.95E-5 (1.76)                   |

Based on the energy balance (13), the natural norms for measuring error in $\Omega_{3D}$ are $C(0, T, L^2(\Omega_{3D}))$ and $L^2(0, T, H^1(\Omega_{3D}))$ for velocity and $L^2(0, T, L^2(\Omega_{3D}))$ for pressure. These norms and, additionally, $L^2(0, T, L^1(\Omega_{3D}))$ for velocity error are shown in Table 1. The error norms in the 1D domains coupled to the inflow and the outflow boundaries of $\Omega_{3D}$ are shown in Table 2. We observe the expected second order of convergence for all variables except for pressure in 3D. We remark that the integral in time error norms were computed approximately using the quadrature rule:

$$\int_0^T ||\nabla (u - u_h)||^2 dt \approx \Delta t \sum_{n=1}^N \|\nabla (u(n\Delta t) - u_h^n)\|_{L^2(\Omega_{1D})}^2, \quad N = T(\Delta t)^{-1}.$$
sufficiently small values of the time step $\Delta t$. The results in Table 4 show that for small $\nu$ the preconditioned BiCGstab method fails to converge unless $\Delta t = 0.01$. More robust preconditioners as in [4,5] may reduce the dependence on $\nu$ and $\Delta t$.

4.2. Flow around circular cylinder

Motivated by the simulation of blood flows over an intravenous filter, we experiment with flows in a 3D domain having an inclusion and coupled with 1D model at the outflow boundary. Interesting statistics for such applications are the drag force acting on the cylinder width $C_0$ may reduce the dependence on the cylinder surface $S \subset \Gamma_{wall}$:

$$ C_{drag} = \frac{2}{DHU^2} \int_S \left( \frac{\partial (u \cdot t)}{\partial n} n_x - p_n \right) ds $$

Here $n = (n_x, n_y, n_z)^T$ is the normal vector to the cylinder surface pointing to $\Omega$ and $t = (-n_z, 0, n_x)^T$ is a tangent vector.

For problem P2, the reference velocity in (25) is $U = 2.25 m/s$.

For these benchmark problems, the paper [29] collects several DNS results based on various finite element, finite volume discretizations of the Navier–Stokes equations and the Lattice Boltzmann method. In [29], the authors provided reference intervals, where the statistics are expected to converge. Using a higher order finite element method and locally refined adaptive meshes, more accurate reference values of $C_{drag}$ and $\Delta p$ were found in [8] for the steady state solution (problem P1) and in [3] for unsteady problem P2. For the computations we use two meshes: a ‘coarse’ and a ‘fine’ ones, both adaptively refined towards the cylinder. The coarser mesh is build of 35,803 tetrahedra, which results in 53,061 velocity d.o.f. and 8767 pressure d.o.f. Both coarse and fine mesh consist of regular tetrahedra. The refinement ratio is about 20 and 60 for the coarse and the fine meshes, respectively.

The benchmarks setups do not specify outflow boundary conditions. Hence, on the outflow boundary we apply the 3D–1D coupling using the new conditions (8), (12) so that numerical performance of the coupling can be verified.

The statistics of interest are the following:

- The difference $\Delta p = p(x_i) - p(x_j)$ between the pressure values in points $x_i = (0.2, 0.205, 0.55)$ and $x_j = (0.2, 0.205, 0.45)$.
- The drag coefficient given by an integral over the surface of the cylinder $S \subset \Gamma_{wall}$:

$$ C_{drag} = \frac{2}{DHU^2} \int_S \left( \frac{\partial (u \cdot t)}{\partial n} n_x - p_n \right) ds $$

For problem P2, the reference velocity in (25) is $U = 2.25 m/s$.

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We remark that the fine mesh has four times as many tetrahedra touching the cylinder as the coarse mesh. The time steps are $\delta t = 0.002$ and $\delta t = 0.001$ for the coarse and the fine meshes, respectively.

The first step in Tables 5 and 6 results for problems P1 and P2 obtained with the coarse and the fine meshes. For all settings, the computed values are within “reference intervals” from [29] (except $C_{drag}^{max}$ for problem P2, but in this case the upper reference bound appears to be tough). The computed drag coefficients were well within 1% of reference values and pressure drop within 2%. This is a good result for the number of the degrees of freedom involved. Indeed, the results shown in [8,3,29] for meshes with about the same number of degrees of freedom show comparable or worse accu-

![Fig. 3. The flow domain $\Omega_{3D}$ in the benchmark problem of the channel flow past a 3D cylinder.](image-url)
racy. In Fig. 4, we show the computed evolution of the drag coefficient for problem P2 and compare it to the reference results. The computed drag coefficients match the reference curve very well. We conclude that the conforming finite element method with the coupling outflow conditions is a reliable and stable approach for the simulation of such flow problems.

5. Simulations of a flow over a model IVC filter

The development of endovascular devices is the challenging problem of cardiovascular medicine. One example is the design of vascular filters implanted in inferior vena cava (IVC) to prevent a blockage of the main artery of the lung or one of its branches by a substance that has traveled from elsewhere in the body through the bloodstream. The filter is typically made of thin rigid metal wires as illustrated in Fig. 5 (left). Numerical simulation is an important tool that helps in finding an optimal filter design. Thin and anisotropic construction of a IVC filter requires adaptive grid refinement and makes computations of flows in such domains not an easy task. In this section, we demonstrate the ability of the numerical method to treat such problems in a stable way. One statistic of interest here is the drag force experienced by a filter. We recall that in this paper we do not account for the elastic properties of the vessel walls, which are otherwise important in practice.

We consider a segment (4.5 cm long) of IVC with elliptic cross section $1.6 \times 2.4$ cm. The filter is placed on the 0.5 cm distance from inflow, it is 2 cm long and the diameter of its 12 wire legs is 0.5 mm. Blood is assumed to be incompressible fluid with dynamic viscosity equal to 0.0055 Pa s and density equal to 1 g/cm$^3$.

A blood flow in IVC is strongly influenced by the contraction of the heart. The IVC have pulsatile waveforms with two peaks and reverse flow [39] occurring on every cardiac cycle. We consider the Doppler blood flow waveforms of IVC reported in [39] and approximate them by a smooth periodic function plotted in Fig. 5 (right). Note that the presence of significant reverse flows in IVC differs this problem from computing artery flows, where such phenomenon does not typically occur.

On the inflow and outflow, the 3D vessel is coupled to 1D models as described in Section 2.3. Each 1D model consists of Eqs. (3)–(5) posed on intervals of 5 cm length. Periodic velocity with waveform as shown in Fig. 5 is prescribed on the upstream part of the 1D model coupled to $I_m$. The maximum 1D model velocity of 12 cm/s yields the maximum inlet velocity in $\Omega_{1D}$ of $0.2$.
about 24 cm/s. This agrees with the measurements in [39]. The coupling conditions are the same regardless of the mean flow direction.

The mesh was adapted towards the filter, so the ratio of largest and smallest element diameters was about $1.1e+2$, the maximum elements anisotropy ratio was about 14. The resulting mesh is illustrated in Fig. 6. The time step in 3D model was set equal to 0.001 s. The BiCGstab iterative method, with preconditioner (21) was used to solve discrete Oseen subproblems. The stopping criterion was the reduction of the residual by the factor of $10^6$. The average number of linear iterations on every time step was about 35. We found that choosing time step larger for this problem, leads to the significant increase of the linear iteration counts and makes ‘long time’ computations non-feasible.

We visualize the computed solutions in Fig. 7 by showing the values of the $x$-component of the velocity in several cutplanes orthogonal to $x$-axis. Behind the filter the velocity $x$-component eventually has negative values, indicating the occurrence of circulation zones and ‘returning’ flows. Note that the solution behind the filter is no longer axial-symmetric: a perturbation to solution induced by non-symmetric tetrahedral grid is sufficient for the von Karman type flow instability to develop behind the filter.

Fig. 8 (left) shows the time evolution of the drag force experienced by the filter. After the instantaneous start, the flow needs few cycles to obtain the periodic regime. In general, the drag force follows the pattern of the inflow waveform. In particular, the filter experiences forces both in downstream and upstream directions at different periods of the cardiac cycle. The right plot in Fig. 8 shows the mean axial velocity in the middle point of the 1D model before and after the 3D domain with cava filter. It is remarkable that after few cycles, when the flow is periodic, the waveforms in the 1D domains coupled to upstream and downstream boundaries are very close. This suggests that the coupling conditions are efficient in conserving averaged flow quantities such as mean flux.

6. Conclusions

We reviewed the 3D and 1D models of fluid flows and some existing coupling conditions for these models. New coupling conditions were introduced and shown to ensure a suitable bound for the cumulative energy of the model. The conditions were found to perform stable in several numerical tests with analytical and benchmark solutions. For the example of the flow around IVC filter, the coupled numerical model was found to capture the periodic flow regime and correct 1D waveforms before and after 3D domain. The model was able to handle ‘opposite direction’ flow, i.e. the flow where the ‘upstream’ boundary (boundary with Dirichlet boundary conditions) becomes the outflow boundary for a period of time. The preconditioned BiCGstab method with one state-of-the-art preconditioner applied to the linearized finite element Navier–Stokes problem performs well. However, often the time step should be taken small enough to make the linear solver converge.
sufficiently fast. Overall, the coupled 3D–1D model together with the conforming finite element method and preconditioned iterative strategy was demonstrated as a reliable tool for the simulation of such biological flows as the flow over an inferior vena cava filter.

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Fig. 8. Left: The evolution of the drag force \( (g \text{ sm/s}^2) \) for the IVC filter. Right: The evolution of the mean axial velocity \( (\text{sm/s}) \) in the middle point of the 1D model before and after the 3D domain with IVC filter.

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References


