Adjoint-based methods for optimization and goal-oriented error control applied to fluid-structure interaction: implementation of a partition-of-unity dual-weighted residual estimator for stationary forward FSI problems in deal.II

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Abstract: In this work, we implement goal-oriented error control and spatial mesh adaptivity for stationary fluid-structure interaction. The a posteriori error estimator is realized using the dual-weighted residual method in which the adjoint equation arises. The fluid-structure interaction problem is formulated within a variational-monolithic framework using arbitrary Lagrangian-Eulerian coordinates. The overall problem is nonlinear and solved with Newton’s method. We specifically consider the FSI-1 benchmark problem in which quantities of interest include the elastic beam displacements, drag, and lift. The implementation is provided open-source published on github https://github.com/tommewick/goal-oriented-fsi. Possible extensions are discussed in the source code and in the conclusions of this paper.

1 Introduction

Fluid-structure interaction (FSI) is well-known and a prime example of a multiphysics problem. It combines several challenges such as different types of partial-differential equations (PDE), interface-coupling, nonlinearities in the equations and due to coupling, Lagrangian and Eulerian coordinates. These result into typical numerical challenges such as robust spatial discretization (in particular for the moving interface), robust time-stepping schemes, efficient and robust linear and nonlinear solution algorithms. Computational work include different coupling concepts, space-time multiscale, reduced order modeling, optimal control, parameter estimation, uncertainty quantification, and efficient solver developments.

In this work, the main objective is the application and open-source implementation of goal-oriented a posteriori error control using the dual-weighted residual (DWR) method. For applications in fluid-structure interaction, we refer to a recent overview of our own work using the adjoint FSI equation in goal-oriented error estimation and optimization was done in [60]. In [49] a variational localization using a partition-of-unity (PU) was proposed, facilitating the application to coupled problems such as fluid-structure interaction. In view of increasing initiatives of open-source developments, another purpose of this work is to provide a documented open-source code. To this end, a stationary fluid-structure interaction problem is considered in order to explain the main steps of a PU-DWR estimator. The problem is formulated within a monolithic framework using arbitrary Lagrangian-Eulerian (ALE) coordinates. For some well-posedness results of
such stationary FSI problems, we refer to \cite{31, 61}. Together with the goal functional under consideration, the FSI formulation serves as PDE constraint and the Lagrange formalism can be applied. Specifically, the monolithic formulation yields a consistent adjoint equation.

For the monolithic, stationary, FSI formulation we follow \cite{17, 56} and for the PU-DWR error estimator, we follow \cite{49}. The basis of our programming code is \cite{28} (see also updates on github\footnote{https://github.com/tommeswick/fsi}) and we take some ideas from deal.II \cite{1, 2} step-14\footnote{https://www.dealii.org/current/doxygen/deal.II-step_14.html}. Our resulting code can be found on github\footnote{https://github.com/tommeswick/goal-oriented-fsi}.

2 Variational-monolithic ALE fluid-structure interaction

2.1 Modeling

For the function spaces in the (fixed) reference domains $\hat{\Omega}, \hat{\Omega}_f, \hat{\Omega}_s$, we define $V := H^1(\hat{\Omega})^d$. In the fluid and solid domains, we define further:

\[
\hat{L}_f := L^2(\hat{\Omega}_f), \quad \hat{L}_f^0 := L^2(\hat{\Omega}_f)/\mathbb{R}, \quad \hat{V}_f := \{\hat{v}_f \in H^1(\hat{\Omega}_f)^d : \hat{v}_f = 0 \text{ on } \hat{\Gamma}_{in} \cup \hat{\Gamma}_D\},
\]

\[
\hat{V}_{f,s} := \{\hat{u}_f \in H^1(\hat{\Omega}_f)^d : \hat{u}_f = \hat{u}_s \text{ on } \hat{\Gamma}_i, \quad \hat{u}_f = 0 \text{ on } \hat{\Gamma}_{in} \cup \hat{\Gamma}_D \cup \hat{\Gamma}_{out}\},
\]

\[
\hat{V}_{f,s}^0 := \{\hat{\psi}_f \in H^1(\hat{\Omega}_f)^d : \hat{\psi}_f = 0 \text{ on } \hat{\Gamma}_i \cup \hat{\Gamma}_{in} \cup \hat{\Gamma}_D \cup \hat{\Gamma}_{out}\},
\]

\[
\hat{V}_s := \{\hat{u}_s \in H^1(\hat{\Omega}_s)^d : \hat{u}_s = 0 \text{ on } \hat{\Gamma}_D\}.
\]

As stationary FSI problem in variational-monolithic ALE form, we have \cite{56}[p. 29]:

Problem 2.1. Find $\{\hat{v}_f, \hat{u}_f, \hat{u}_s, \hat{p}_f\} \in \{\hat{V}^0_f + \hat{V}^0_{f,s}\} \times \{\hat{u}^0_f + \hat{V}^0_{f,s}\} \times \{\hat{u}^0_s + \hat{V}^0_s\} \times \hat{L}_f^0$, such that

\[
(\hat{\rho}_f \hat{J}(\hat{F}^{-1}\hat{v}_f \cdot \hat{n}), \hat{\psi}_f)_{\hat{\Omega}_f}
\]

\[
+ (\hat{J}\hat{\sigma}_f \hat{F}^{-T} \cdot \hat{\nabla} \hat{\psi}^\nu)_{\hat{\Omega}_f} - (\hat{g}_f, \hat{\psi}^\nu)_{\hat{\Gamma}_N} - (\hat{\rho}_f \hat{f}_f, \hat{\psi}^\nu)_{\hat{\Gamma}_D} = 0 \quad \forall \hat{\psi}^\nu \in \hat{V}^0_{f,s},
\]

\[
(\hat{F}^{-T} \cdot \hat{\nabla} \hat{\psi}^\nu)_{\hat{\Omega}_f} - (\hat{\rho}_s \hat{f}_s, \hat{\psi}^\nu)_{\hat{\Omega}_s} = 0 \quad \forall \hat{\psi}^\nu \in \hat{V}^0_s,
\]

\[
(\hat{\sigma}_{\text{mesh}} \cdot \hat{\nabla} \hat{\psi}^\nu)_{\hat{\Omega}_f} + (\hat{u}_s \cdot \hat{n}, \hat{\psi}^\nu)_{\hat{\Gamma}_i} = 0 \quad \forall \hat{\psi}^\mu \in \hat{V}^0_{s, \hat{\Gamma}_i},
\]

\[
(\hat{\div}(\hat{J}\hat{F}^{-1}\hat{v}_f), \hat{\psi}^\nu)_{\hat{\Omega}_f} = 0 \quad \forall \hat{\psi}^\nu \in \hat{L}^0_f,
\]

with $\hat{F} = \hat{I} + \hat{\nabla} \hat{u}_f$, $\hat{J} = \det(\hat{F})$, $\hat{\sigma}_f = -\hat{\rho}_f \hat{I} + \hat{\rho}_f \hat{v}_f \hat{\nabla} \hat{y} \hat{F}^{-1} + \hat{F}^{-T} \hat{\nabla} \hat{v}_f$, $\hat{\Sigma} = 2\mu_s \hat{E} + \lambda_s tr(E) \hat{I}$, $\hat{E} = 0.5(\hat{F}^T \hat{F} - \hat{I})$, $\hat{\sigma}_{\text{mesh}} = \alpha_u \hat{\nabla} \hat{u}_f$, volume forces $\hat{f}_f$ and $\hat{f}_s$ (both zero in this work), flow correction term $\hat{f}_f$ (do-nothing \cite{33}), densities $\hat{\rho}_s, \hat{\rho}_f$, kinematic viscosity $\nu_f$, and the Lamé parameters $\mu_s, \lambda_s$. All explanations are provided in \cite{56}[Chapter 3].

2.2 Discretization and numerical solution

For spatial discretization, a conforming Galerkin finite element scheme on quadrilateral mesh elements is employed \cite{12}. Specifically, we use $Q_2^s$ elements for $\hat{v}$ and $\hat{u} := \hat{u}_f + \hat{u}_s$, and $Q_1^f$ elements for $\hat{p}$. For the flow problem $(\hat{v}, \hat{p})$, this is the well-known inf-sup stable Taylor-Hood element; see e.g., \cite{29}. Due to variational-monolithic coupling and globally-defined finite elements, the fluid pressure must be extended to the solid domain, which is achieved via $\alpha_u[(\hat{\nabla} \hat{p}_s, \hat{\nabla} \hat{\psi}^p) + (\hat{\rho}_s, \hat{\psi}^p)]$, and $\alpha_u$ (as before) small, positive. This is only for convenience, an alternative is to work with the FE\_NOTHING element in deal.II. The nonlinear problem is solved with Newton’s method. Therein, for simplicity in this work, we utilize a sparse direct solver \cite{14}. For algorithmic descriptions of our implementation, we refer to \cite{56}.
3 PU-DWR goal-oriented error control

The Galerkin approximation reads: Find $\hat{U}_h = \{\hat{v}_{f,h}, \hat{u}_{f,h}, \hat{u}_{s,h}, \hat{p}_{f,h}\} \in \hat{X}^0_{h,D}$, where $\hat{X}^0_{h,D} := \{\hat{v}_{f,h}^D + \hat{V}_{f,\Sigma,h} \} \times \{\hat{u}_{f,h}^D + \hat{V}_{f,\Sigma,h} \} \times \{\hat{u}_{s,h}^D + \hat{V}_{s,h} \} \times \hat{L}^0_{f,h}$, such that

$$A(\hat{U}_h)(\hat{\Psi}_h) = \hat{F}(\hat{\Psi}_h) \quad \forall \hat{\Psi}_h \in \hat{X}_h,$$

where $\hat{X}_h$ is the test space with homogeneous Dirichlet conditions.

3.1 Goal functional

The solution $\hat{U}_h$ is used to calculate an approximation $J(\hat{U}_h)$ of the goal-functional $J(\hat{U}) : \hat{X} \to \mathbb{R}$. This functional is assumed to be sufficiently differentiable. The drag value as goal functional reads

$$J(\hat{U}) := \int_{\hat{S}} \hat{\sigma}_f \hat{F}^{-T} \hat{n}_f \hat{d}s,$$

where $\hat{n}_f$ is the outward point normal vector of the cylinder boundary $\hat{S}$ and the FSI interface $\hat{\Gamma}_i$. Moreover, $\hat{d}$ is a unit vector perpendicular to the mean flow direction. For the drag, we use $\hat{d} = (1, 0)$.

3.2 Error representation

We use the (formal) Euler-Lagrange method, to derive a computable representation of the approximation error $J(\hat{U}) - J(\hat{U}_h)$. The task is: Find $\hat{U} \in \hat{X}_D^0$ such that

$$\min \{ J(\hat{U}) - J(\hat{U}_h) \} \quad \text{s.t.} \quad \hat{A}(\hat{U})(\hat{\Psi}) = \hat{F}(\hat{\Psi}) \quad \forall \hat{\Psi} \in \hat{X},$$

from which we obtain the optimality system

$$\mathcal{L}_\delta^p(\hat{U}, \hat{Z})(\delta \hat{Z}) = \hat{F}(\delta \hat{Z}) - \hat{A}^T(\delta \hat{Z}) = 0 \quad \forall \delta \hat{Z} \in \hat{X}, \quad \text{(Primal problem)},$$

$$\mathcal{L}_\delta^a(\hat{U}, \hat{Z})(\delta \hat{U}) = J^a(\delta \hat{U}) - \hat{A}_h^a(\delta \hat{U}, \hat{Z}) = 0 \quad \forall \delta \hat{U} \in \hat{X}, \quad \text{(Adjoint problem)}.$$ Using the main theorem from [36], we obtain:

**Theorem 3.1.** We have the error identity:

$$J(\hat{U}) - J(\hat{U}_h) = \frac{1}{2} \rho(\hat{U}_h)(\hat{Z} - \hat{\Phi}_h) + \frac{1}{2} \rho^*(\hat{U}_h, \hat{Z}_h)(\hat{U} - \hat{\Psi}_h) + \mathcal{R}^{(3)}_h,$$

for all $\{\hat{\Psi}_h, \hat{\Phi}_h\} \in \hat{X}_h \times \hat{X}_h$ and with the primal and adjoint residuals:

$$\rho(\hat{U}_h)(\hat{Z} - \hat{\Phi}_h) := -A(\hat{U}_h)(\cdot) + \hat{F}(\cdot),$$

$$\rho^*(\hat{U}_h, \hat{Z}_h)(\hat{U} - \hat{\Psi}_h) := J^a(\hat{U}_h)(\cdot) - A^a(\hat{U}_h)(\cdot, \hat{Z}_h) + \hat{F}(\cdot).$$

The remainder term is $\mathcal{R}^{(3)}_h$ is of cubic order. This error identity can be used to define the error estimator $\eta$, which can be further utilized to design adaptive schemes.

**Corollary 3.2** (Primal error). The primal error identity reads:

$$J(\hat{U}) - J(\hat{U}_h) = \rho(\hat{U}_h)(\hat{Z} - \hat{\Phi}_h) + \mathcal{R}^{(2)}_h.$$
3.3 Adjoint equation, discretization, and numerical solution

The adjoint equation reads: Find $\hat{z} = (\hat{z}^v, \hat{z}^u, \hat{z}^p) \in \hat{X}$ such that

$$J'(\hat{U})(\hat{\Phi}) = \hat{A}'_G(\hat{U})(\hat{\Phi}, \hat{z}) \quad \forall \hat{\Phi} \in \hat{X},$$

and the explicit form can be found in [56, 60].

For the discretization, we briefly mention that higher-order information for the adjoint solution must be employed due to Galerkin orthogonality; in this work $\hat{X}_h \subset \hat{X}^{(2)}_h \subset \hat{X}$. For simplicity, this is realized with global-higher order finite elements and in order to ensure again inf-sup stability, we use $Q_h^1$ elements for $\hat{z}^v$ and $\hat{z}^u$, and $Q_h^2$ elements for $\hat{z}^p$. It is clear that this is an expensive choice. For the numerical solution, the same solvers as for the primal problem are taken (see Section 2.2), namely a Newton-type method and sparse direct solver. Since the adjoint problem is linear, Newton’s method converges in one step. This is a trivial information, but for debugging reasons useful.

3.4 Localization

A PU localization [49] for stationary FSI reads:

**Proposition 3.1.** We have for the primal error part $\rho(\hat{U}_h)(\cdot)$ the a posteriori error estimate

$$|J(\hat{U}) - J(\hat{U}_h)| \leq \eta := \sum_{i=1}^M \eta_i \leq \sum_{i=1}^M |\eta_i|$$

where $M$ is the dimension of the PU finite element space $\hat{V}_{PU}$ (composed of $Q_h^1$ functions $\chi_i$) and with the PU-DoF indicators

$$\eta_i = -A(\hat{U}_h)((\hat{z}_h^{(2)} - i_h\hat{z}_h^{(2)})\hat{\psi}_i) + \hat{F}((\hat{z}_h^{(2)} - i_h\hat{z}_h^{(2)})\hat{\psi}_i)$$

$$= -\langle \rho_f \hat{F}^{-1}\hat{\psi}_i \cdot \hat{\nabla} \hat{\psi}_i, \hat{\psi}_i \rangle_{\Omega_f} - \langle \sigma_f \hat{\nabla} \hat{\psi}_i, \hat{\psi}_i \rangle_{\Omega_f} + \langle g_f, \hat{\psi}_i \rangle_{\Gamma_N}$$

$$+ \langle \hat{F} \hat{\psi}_i \rangle_{\Omega_s} - \langle \sigma_{msh} \hat{\nabla} \hat{\psi}_i, \hat{\psi}_i \rangle_{\Omega_f} - \langle \text{div}(\hat{F}^{-1}\hat{\psi}_i), \hat{\psi}_i \rangle_{\Omega_f}$$

$$+ \langle \rho_f \hat{J}_{ff}, \hat{\psi}_i \rangle_{\Omega_f} + \langle \rho_s f_s, \hat{\psi}_i \rangle_{\Omega_s}$$

with the interpolation $i_h : \hat{X}_h^{(2)} \rightarrow \hat{X}_h$ and the weighting functions are defined as

$$\hat{\psi}_i^v := (\phi_{2h,v} - \phi_{h,v}) \chi_i, \quad \hat{\psi}_i^u := (\phi_{2h,u} - \phi_{h,u}) \chi_i, \quad \hat{\psi}_i^p := (\phi_{2h,p} - \phi_{h,p}) \chi_i.$$

3.5 Adaptive algorithm

1. Compute the primal solution $\hat{U}_h$ and the (higher-order) adjoint solution $\hat{z}_h^{(2)}$ on the present mesh $\mathcal{T}_h$.

2. Evaluate $|\eta| := |\sum_i \eta_i|$ in [4].

3. Check, if the stopping criterion is satisfied: $|J(\hat{U}) - J(\hat{U}_h)| \leq |\eta| \leq TOL$, then accept $U_h$ within the tolerance $TOL$. Otherwise, proceed to the following step.

4. Mark all elements $K_i$ for refinement that touch DoFs $i$ with indicator $\eta_i$ with $\eta_i \geq \frac{\alpha m}{M_{el}}$ (where $M_{el}$ denotes the total number of elements of the mesh $\mathcal{T}_h$ and $\alpha \approx 1$). Alternatively, pure DoF-based refinement in $i$ can be carried out.
4 Numerical tests

In this section, we consider the FSI-1 benchmark \cite{36} (see also the books \cite{11, 10} and our own former results \cite{47, 58}) and the 2D-1 benchmark \cite{50}. The drag value is taken as goal functional. As previously mentioned, this paper is accompanied with a respective open-source implementation on github\footnote{https://github.com/tommeswick/goal-oriented-fsi} based on the finite element library deal.II \cite{1, 2} and our previous fluid-structure interaction code \cite{58}, which is also available on github\footnote{https://github.com/tommeswick/fsi}.

4.1 FSI-1 benchmark

The configuration, all parameters, and reference values can be found in \cite{36}. The reference value for computing the true error was computed on a five times refined mesh and is 1.5370185576528707e+01 (see also in the provided github code). Our results from the file dwr_results.txt are:

| Dofs | True err | Est err | Est ind | Eff | Ind |
|------|----------|---------|---------|-----|-----|
| 13310 | 2.58e-01 | 1.43e-01 | 4.37e-01 | 5.54e-01 | 1.69e+00 |
| 20921 | 9.00e-02 | 4.75e-02 | 1.60e-01 | 5.28e-01 | 1.77e+00 |
| 37874 | 3.20e-02 | 1.09e-02 | 5.96e-02 | 3.40e-01 | 1.86e+00 |
| 68754 | 1.84e-02 | 4.57e-03 | 2.77e-02 | 2.48e-01 | 1.51e+00 |

Furthermore, the terminal output yields

Disx : 2.2656126465725842e-05
Disy : 8.1965770448936843e-04
P-Diff: 1.4819455817646477e+02
P-front: 1.4819455817646477e+02

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Face drag: 1.5351806985399641e+01
Face lift: 7.3933527637991259e-01

where Face drag represents the chosen goal functional. While the error reductions in the True err $J(\hat{U}) - J(\hat{U}_h)$ and the estimated error $\eta$ are reasonable, the effectivity index Eff has room for improvement. The indicator index Ind (for the definition see \cite{49}) performs quite well. The main reason for the intermediate effectivity indices might be the accuracy of the reference value. Second, we notice that only the primal error part $\rho$ (Corollary 3.2) was used. As shown in our recent studies for quasi-linear problems, the adjoint error part $\rho^*$ might play a crucial role in order to obtain nearly perfect effectivity indices for highly nonlinear problems \cite{18}. Graphical solutions of the primal solution, including the adaptively refined mesh, and the adjoint solution are displayed in Figure 1.

4.2 Adaptation to flow benchmark 2D-1

The provided code can be adapted with minimal changes to the 1996 flow around cylinder benchmark 2D-1 \cite{50}. In the *.inp file the material ids for solid must be set to 0 (flow), and the inflow and material parameters are adapted correspondingly. Of course, in this code, the displacement variables are still computed despite that they are zero everywhere, which increases the computational cost in comparison to a pure fluid flow code. Due to the zero displacements $\hat{u} = 0$, the ALE mapping is the identity, yielding $\hat{F} = \hat{I}$ and $det(\hat{F}) = 1$. Consequently, there is no mesh deformation and the Navier-Stokes equations fully remain in
Eulerian coordinates. Here, extracting information from `dwr_results.txt`, the findings for the drag value as goal functional are:

| Dofs  | True err | Est err | Est ind | Eff   | Ind   |
|-------|----------|---------|---------|-------|-------|
| 1610  | 3.51e-01 | 2.97e-01| 6.20e-01| 8.44e-01| 1.76e+00|
| 2586  | 8.80e-02 | 7.27e-02| 2.21e-01| 8.26e-01| 2.51e+00|
| 4764  | 1.89e-02 | 1.54e-02| 7.11e-02| 8.14e-01| 3.75e+00|
| 10830 | 3.23e-03 | 2.95e-03| 1.82e-02| 9.13e-01| 5.62e+00|

The pressure, drag (goal functional), and lift values are taken from the terminal output:

P-Diff: \(1.1743527755157424e-01\)
P-front: \(1.3213237901562136e-01\)
P-back: \(1.4697101464047121e-02\)

Face drag: \(5.5754969431700365e+00\)
Face lift: \(1.0717678080199560e-02\)

These values fit well with the reference values given in [50]. Moreover, we observe very stable effectivity indices, which indicate that the primal error estimator \(\rho\) (Corollary 3.2) is for incompressible Navier-Stokes a sufficient choice. Indeed, using this part only, was already suggested in early work [6, 9]. Finally, we notice that extensions to multiple goal functionals for the 2D-1 benchmark were undertaken in [17, 19].

5 Conclusions

In this work, we developed and implemented PU-DWR goal-oriented error control and spatial mesh adaptivity for stationary fluid-structure interaction. An important part is the open-source programming code published on github. As numerical example, the FSI-1 benchmark is chosen. Therein, mesh adaptivity performs as expected and also the error reductions in the true error and estimated error are good. However, the effectivity index may be improved. Extensions of this work include inter alia the implementation of the adjoint error part \(\rho^*\), local-higher order interpolations for the adjoint rather than using global-higher order finite elements, parallel iterative/multigrid linear solvers within Newton’s method, and a 3D implementation. The latter is implementation-wise not difficult with deal.II’s dimension-independent programming, but the linear solver becomes really important.

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