ACSE9 - IRP Project Plan

Author: Keer Mei

Supervisor: Prof. Matthew Piggott

June 28, 2019

1 Introduction

The goal of this project is to develop a Python framework using the Firedrake API for solving problems involving coupled partial differential equations (PDEs). Currently, Firedrake users must have some degree of familiarity with its functionality as well as technical knowledge of the finite element method. A Python framework provides clarity and ease of use for users who want to adopt the finite element method by allowing them to issue high level instructions for setting up and solving their own coupled problems.

Inevitably, many different coupled problems will share the same or similar underlying PDEs depending on the governing physical process. In this regard, the Firedrake framework can also provide developers with a flexible basis to implement their own coupled problems using the already developed foundation classes.

Included within the scope of this project will be demonstrations of how the Firedrake framework is used to solve several different coupled problems.

2 Background

Firedrake employs the Unified Form Language (UFL) which is a domain-specific language that contains built-in support for automatic differentiation and producing variational forms of partial differential equations (PDEs). This allows the user to generate high-level code that resembles closely the underlying mathematical equations.

Although the existing Firedrake API provides an intuitive interface for users to deploy the individual steps of the finite element method, it is not a trivial process. The knowledge to build a mesh, specify initial/boundary conditions, create multiple and/or mixed function spaces, specify solver types, as well as a multitude of other considerations makes using

Firedrake as a research tool for coupled PDEs relatively complicated. A framework will simplify this process by giving users PDE classes with inter-dependencies without needing to worry about how each individual functionality comply with others. An example of a similar framework has been implemented in FEniCS to solve turbulent flow models involving coupled PDEs [1].

2.1 Nomenclature

The following table provides definitions for the variables used in the equations described throughout section 2:

Symbol	Definition
u, v	velocity
p	pressure
ρ	density
μ	viscosity
c_1, c_2, c_3	concentrations of chemical components 1, 2, and 3
D	diffusivity coefficient
f_1, f_2, f_3	source terms of chemical components 1, 2, and 3
K	reaction coefficient
U	average velocity
\tilde{u}	mean velocity component
u'	fluctuating velocity component
z	displacement of water surface
ω	Coriolis parameter
$ au_u, au_v$	sea bed friction stresses
H	total water depth
A	horizontal eddy viscosity

Table 1: Table of nomenclature.

2.2 Example problem: flow coupled reactions

One example of a coupled problem is the two-dimensional incompressible channel flow past a cylinder with a first-order chemical reaction [2]. In this problem, we must first solve the 2-D Navier-Stokes (N-S) Equation, given by:

$$\frac{\partial u}{\partial t} + u \cdot \nabla u = \frac{1}{\rho} \nabla p + \mu \nabla^2 u, \tag{1}$$

$$\nabla \cdot u = 0. \tag{2}$$

The N-S equation will have to be solved on a finite mesh discretizing the domain, with a specified mesh size, e.g. as shown in figure 1:

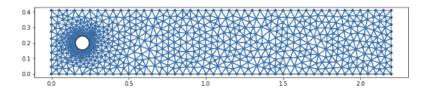


Figure 1: Channel and cylinder mesh.

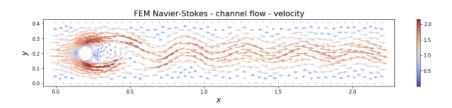


Figure 2: Velocity profile of flow past cylinder.

Next, using the velocity (u) obtained from the solution to the N-S Equation, we can solve for the concentrations of the different chemical components using their mass balance:

$$\frac{\partial c_1}{\partial t} + u \cdot \nabla c_1 - \nabla \cdot (D\nabla c_1) = f_1 - Kc_1c_2, \tag{3}$$

$$\frac{\partial c_2}{\partial t} + u \cdot \nabla c_2 - \nabla \cdot (D \nabla c_2) = f_2 - K c_1 c_2, \tag{4}$$

$$\frac{\partial c_2}{\partial t} + u \cdot \nabla c_2 - \nabla \cdot (D \nabla c_2) = f_2 - K c_1 c_2,
\frac{\partial c_3}{\partial t} + u \cdot \nabla c_3 - \nabla \cdot (D \nabla c_3) = f_3 + K c_1 c_2 - K c_3.$$
(4)

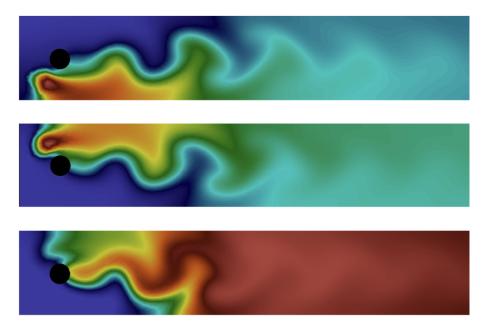


Figure 3: Reactants concentration profiles [2].

2.3 Example problem: turbulent flow

Another example of a coupled PDE problem is the modelling of turbulent flow. In research, there exists a number of different turbulence models such as the Reynolds-averaged Navier-Stokes (RANS), Large Eddy Simulations (LES), and Direct Numerical Simulations (DNS) [5]. Each of these models provide their own advantages and disadvantages and may not be suitable for all scenarios. In the RANS model for example [1], a modified version of the N-S equation must be solved:

$$\frac{\partial U}{\partial t} + U \cdot \nabla U = \frac{1}{\rho} \nabla p + \mu \nabla^2 U, \tag{6}$$

$$\nabla \cdot U = 0, \tag{7}$$

$$U = \tilde{u} + u'. \tag{8}$$

where U is an average velocity separated into mean (\tilde{u}) and fluctuating (u') components. The difficulty lies in defining the transport models (comprising of coupled PDEs) that describe the interaction between mean and fluctuating components.

2.4 Example problem: hydrodynamic coupled radioactivity transport

One application of a hydrodynamics model is in the quantification of radioactivity dispersion in the marine environment. The solution to the hydrodynamic equations provide the magnitude of the water currents at each point of the model domain, which are used to solve for the transport equation [7]. The following is an example of 2D depth-averaged hydrodynamic equations:

$$\frac{\partial z}{\partial t} + \frac{\partial}{\partial x}(Hu) + \frac{\partial}{\partial y}(Hv) = 0, \tag{9}$$

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} + v \frac{\partial u}{\partial y} + g \frac{\partial z}{\partial x} - \omega v + \frac{\tau_u}{\rho H} = A(\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2}), \tag{10}$$

$$\frac{\partial v}{\partial t} + u \frac{\partial v}{\partial x} + v \frac{\partial v}{\partial y} + g \frac{\partial z}{\partial x} - \omega u + \frac{\tau_v}{\rho H} = A(\frac{\partial^2 v}{\partial x^2} + \frac{\partial^2 v}{\partial y^2}). \tag{11}$$

Once the water currents have been resolved on the domain, the next step is to solve the 2D transport equation describing the concentration of radioactivity [7]:

$$\frac{\partial c}{\partial t} + \frac{\partial (uc)}{\partial x} + \frac{\partial (vc)}{\partial y} = D(\frac{\partial^2 c}{\partial x^2} + (\frac{\partial^2 c}{\partial y^2})$$
(12)

2.5 Existing challenges

As mentioned, the existing Firedrake API equipped with UFL is well suited for converting mathematical expressions into code. However, solving PDEs using Firedrake is still a complicated process that involves multiple steps, which creates opportunities for programming errors. The following excerpt demonstrates the basic necessities for solving the N-S reactions coupled problem described in section 2.2:

```
import firedrake as fd
# physical constants
nu = fd.Constant(0.001)
\# time step
dt = 0.001
# define a firedrake constant equal to dt so that variation forms
# not regenerated if we change the time step
k = fd.Constant(dt)
# diffusion rate
eps = fd.Constant(0.01)
# reaction rate
K = fd.Constant(10.0)
# create a mesh
mesh = fd. Mesh("flow_past_cylinder.msh")
# create Function spaces
V = fd. VectorFunctionSpace (mesh, "CG", 2)
Q = fd. FunctionSpace (mesh, "CG", 1)
u = fd. TrialFunction(V)
v = fd. TestFunction(V)
p = fd.TrialFunction(Q)
q = fd \cdot TestFunction(Q)
# creating coupled variables and equations from the reaction formulas
P1 = fd.FiniteElement("CG", 'triangle', 1)
element = fd. MixedElement ([P1, P1, P1])
H = fd. FunctionSpace (mesh, element)
v_{-1}, v_{-2}, v_{-3} = fd. TestFunctions (H)
c = fd \cdot Function(H)
c_n = fd.Function(H)
```

```
n = fd.FacetNormal(mesh)
f = fd.Constant((0.0, 0.0))
u_{mid} = 0.5*(u_{n} + u)
x, y = fd.SpatialCoordinate(mesh)
# create source terms
f_{-1} = fd. conditional(pow(x-0.1, 2)+pow(y-0.1,2)<0.05*0.05, 0.1, 0)
f_{-2} = fd. conditional(pow(x-0.1, 2)+pow(y-0.3,2)<0.05*0.05, 0.1, 0)
f_3 = fd.Constant(0.0)
# Define boundary conditions
bcu = [fd.DirichletBC(V, fd.Constant((0,0)), (1, 4)),
# Define variational forms
F1 = fd.dot((u - u_n)/k, v) * fd.dx \setminus
    + \text{ fd.dot}(\text{fd.dot}(u_n, \text{ fd.nabla\_grad}(u_n)), \text{ } v) * \text{ fd.dx} \setminus
    + fd.inner(sigma(u_mid, p_n), fd.sym(fd.nabla_grad(v))) * fd.dx \
    + ...
a1, L1 = fd.system(F1)
a2 = fd.dot(fd.nabla_grad(p), fd.nabla_grad(q)) * fd.dx
L2 = fd.dot(fd.nabla_grad(p_n), fd.nabla_grad(q)) * fd.dx 
    - (1/k) * fd.div(u_{-})*q*fd.dx
a3 = fd.dot(u, v) * fd.dx
L3 = fd.dot(u_{-}, v) * fd.dx
     - k * fd.dot(fd.nabla_grad(p_- - p_n), v) * fd.dx
F4 = ((c_1 - c_n 1) / k) * v_1 * fd.dx 
    + ((c_2 - c_n 2) / k) * v_2 * fd. dx
    + ((c_3 - c_n 3) / k) * v_3 * fd.dx
F4 += ...
# Define problems classes
prob1 = fd. Linear Variational Problem (a1, L1, ...)
prob2 = fd. Linear Variational Problem (a2, L2, ...)
prob3 = fd. Linear Variational Problem (a3, L3, ...)
prob4 = fd. Nonlinear Variational Problem (F4, c, ...)
# Define solvers classes
solve1 = fd.LinearVariationalSolver(prob1, solver_parameters = \{...\})
```

```
solve2 = fd.LinearVariationalSolver(prob2, solver_parameters = {...})
solve3 = fd.LinearVariationalSolver(prob3, solver_parameters = {...})
solve4 = fd.NonlinearVariationalSolver(prob4, solver_parameters = {...})

# Time loop
t = 0.0
t_end = 5.
while t < t_end :
    t += dt
    solve1.solve()</pre>
```

From the excerpt above, several difficulties with using the current Firedrake API become obvious:

- 1. First, Firedrake users must understand the basics of how to set up function spaces, boundary conditions, conditional source terms, etc...
- 2. Next, although it is not mentioned, the variational forms written in this example uses a pressure projection method, AKA Chorin projection, to solve for velocity. This is knowledge that the users must be aware of before even attempting to use Firedrake.
- 3. What are the appropriate solver parameters that an user should specify? How does an user know if the time step size or the mesh grid size is appropriate for their problem?
- 4. Finally, the current implementation is simply too long and onerous to solve for larger, more coupled problems. Setting up a problem with hundreds of PDEs is a significant amount of work, but many of the steps will be in common.

The end goal is to use the Firedrake framework to simplify the problem and address the challenges above. Similar to the FEniCS framework [1], the user should only be required to specify high level programming instructions in a few short lines of code. Below is an example of how the FEniCS framework solves the N-S equation:

2.6 Adaptability of using a Framework

Improving code readability and reducing redundancy are just some of the benefits to be gained in adopting a programmable framework. Another key advantage of adopting a programmable framework for Firedrake is to create generality, allowing Firedrake to be adapted quickly to a variety of coupled problems.

For example, the FEniCS framework provides users with generalized Python classes that are able to adapt any coupled problem. The three main object classes in the FEniCS framework are: problems, solvers, and schemes [8]. The problem class is responsible for describing the domain and model inputs to the coupled problem. The solver class defines the complete set of PDEs and refers to individual schemes to set up and solve parts of the overall system of PDEs [1].

With the three generalized Python objects, solving for different systems of PDEs become trivial. Any type of coupled problem can be represented by a combination of problem, solver, and scheme.

One example demonstrated in [1] is how to couple the N-S equations with a turbulent model using the three objects:

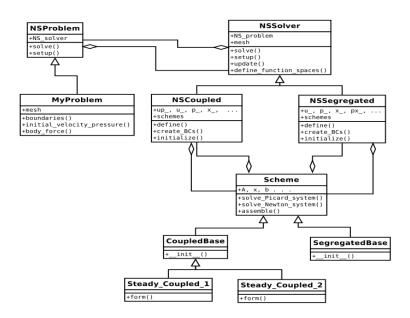


Figure 4: N-S problem dependencies [1].

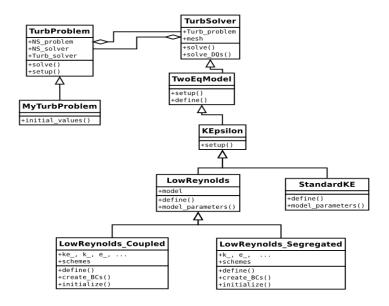


Figure 5: Turbulent model dependencies [1].

The objects used to represent the N-S problem in figure 4 provide a basis to define the turbulent models in figure 5. Solving the turbulent problem involves solving the underlying N-S problem as well. By swapping the turbulent model for a hydrodynamics model, or a reactions model, the problem can be easily redefined into a different coupled problem using the framework objects.

3 Goals and Milestones

The following are high level targets that have been set by myself based on discussions between myself and my supervisor. I am confident that meeting these deadlines will ensure a smooth progression towards the final deliverable. *Disclaimer*: this is an initial blueprint and certain milestones may not be completed in the planned order or to the same specification.

3.1 June 28, 2019 Milestone

The goal set for this initial period was to accomplish the following:

- 1. Familiarize myself with writing variational forms of PDEs by reviewing lecture notes. Gain a better understanding of the Firedrake API by creating my own python notebook and attempt to duplicate tutorial problems from Lecture and the Firedrake website. (Done)
- 2. Conduct literature review to collect a variety of coupled PDE problems (RANs turbulence, radionuclide spread in aquatic ecosystems [6], oil spill models [4], reaction

models, etc..). Determine potential demonstration problems to be included and solved within the Firedrake framework. (Ongoing)

3.2 July 19, 2019 Milestone

By the end of June, I should have acquainted myself enough with the Firedrake API to begin programming a framework. The goal during this period will be to focus on producing the outline of the Firedrake framework.

- 1. Produce a simple N-S problem and solver class that resembles the examples developed within the FEniCS framework [1]. Once this is accomplished, expand the dependencies and build another class that is coupled with the N-S problem and solver (e.g. either a RANS class or a reactions class). (Not started)
- 2. Finalize literature review and determine the direction of the demonstration problems to be solved either a complicated coupled problem that spans multiple disciplines or a variety of simpler coupled problems similar to the FEniCS [2] and Firedrake [3] tutorials. (Not started)

3.3 August 9, 2019 Milestone

By the start of August, the goal is to have several different classes created within the framework. These classes should capture the complete set of coupled PDEs required for the demonstration problems. These classes will allow users to freely couple problems in the available domains and should be a basic foundation for developers to continue building their own coupled problems. The outline of the final framework should be evident at this point with the following features:

- 1. A way to verify the convergence of the solvers would be beneficial. Although it has not yet been discussed, a Manufactured Solution for some simple examples may be appropriate here to test the framework. (Not started)
- 2. To be determined: whether any extra functionality or nice to have features should be included. (Not started)

3.4 August 30, 2019 Milestone

This period ends with the final submission deadline. The goal during this period of the project will be to update all of the code documentation and complete the final report.

3.5 September 7, 2019 Milestone

This final milestone period will be used to collect the major accomplishments that will be highlighted in the final presentation.

4 Measures of success

The following is a gradient of evaluation for this project, based on my own expectations of what should be accomplished by the Firedrake framework:

- Satisfactory: The programmable framework is functional, and a couple of different problems have been developed. Demonstration examples are available, documentation is complete, continuous integration is implemented. Manufactured tests are available for convergence analysis on demonstration examples.
- Beyond Expectations: All metrics of Satisfactory have been met with the addition of: an analysis of different solver parameters and mesh sizing. Built-in functionality for uniform adaptable mesh sizing and time steps would be a bonus. A generalized MMS class to adapt to any type of coupled problem would be a bonus.
- Exceptional: All metrics of Beyond Expections have been met with the addition of: non-uniform adaptable mesh and the ability to locally refine mesh on a problem. Adapting the framework onto multiple domains (e.g. solid-liquid interactions, gasliquid interactions). Parallelisation of solvers. (Note: these are ideal "nice to have" functionality and I have not determined their feasibility with the existing Firedrake API)

References

- [1] Mikael Mortensen, Hans Petter Langtangen, Garth N. Wells. A FEniCS-Based Programming Framework for Modelling Turbulent Flow by the Reynolds-Averaged Navier-Stokes Equations Advances in Water Resources 2011;34(9):1082-1101. Available from: https://www.sciencedirect.com/science/article/pii/S030917081100039X?via%3Dihub. [Accessed 2019]
- [2] Hans Petter Langtangen and Anders Logg. Solving PDEs in Python. Springer; 2017. Available from: https://fenicsproject.org/tutorial/. [Accessed 2019]
- [3] Firedrake. Introductory Jupyter notebooks. Available from: https://www.firedrakeproject.org/notebooks.html. [Accessed 2019]
- [4] DHI. MIKE21 MIKE3 FlowModelFM.OilSpillModule.Short Description. Agem Alle 5, DK-2970 Horsholm, Den-2017. Available from: https://www.mikepoweredbydhi.com/mark.

- /media/shared%20content/mike%20by%20dhi/flyers%20and%20pdf/product-documentation/short%20descriptions/mike213_os_fm_short_description.pdf. [Accessed 2019]
- [5] Walter Frei. Which Turbulence Model Should I Choose for My CFD Application. Available from: https://uk.comsol.com/blogs/which-turbulence-model-should-choose-cfd-application/. [Accessed 2019]
- [6] Anders Christian Erichsen, Flemming Møhlenberg, Rikke Margrethe Closter, Johannes Sandberg. Models for transport and fate of carbon, nutrients and radionuclides in the aquatic ecosystem at Öregrundsgrepen. Svensk Kärnbränslehantering AB: R-10-10, 2010.
- [7] Raul Perianez. Modelling the Dispersion of Radionuclides in the Marine Environment. Springer.
- [8] Mikael Mortensen. CBC.PDESys. Available from: https://launchpad.net/cbcpdesys. [Accessed 2019]