

Mandatory assignment 1

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1 Exercise 1

1.1 Exercise 1a)

See hand written notes

1.2 Exercise 1b)-1d)

Since the equation is independent of y , we can solve it using a 1D domain. To solve the equation, I implemented a 1D FEniCS solver using both the Galerkin and streamwise upwind Petrov-Galerkin method using first order Lagrange elements.

For the Petrov-Galerkin method, I used the same variational formulation as for the standard Galerkin-method, but with test functions given by

$$v = v_g + \beta h \frac{\partial v}{\partial x}, \quad (1)$$

where h was the mesh resolution, $\beta = 0.5$ was the stabilisation coefficient and v_g was the standard first order Lagrangian test functions.

To estimate the parameters for the error estimate, I used the Nelder-Mead algorithm [3, 1] from SciPy [5] to minimise the loss function

$$\sum_{h \in \mathcal{H}} f(\|u_h - u\|, Ch^p)^2, \quad (2)$$

where $\mathcal{H} = \{1/8, 1/16, 1/32, 1/64\}$ is the set of all mesh resolutions and u and u_h are the true and estimated solution to the PDE, respectively. $C = C_\alpha$ and $p = \alpha$ for $\|\cdot\| = \|\cdot\|_1$ and $C = C_\beta$ and $p = \beta$ for $\|\cdot\| = \|\cdot\|_0$. f was a function used to penalise under-estimates of the error more severely than over-estimates, and were given by

$$f(a, b) = \begin{cases} a - b & a < b \\ \rho(a - b) & \text{otherwise} \end{cases}, \quad (3)$$

ρ was tuned manually to ensure that the error estimate was an upper bound for the true error, and was set to 10^8 . The initial guess for the Nelder-Mead

Table 1: Estimated parameters

| Scheme | μ | C_α | α | C_β | β |
|-----------------|-------|------------|----------|-----------|---------|
| Galerkin | 0.1 | 6.13 | 0.99 | 1.48 | 1.98 |
| | 0.3 | 1.88 | 1.09 | 0.30 | 2.00 |
| | 1.0 | 1.11 | 1.31 | 0.09 | 2.00 |
| Petrov-Galerkin | 0.1 | 5.94 | 0.87 | 1.26 | 1.12 |
| | 0.3 | 1.82 | 0.99 | 0.88 | 1.24 |
| | 1.0 | 0.90 | 1.24 | 0.32 | 1.48 |

algorithm was $C = 5$ and $p = 1$. The various estimated parameters are shown in Table 1.

From the error estimates in Table 1, we see that the traditional Galerkin approach has better convergence properties than the Petrov-Galerkin method. However, from the convergence plots in Figure 1, we see that the Petrov-Galerkin method shows promise for rough meshes.

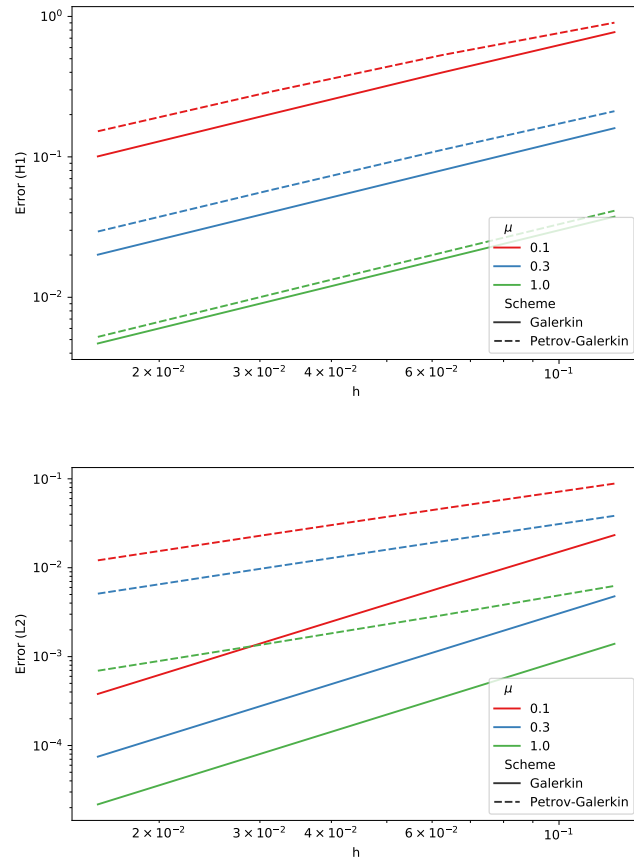


Figure 1: Error norm as a function of mesh resolution. Top: H1 norm, bottom: L2 norm

2 Exercise 2

2.1 Exercise 2a)

For both the explicit and the implicit scheme, based on the incremental pressure correction method described in [2]. The solver was made for benchmark task 2.2 b) in [4]. The code is available on the GitHub repository for the assignment.

To obtain the initial conditions for the solver, I solved Stokes equation with the same parameters as for the Navier-Stokes equation. By default, it uses the standard P2-P1 Taylor-Hood elements for the velocity-field and pressure, respectively, but with an option to select first order elements for both velocity and pressure. By using P1-P1 elements, we get NaN-values in the pressure field, illustrating the superiority of Taylor-Hood elements compared to all linear elements.

2.2 Exercise 2b)

I did not find a good way to solve this problem. However, I start with some arguments towards showing that the stability conditions for the explicit scheme is $C\Delta t\Delta x^{-2}$. We have the equation

$$\dot{u} = -u \cdot \nabla u - \nabla p + \nabla \cdot (\mu \nabla u) + f. \quad (4)$$

Consider a velocity field where ∇p is constant in time, then, we can combine f and ∇p into a single variable, $\omega = f - \nabla p$ and obtain the system

$$\dot{u} = -u \cdot \nabla u + \nabla \cdot (\mu \nabla u) + \omega. \quad (5)$$

Since we know that the stability requirements for

$$\dot{u} = \nabla \cdot (\mu \nabla u) + \omega. \quad (6)$$

solved with an explicit solver is that $\Delta t \leq C\Delta x^2$, and that $u \cdot \nabla u$ is a non-linear term that does not have stabilising properties, we can postulate that the same stability criterion is a best-case-scenario for solving the Navier-Stokes equation.

2.3 Exercise 2c)

To compute the drag coefficient, we first need to compute the drag force. To accomplish this, I used FEniCS to solve the integral

$$\int_{\Gamma_c} [(pI - \epsilon(u)) \cdot n]_x ds, \quad (7)$$

where Γ_c is the cylinder boundary, p is the pressure, $\epsilon(u)$ is the the strain tensor, n is the outwards facing unit normal of the cylinder and $[\cdot]_x$ denotes the x-component of a vector. Moreover, to improve the simulation accuracy, I start by running the simulation for 1.5 s with a low-resolution mesh ($h \approx 1/64$). The

result of the final time-step was subsequently used as the initial condition for a simulation with a higher resolution mesh ($h \approx 1/256$) for which the simulation was run for 0.5 s. The maximal drag coefficient was estimated to be 3.64 and the maximal pressure difference was 3.05 Pa.

References

- [1] Fuchang Gao and Lixing Han. “Implementing the Nelder-Mead simplex algorithm with adaptive parameters”. In: *Computational Optimization and Applications* 51.1 (2012), pp. 259–277.
- [2] Hans Petter Langtangen, Kent-Andre Mardal, and Ragnar Winther. “Numerical methods for incompressible viscous flow”. In: *Advances in water Resources* 25.8-12 (2002), pp. 1125–1146.
- [3] John A Nelder and Roger Mead. “A simplex method for function minimization”. In: *The computer journal* 7.4 (1965), pp. 308–313.
- [4] Michael Schäfer et al. “Benchmark computations of laminar flow around a cylinder”. In: *Flow simulation with high-performance computers II*. Springer, 1996, pp. 547–566.
- [5] Pauli Virtanen et al. “SciPy 1.0: fundamental algorithms for scientific computing in Python”. In: *Nature methods* 17.3 (2020), pp. 261–272.