

# MEK 4250 Elementmethod

## Mandatory Assignment

Andreas Slyngstad

7. mai 2016

### 1 Exercise 1

In these set of exercises we will study the Stokes problem defined as

$$\begin{aligned} -\Delta u + \nabla p &= f \quad \text{in } \Omega \\ \nabla \cdot v &= 0 \quad \text{in } \partial\Omega \\ u &= g \quad \text{in } \partial\Omega_N \\ \frac{\partial u}{\partial x} - pn &= h \quad \text{in } \partial\Omega_N \end{aligned}$$

First off we will define the weak formulation for the stokes problem. Let  $u \in H_{D,g}^1$  and  $p \in L^2$ . Then the stokes problem can be defined as

$$\begin{aligned} a(u, v) + b(p, v) &= f(v) \quad v \in H_{D,0}^1 \\ b(q, u) &= 0 \quad q \in L^2 \end{aligned}$$

Where a and b defines the bilinear form, and f defines the linear form as

$$\begin{aligned} a(u, v) &= \int \nabla u : \nabla v \, dx \\ b(p, v) &= \int p \nabla \cdot v \, dx \\ f(v) &= \int f v \, dx + \int_{\Omega_N} h v \, ds \end{aligned}$$

Further we will define to properties which will be usefull for solving the exercises

#### **Cauchy-Schwartz inequality**

Let V be a inner product space, then

$$|\langle u, v \rangle| \leq \|u\| \cdot \|v\| \quad \forall u, v \in V$$

**Poincare's Inequality** Let  $v \in H_0^1(\Omega)$

$$\|v\|_{L^2(\Omega)} \leq C \|v\|_{H^1(\Omega)}$$

### Exercise 7.1

In this section we are to prove the conditions (7.14-7.16) from the course lecture notes. Starting off with the first **Condition 7.14**

$$a(u_h, v_h) \leq C_1 \|u_h\|_{V_h} \|v_h\|_{V_h} \quad \forall u_h, v_h \in V_h$$

As for in all of these conditions we will assume that  $V_h \in H_0^1$ , and for later conditions that  $Q_h \in L^2$ . First off we write out the term  $a(u_h, v_h)$ , and we observe we can use the Cauchy-Schwartz inequality since  $V$  is an inner product space.

$$\begin{aligned} a(u_h, v_h) &= \int_{\Omega} \nabla u_h : \nabla v_h \, dx = \langle \nabla u_h, \nabla v_h \rangle \\ \langle \nabla u_h, \nabla v_h \rangle &\leq |\langle \nabla u_h, \nabla v_h \rangle|_0 \leq \|\nabla u_h\| \cdot \|\nabla v_h\| \end{aligned}$$

Now since we have defined that  $u_h, v_h \in V_h \in H_0^1$  we can use the Poincare inequality. First we observe that

$$\begin{aligned} \|\nabla u_h\|_{L^2}^2 &\leq \|u_h\|_{H^1}^2 = \|u_h\|_{L^2}^2 + \|\nabla u_h\|_{L^2}^2 \leq C_1 \|u_h\|_{H^1}^2 + \|\nabla u_h\|_{L^2}^2 = C_1 \|\nabla u_h\|_{L^2}^2 + \|\nabla u_h\|_{L^2}^2 \\ &= (C_1 + 1) \|\nabla u_h\|_{L^2}^2 \leq D \|u_h\|_{H^1}^2 \end{aligned}$$

#### Condition 7.14

$$b(u_h, q_h) \leq C_2 \|u_h\|_{V_h} \|q_h\|_{Q_h} \quad V_h \in H_0^1, \quad Q_h \in L^2,$$

By direct insertion we get

$$b(u_h, q_h) = \int p \nabla \cdot v \, dx = \langle p, \nabla \cdot u \rangle \leq |\langle p, \nabla \cdot u \rangle|_0$$

Using the Cauchy-Schwartz inequality we can show that

$$|\langle q, \nabla \cdot u \rangle| \leq \|q\|_0 \cdot \|\nabla \cdot u\|_0$$

Hence, it holds to show that

$$\begin{aligned} \|q\|_0 \cdot \|\nabla \cdot u\|_0 &\leq C_2 \|u_h\|_1 \|q_h\|_0 \\ \|\nabla \cdot u\|_0 &\leq C_2 \|u_h\|_1 \end{aligned}$$

We choose square the left side of the inequality and expand the norm, in hope of finding a term to determine the upper bound of  $b$ . Applying the Poincare inequality on line 2, we can determine that the bound must be determined by some constant  $C_2$

$$\begin{aligned} \|\nabla \cdot u\|_0^2 &\leq \|u\|_1^2 = \|u\|_0^2 + \|\nabla u\|_0^2 \\ \|u\|_0^2 + \|\nabla u\|_0^2 &\leq C_2^2 \|u\|_1^2 + \|\nabla u\|_0^2 = (C_2^2 + 1) \|\nabla u\|_0^2 = C_2^2 \|\nabla u\|_0^2 \end{aligned}$$

Hence, to show the implied boundedness of  $b$ , it holds to show  $\|\nabla \cdot u\|_0^2 \leq (C_2^2 + 1) \|\nabla u\|_0^2$ . For simplicity we write out the terms for the  $R^2$  case, but the same proof can be showed for the general case  $R^n$ .

$$\begin{aligned} \|\nabla \cdot u\|_0^2 &= \int \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 v}{\partial y^2} \right)^2 dx \\ \|\nabla u\|_0^2 &= \int \left( \frac{\partial^2 u}{\partial x^2} \right)^2 + \left( \frac{\partial^2 u}{\partial y^2} \right)^2 + \left( \frac{\partial^2 v}{\partial x^2} \right)^2 + \left( \frac{\partial^2 v}{\partial y^2} \right)^2 dx \end{aligned}$$

Remembering that since we are in a normed vector space  $V$ , the triangle inequality holds.

$$\|x + y\| \leq \|x\| + \|y\| \quad \forall x, y \in V$$

Rearrangeing the terms we observe that

$$\begin{aligned} \|\nabla \cdot u\|_0^2 &= \int \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 v}{\partial y^2} \right)^2 dx \\ \|\nabla u\|_0^2 &= \int \left( \frac{\partial^2 u}{\partial x^2} \right)^2 + \left( \frac{\partial^2 v}{\partial y^2} \right)^2 dx + \int \left( \frac{\partial^2 v}{\partial x^2} \right)^2 + \left( \frac{\partial^2 u}{\partial y^2} \right)^2 dx \\ \sqrt{\int \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 v}{\partial y^2} \right)^2 dx} &\leq \sqrt{\int \left( \frac{\partial^2 u}{\partial x^2} \right)^2 dx} + \sqrt{\int \left( \frac{\partial^2 v}{\partial y^2} \right)^2 dx} \end{aligned}$$

Hence  $\|\nabla \cdot u\|_0^2 \leq C_2^2 \|\nabla u\|_0^2$ , and the inequality holds.

**Condition 7.16, Coersivity of a**

$$a(u_h, u_h) \geq C_3 \|u_h\|_{V_h}^2 \quad \forall u_h \in V_h$$

Expanding the  $H^1$  norm of  $u_h$ , and applying the Cauchy-Schwartz inequality we find the relations

$$\begin{aligned} \|u_h\|_1^2 &= \|u_h\|_0^2 + \|\nabla u_h\|_0^2 \leq C \|u_h\|_1^2 + \|\nabla u_h\|_0^2 \\ &= C \|\nabla u_h\|_0^2 + \|\nabla u_h\|_0^2 = (C + 1) \|\nabla u_h\|_0^2 = C_3 \|\nabla u_h\|_0^2 \end{aligned}$$

Dividing the constant on both sides of the inequality and let  $C_3 = 1/C_3$  we get our final result which proofs the coersivity of a

$$C_3 \|u_h\|_1^2 \leq \|\nabla u_h\|_0^2 = \int \nabla u_h : \nabla u_h dx = a(u_h, u_h)$$

### 1.1 Exercise 7.6

Looking on the poiseuille flow, we are to investigate if the anticipated convergence rates applies for the problem

$$\begin{aligned} -\Delta u - \nabla p &= f \text{ in } \Omega \\ \nabla \cdot v &= 0 \text{ in } \partial\Omega \end{aligned}$$

We will make use of the "manufactured solution" approach defining the analytical solution in  $\Omega$  and finding the appropriate source function as

$$\begin{aligned} u &= (\sin(\pi y), \cos(\pi x)); \quad p = \sin(2\pi x) \\ f &= (\pi^2 \sin(\pi y) - 2\pi \cos(2\pi x), \pi^2 \cos(\pi x)) \end{aligned}$$

We will this exercise examine the error estimate

$$\|u - u_h\|_1 + \|p - p_h\|_0 \leq Ch^k \|u\|_{k+1} + Dh^{l+1} \|p\|_{l+1}$$

Where  $k, l$  denotes the polynomial degree of the velocity and pressure. The error estimate will be examined using  $(P_4 - P_3)$ ,  $(P_4 - P_2)$ ,  $(P_3 - P_2)$  and  $(P_3 - P_1)$  elements for velocity and pressure accordingly. The problem will be solved on a UnitSquareMesh(N, N), for  $N = [8, 16, 32, 64]$ . Order of convergence is calculated for each choice of elements, between to neighbouring N values defined as

$$r = \frac{\log\left(\frac{E[i+1]}{E[i]}\right)}{\log\left(\frac{h[i+1]}{h[i]}\right)}$$

Where E is a list of error norms and h is a list of the facet length in the mesh defined as  $\frac{1}{N}$

Tabell 1: Convergence rate velocity,  $E = \|u - u_h\|_1$

N	Conv 4 to 8	Conv 8 to 16	Conv 16 to 32	Conv 32 to 64
4-3	4.53315	4.2951	4.11514	4.03954
4-2	2.6871	2.88045	2.96138	2.98766
3-2	2.57912	2.84778	2.9518	2.98466
3-1	2.17924	2.08939	2.04646	2.02452

Tabell 2: Convergence rate velocity,  $E = \|p - p_h\|_0$

N	Conv 4 to 8	Conv 8 to 16	Conv 16 to 32	Conv 32 to 64
4-3	4.22147	4.08419	4.02915	4.01063
4-2	2.71627	2.89978	2.97342	2.99382
3-2	2.71694	2.90114	2.97394	2.99397
3-1	2.35349	2.19402	2.09554	2.04362

To describe these results we have to take a closer look on the error estimate. Let us choose the dataset from the  $(P_4 - P_3)$  computation. For  $k = 4$  and  $p = 3$  we get.

$$\|u - u_h\|_1 + \|p - p_h\|_0 \leq Ch^4 \|u\|_5 + Dh^4 \|p\|_4$$

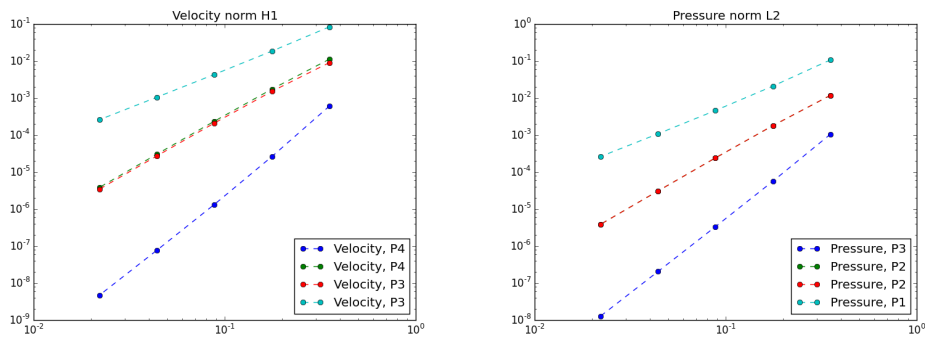
Observing that  $C, D, \|u\|_5$  and  $\|p\|_4$  are some constants and will not effect the trend of the error estimate, we see that the error is limited to  $h^4$ , both for the velocity and pressure norm. From our result this seems reasonable as we can see the order of convergence approach 4.

What about if we choose some elements which differ in degree by 2? Let's choose the  $(P_4 - P_2)$  computation and see what is going on. Our error estimate then yields

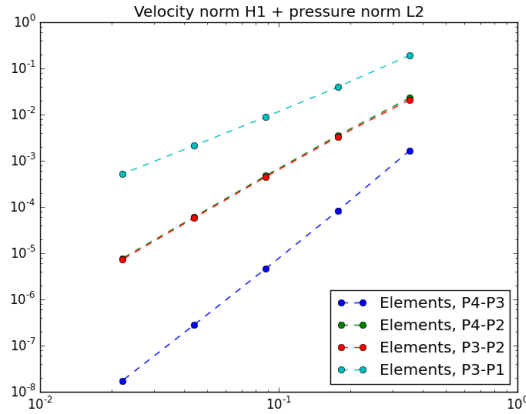
$$\|u - u_h\|_1 + \|p - p_h\|_0 \leq Ch^4\|u\|_5 + Dh^3\|p\|_3$$

Now as the velocity norm on the right side of the inequality is dominated by the  $h$  term, we observe that this term will go faster to zero than the pressure norm. Hence our calculations should be dominated by the pressure norm. From this logic we would expect the order of convergence to be limited by 3. As we can see from the numerical results, this seems legit as we can see that the rate of convergence approaches 3 as the mesh gets finer.

loglogplot



loglogplot



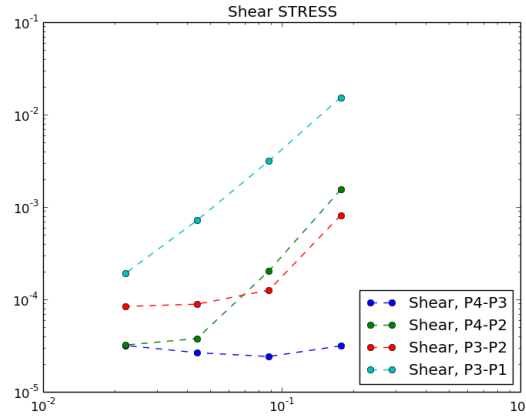
## 1.2 Exercise 7.7

In this exercise we where to calculate the order of convergence for the shear stress in the same domain presented en exercise 7.6. Let  $P$  denote the stress tensor. The wall stress on a surface in the domain is given by  $P \cdot \mathbf{n}$ , where  $\mathbf{n}$  is the normal vector pointing perpendicular to the surface out of the domain. Hence we have the following relations

$$\begin{aligned} \mathbf{P}_n &= P \cdot \mathbf{n} && \text{Shear stress} \\ P_{nn} &= \mathbf{n} \cdot \mathbf{P}_n && \text{Normal stress component} \\ P_{nt} &= |\mathbf{P}_n \times \mathbf{n}| && \text{Tangential stress component} \end{aligned}$$

In this exercise we are interested in the wall shear stress, hence we will for this time focus on the tangential stress.

loglogplot



## Exersize 2, Linear Elasticity

In this exercise we are to take a closer look on linear elasticity, and familiarize ourselves with the numerical artifact locking. We are presented with the following problem

$$\begin{aligned} -\mu \Delta \mathbf{u} - \lambda \nabla \nabla \mathbf{u} &= \mathbf{f} \text{ in } \Omega \\ \mathbf{u} &= u_e \text{ on } \partial\Omega \\ u_e &= \left( \frac{\partial \phi}{\partial y}, -\frac{\partial \phi}{\partial x} \right) \quad \phi = \sin(\pi xy) \end{aligned}$$

First and foremost since we are making a "manufactured solution", we need to determine the source term  $\mathbf{f}$ . The equation to solve for  $\mathbf{f}$  is  $\mathbf{f} = \mu \Delta \mathbf{u}_e$ , due to by construction  $\nabla \cdot \mathbf{u}_e = 0$  Hand calculations and verification with sympy gives us the following result.

$$\begin{aligned} \mathbf{f} &= \mu (\pi^3 x^3 \cos(\pi xy) - \pi^2 y (2 \sin(\pi xy) + \pi xy \cos(\pi xy))) \mathbf{i} + \\ &\quad \mu (\pi^3 y^3 \cos(\pi xy) + \pi^2 x (2 \sin(\pi xy) + \pi xy \cos(\pi xy))) \mathbf{j} \end{aligned}$$

Now how do we assess the problem? Why not try stright forward Galerkin method on the problem, since we have had many good results with this approach. Integration by parts gives us the following variational form and numerical results.

$$\mu \int \Delta uv \, dx - \lambda \int \nabla \nabla \cdot uv \, dx = \int f v \, dx$$

$$\mu \langle \nabla u_h, \nabla v_h \rangle + \lambda \langle \nabla \cdot u_h, \nabla \cdot v_h \rangle = \langle f, v \rangle$$

Now for our numerical calculations, the problem will be solved on a UnitSquareMesh(N,N) for  $N = [8, 16, 32, 64]$ , for choices of  $\lambda = [1, 100, 10000]$ . We choice second order polynomials for the velocity.

Tabell 3: L2 norm velocity

$\lambda / N$	8	16	32	64
1	0.000662282	4.40431e-05	2.81081e-06	1.76953e-07
10	0.00290596	0.000209755	1.38052e-05	8.78212e-07
100	0.0142522	0.00147779	0.000115406	7.83639e-06
1000	0.0269102	0.00513695	0.000689253	6.3586e-05
10000	0.029816	0.00716993	0.00157688	0.000272168

Tabell 4: Convergence rate velocity

$\lambda / N$	Conv 8 to 16	Conv 16 to 32	Conv 32 to 64
1	3.91046	3.96985	3.98955
10	3.79224	3.92542	3.9745
100	3.26967	3.67864	3.88039
1000	2.38917	2.89781	3.43825
10000	2.05606	2.18488	2.53451

As we can see, the first run with  $\lambda = 1$  gives us reasonable results. The  $L_2$  norm decreases nicely for finer mesh resolution, and the convergence rate approaches 3 which is reasonable. For  $\lambda = 10$  we observe the same trend, but as  $\lambda$  gets bigger, we definatly see that the error gets worse. Same goes for the convergence rate, sa we can see drop over a grade in comparison with  $\lambda \leq 10$