

# From FEM Discretizations to Saddle-Point Matrices

Iterative Methods for Large-Scale Saddle-Point Problems

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George Pólya (1887–1985)

*“In order **to solve** this differential equation you look at it till a solution occurs to you.”*

How to Solve It (Princeton 1945)



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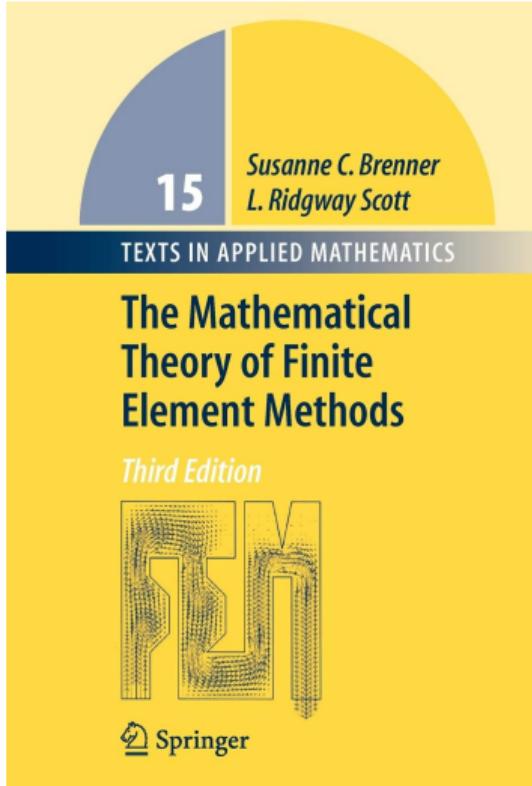
We are gonna settle for **approximating its solution**.

# Overview

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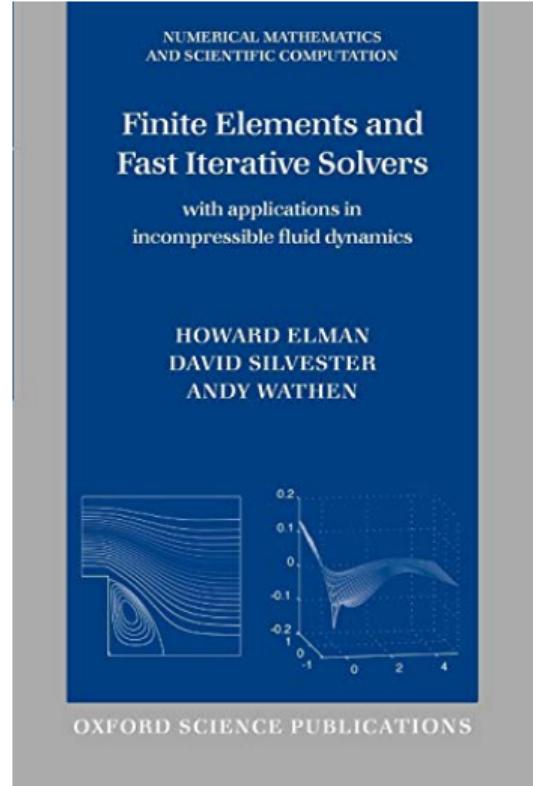
- 1. Basic Concepts**
- 2. Finite Element Spaces**
- 3. Variational crimes**
- 4. Mixed methods**
  - 4.1 The Poisson Equation
  - 4.2 The Stokes Equation
    - Stable discretizations
    - Stabilized discretizations
  - 4.3 The Navier-Stokes Equation

# The main sources



S. C. Brenner and L. R. Scott (2008). *The mathematical theory of finite element methods*. Third. Vol. 15. Texts in Applied Mathematics. Springer, New York, pp. xviii+397. ISBN: 978-0-387-75933-3

H. C. Elman, D. J. Silvester, and A. J. Wathen (2014). *Finite elements and fast iterative solvers: with applications in incompressible fluid dynamics*. Second. Numerical Mathematics and Scientific Computation. Oxford University Press, Oxford, pp. xiv+479. ISBN: 978-0-19-967880-8



# Basic Concepts

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Consider the **two-point boundary value problem** (BVP):

$$\begin{cases} -\frac{d^2u}{dx^2} = f, & \text{in } (0, 1), \\ u(0) = 0, & u'(1) = 0. \end{cases}$$

If  $u$  is the solution and  $v \in V$  is a sufficiently regular function for which  $v(0) = 0$ , then **integration by parts** yields:

$$\begin{aligned} (f, v) &= \int_0^1 f(x)v(x) dx = - \int_0^1 u''(x)v(x) dx \\ &= \int_0^1 u'(x)v'(x) dx = a(u, v). \end{aligned}$$

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# Sobolev Spaces: multi-index notation

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First some notation

Given a **multi-index**  $\alpha \in \mathbb{N}^n$  we denote with

$$|\alpha| = \sum_{i=1}^n \alpha_i,$$

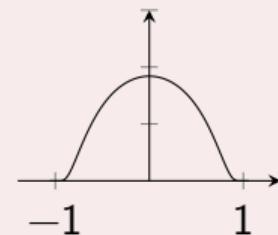
the length of the multi-index. For a function  $\varphi \in \mathcal{C}^\infty$ , we denote the usual **pointwise partial derivative** by

$$D^\alpha \varphi = D_x^\alpha \varphi = \left( \frac{\partial}{\partial x} \right)^\alpha \varphi = \varphi^{(\alpha)} = \partial_x^\alpha \varphi = \left( \frac{\partial}{\partial x_1} \right)^{\alpha_1} \cdots \left( \frac{\partial}{\partial x_n} \right)^{\alpha_n} \varphi.$$

# Sobolev Spaces: building blocks

Definition: compact support functions

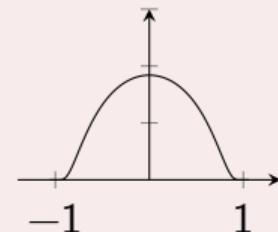
Let  $\Omega \subseteq \mathbb{R}^n$  a domain. We denote by  $\mathcal{D}(\Omega)$  or  $\mathcal{C}_0^\infty(\Omega)$  the set of  $\mathcal{C}^\infty(\Omega)$  **functions with compact support** in  $\Omega$ , i.e., the  $\mathcal{C}^\infty(\Omega)$  functions for which the closure of the set of the points in which they are not zero is compact in  $\Omega$ .



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## Definition: locally integrable functions

Given a domain  $\Omega$  we define the set of **locally integrable functions** as

$$\mathbb{L}_{\text{loc}}^1(\Omega) = \{f : f \in \mathbb{L}^1(K) \forall K \subset \overset{\circ}{\Omega} \text{ } K \text{ compact}\}.$$

# Sobolev Spaces: **weak derivatives**

---

## Definition: weak derivative

We say that a function  $f \in \mathbb{L}_{\text{loc}}^1(\Omega)$  has a **weak derivative**,  $D_w^\alpha f$  provided that there exists a function  $g \in \mathbb{L}_{\text{loc}}^1(\Omega)$  such that

$$\int_{\Omega} g(x)\varphi(x)dx = (-1)^{|\alpha|} \int_{\Omega} f(x)\varphi^{(\alpha)}(x)dx, \quad \forall \varphi \in \mathcal{C}_0^\infty(\Omega).$$

If such  $g$  exists then we define  $D_w^\alpha f = g$ .

# Sobolev Spaces: *weak derivatives*

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A couple of **examples**:

- $f(x) = 1 - |x|$  admits as first weak derivative  $D_w^1 f = g = \chi_{x<0} + \chi_{x>0}$ ,
- If  $f \in \mathcal{C}^{|\alpha|}(\Omega)$  for an arbitrary  $\alpha$ , then  $D_w^\alpha f = D^\alpha f$ .

# Sobolev space

Definition: Sobolev norms and spaces

Let  $k \in \mathbb{N}$ ,  $f \in \mathbb{L}_{\text{loc}}^1(\Omega)$ , suppose that the weak derivative  $D_w^\alpha f$  exists for all  $|\alpha| \leq k$ . We define the **Sobolev norm**

$$\|f\|_{W_p^k(\Omega)} = \begin{cases} \left( \sum_{\alpha: |\alpha| \leq k} \|D_w^\alpha f\|_{\mathbb{L}^p(\Omega)}^p \right)^{\frac{1}{p}}, & 1 \leq p < +\infty \\ \max_{\alpha: |\alpha| \leq k} \|D_w^\alpha f\|_{\mathbb{L}^\infty(\Omega)}, & p = \infty. \end{cases}$$

We define the **Sobolev space**  $W_p^k(\Omega)$  as

$$W_p^k(\Omega) = \left\{ f \in \mathbb{L}_{\text{loc}}^1(\Omega) : \|f\|_{W_p^k(\Omega)} < \infty \right\}.$$

# Sobolev space: a collection of results

## Theorem(s)

- (i) The Sobolev space  $W_p^k(\Omega)$  is a Banach space,
- (ii) Let  $\Omega$  be any open set, then  $C^\infty(\Omega) \cap W_p^k(\Omega)$  is dense in  $W_p^k(\Omega)$  for  $p < \infty$ ,
- (iii)  $k, m \in \mathbb{N}$ ,  $k \leq m$ ,  $1 \leq p \leq \infty \Rightarrow W_p^m(\Omega) \subset W_p^k(\Omega)$ ,
- (iv)  $\Omega$  bounded,  $k \in \mathbb{N}$ ,  $1 \leq p \leq q \leq \infty \Rightarrow W_q^k(\Omega) \subset W_p^k(\Omega)$ ,

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## Definition: Lipschitz boundary

Let  $n \in \mathbb{N}$ ,  $\Omega \subset \mathbb{R}^n$ .  $\Omega$  is a **Lipschitz domain** if  $\forall p \in \partial\Omega$  exists a hyperplane  $H$  of dimension  $n - 1$  through  $p$ , a Lipschitz-continuous function  $g : H \rightarrow \mathbb{R}$  over that hyperplane, and reals  $r > 0$  and  $h > 0$  such that

- $\Omega \cap C = \{x + y\vec{n} \mid x \in B_r(p) \cap H, -h < y < g(x)\}$ ,
- $(\partial\Omega) \cap C = \{x + y\vec{n} \mid x \in B_r(p) \cap H, g(x) = y\}$ ,

where  $\vec{n}$  is a unit vector that is normal to  $H$ ,  $B_r(p) := \{x \in \mathbb{R}^n \mid \|x - p\| < r\}$  is the open ball of radius  $r$ ,  $C := \{x + y\vec{n} \mid x \in B_r(p) \cap H, -h < y < h\}$ .

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- (iv)  $\Omega$  bounded,  $k \in \mathbb{N}$ ,  $1 \leq p \leq q \leq \infty \Rightarrow W_q^k(\Omega) \subset W_p^k(\Omega)$ ,
- (v) If  $\Omega \subset \mathbb{R}^n$  has a Lipschitz boundary,  $\forall k \in \mathbb{N}$ ,  $1 \leq p \leq \infty$ , there exist  $E : W_p^k(\Omega) \rightarrow W_p^k(\mathbb{R}^n)$  satisfying  $Ev|_\Omega = v \quad \forall v \in W_p^k(\Omega)$ , and  $\|Ev\|_{W_p^k(\mathbb{R}^n)} \leq C\|v\|_{W_p^k(\Omega)}$  with  $C$  independent of  $v$ ,
- (vi) If  $\Omega \subset \mathbb{R}^n$  has a Lipschitz boundary,  $\forall k \in \mathbb{N}$ ,  $1 \leq p < \infty$ , and  $m < k$ , then

$$\exists C > 0 : \forall u \in W_p^k(\Omega) \quad \|u\|_{W_\infty^m(\Omega)} \leq C\|u\|_{W_p^k(\Omega)} \quad \begin{cases} k - m \geq n, & p = 1, \\ k - m > \frac{n}{p}, & p > 1. \end{cases}$$

And there exist a function in  $C^m$  in the  $\mathbb{L}^p$  equivalence class of  $u$ .

# Sobolev space: finally we have got an answer!

If you have forgotten the question, we were trying to understand for what  $V$  the solution  $u$  characterized by

$$\text{find } u \in V \text{ such that } a(u, v) = (f, v) \quad \forall v \in V.$$

was a meaningful solution to our initial BVP.

The space

$$V = \{v \in W_2^1(\Omega) : v(0) = 0\}.$$

By the **extension property** and the **Sobolev inequality** we now know that **pointwise values are well defined** for functions  $W_2^1(\Omega)$ .

But all this machinery was needed just to **validate the formulation**, how do we go to a **discrete solution**?

# Building a discrete space

---

To move to a discrete setting, we need to select a **finite subspace**  $S \subset V$ . With this, we can impose the **Ritz-Galerkin** conditions:

$$\text{find } u_S \in S \text{ such that } a(u_S, v) = (f, v) \quad \forall v \in S.$$

- Since  $S$  is finite-dimensional, there exists a basis  $\phi_1, \dots, \phi_n$  of  $S$ ,
- Thus,  $u_S = \sum_{i=1}^n U_i \phi_i \in S$ ,  $U_i \in \mathbb{R}$  for  $i = 1, \dots, n$ ,
- Ritz-Galerkin conditions are now a **system of linear equations** for the unknown coefficients  $U_i$ :

$$K\mathbf{U} = \mathbf{F},$$

with

- $\mathbf{U} = (U_1, \dots, U_n)^T \in \mathbb{R}^n$ ,
- $\mathbf{F} = (F_1, \dots, F_n)^T \in \mathbb{R}^n$ , for  $F_i = (f, \phi_i)$ ,
- $\mathbf{K} = (K_{ij}) \in \mathbb{R}^{n \times n}$ , for  $K_{ij} = a(\phi_i, \phi_j)$ .

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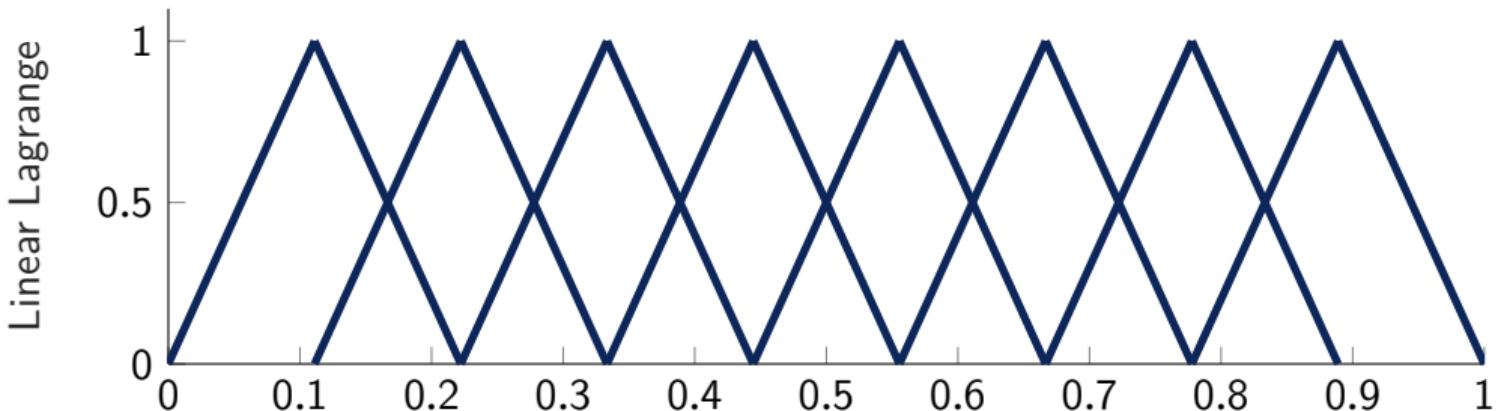
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What are examples of such  $S$ ?

# Lagrange basis

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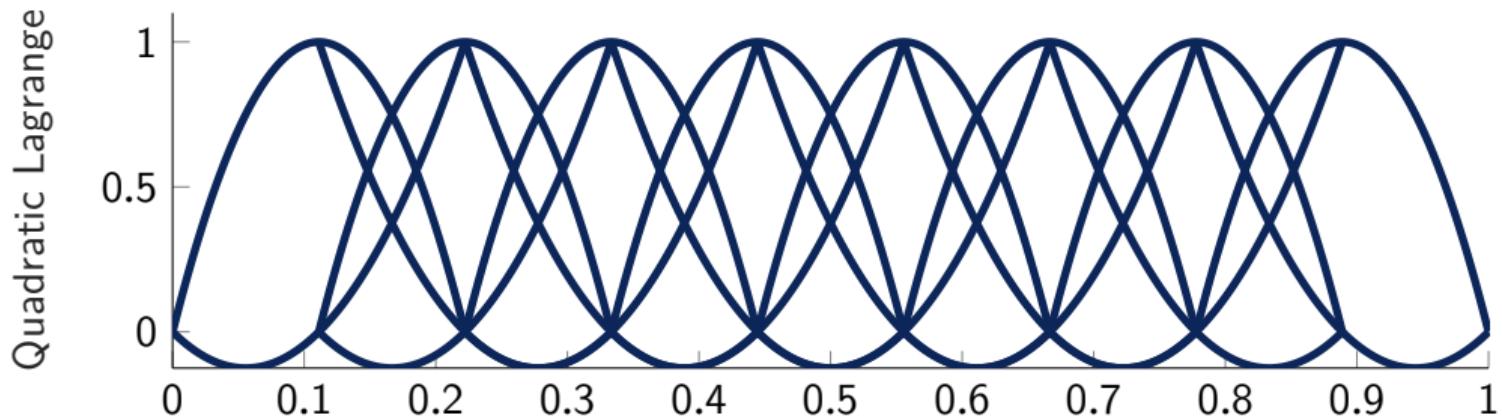


Let  $0 = x_0 < x_1 < x_2 < \dots < x_n = 1$ , we consider the linear space of functions  $v \in S$  s.t.

- (i)  $v \in C^0([0, 1])$ ,
- (ii)  $v|_{[x_{i-1}, x_i]}$  is a linear polynomial,  $i = 1, \dots, n$ , and
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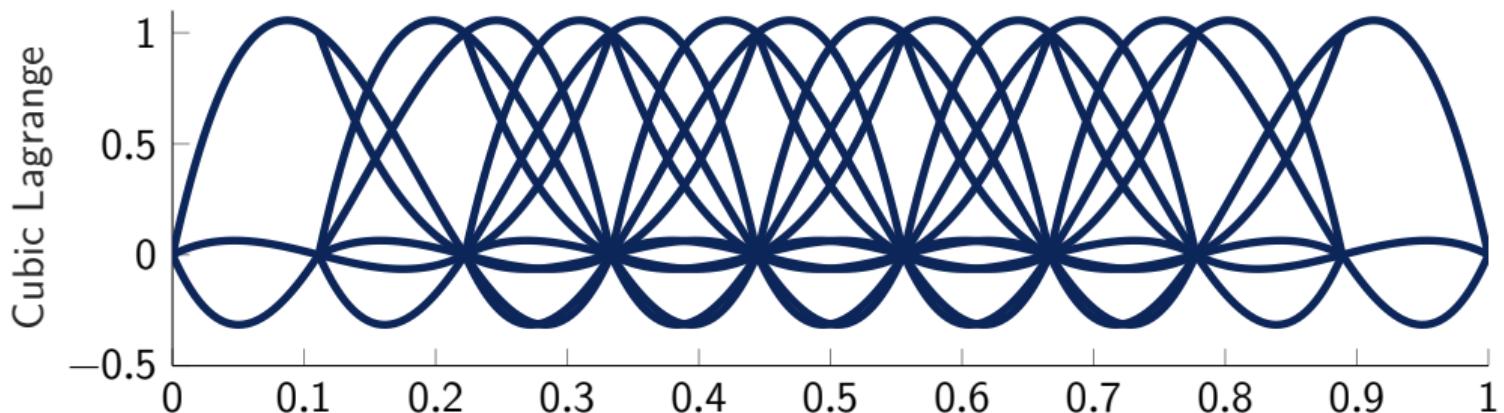


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- (ii)  $v|_{[x_{i-3}, x_i]}$  is a cubic polynomial,  $i = 3, \dots, n$ , and
- (iii)  $v(0) = 0$ .

# Convergence and approximation properties

We have a **theoretical framework for solutions**, examples of **discrete spaces**, but **what about convergence?**

## Sobolev meets Hilbert

$W_p^k$  is a Hilbert space for  $p = 2$ , with inner product

$$\langle f, g \rangle_{W_2^k(\Omega)} = \sum_{|\alpha| \leq k} (D^\alpha f, D^\alpha g).$$

We write:  $H^k(\Omega) \equiv W_2^k(\Omega)$ , and  $H_0^k(\Omega) = \{v \in W_2^k(\Omega) : v \equiv 0 \text{ on } \partial\Omega\}$ .

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## Our $V$ space

$$V = H_0^1([0, 1]).$$

# Convergence and approximation properties

## Variational problem

For a given Hilbert space  $V$ , a bilinear form  $a : V \times V \rightarrow \mathbb{R}$  and a linear functional  $F : V \rightarrow \mathbb{R}$ , find  $u \in V$  such that:

$$a(u, v) = F(v), \quad \text{for all } v \in V. \tag{VP}$$

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## Theorem (Lax-Milgram).

Let  $V$  be a Hilbert space,  $a : V \times V \rightarrow \mathbb{R}$  a bilinear form, and  $F : V \rightarrow \mathbb{R}$  a linear functional s.t.

**Coercivity**  $\exists c_1 > 0$  s.t.  $a(v, v) \geq c_1 \|v\|_V^2$  for all  $v \in V$ .

**Continuity**  $\exists c_2, c_3 > 0$  s.t.  $a(v, w) \leq c_2 \|v\|_V \|w\|_V$ , and  $|F(v)| \leq c_3 \|v\|_V$  for all  $v, w \in V$ .

$\Rightarrow \exists! u \in V$  satisfying (VP), and  $\|u\|_V \leq \frac{1}{c_1} \|F\|_{V^*}$ .

# Convergence and approximation properties

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In the **conforming Galerkin approach** we chose a (finite-dimensional) closed subspace  $V_h \subset V$  and look for  $u_h \in V_h$  satisfying

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## Theorem

Under the assumptions of the Lax-Milgram, for any closed subspace  $V_h \subset V$ , there exists a unique solution  $u_h \in V_h$  of  $(\text{VP}_h)$  satisfying

$$\|u_h\|_V \leq \frac{1}{c_1} \|F\|_{V^*}.$$

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## Céa's lemma

Let  $u_h$  be the solution of  $(\text{VP}_h)$  for given  $V_h \subset V$  and  $u$  be the solution of variational problem  $(\text{VP})$ . Then,

$$\|u - u_h\|_V \leq \frac{c_2}{c_1} \inf_{v_h \in V_h} \|u - v_h\|_V,$$

where  $c_1$  and  $c_2$  are the constants from the **coercivity** and **continuity** assumptions.

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## The conforming idea

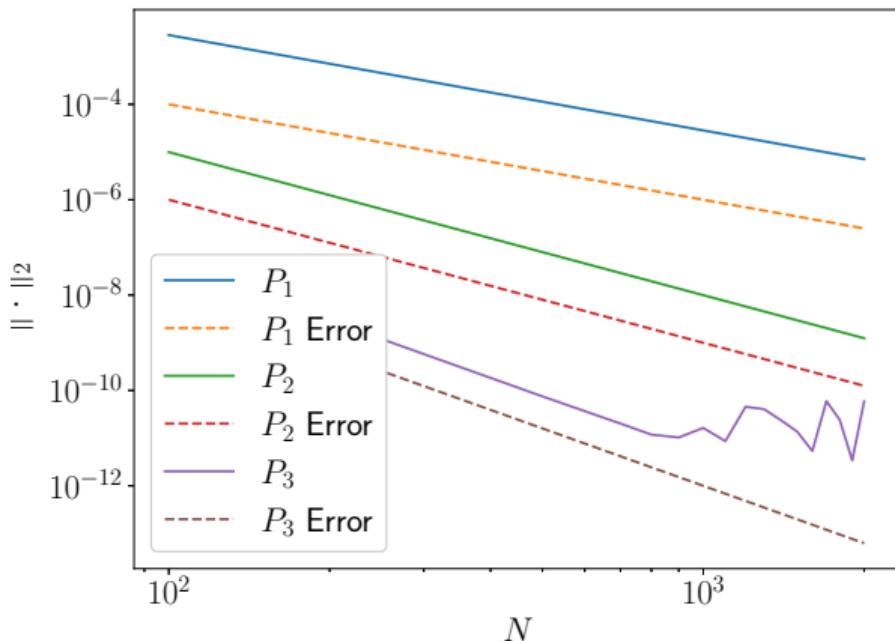
The **error** of the **conforming Galerkin** approach is **determined** by the **approximation error** of the exact solution **in**  $V_h$ .

# Error estimate on 1D problem

The **test problem** we consider is

$$\begin{cases} -u_{xx} = f(x), & x \in \Omega, \\ u(x) = g(x), & x \in \partial\Omega \end{cases}$$

for  $f(x) = 2 \cos(x)/e^x$  and  $g(x) = \sin(x)/e^x$  on  $\Omega = (0, 10)$ . We discretize it with Lagrangian 1, 2 and 3 elements and report the computed error:  $\|u - u_{\text{ex}}\|_{L^2(\Omega_h)}$  on the uniform grid with  $N$  points.



# FEniCSx Code Example

---

We can implement this simple case in the FEniCSx Library in few lines of code

1. First we need to load some packages

```
from mpi4py import MPI # Needed for the MPI environment
import numpy as np # The numpy package support
from dolfinx import mesh # Handler for the meshes
from dolfinx import fem # FEM building blocks
from dolfinx.fem import FunctionSpace # FEM Function Spaces
import ufl # Language for building up variational formulations
```

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```
nx = 500
Omegah = mesh.create_interval(comm=MPI.COMM_WORLD, nx=nx,
    ↪ points=(0,10))
V = FunctionSpace(Omegah, ("CG", 1))
```

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3. Then we need a bit of work to impose **essential boundary conditions**

```
g = fem.Function(V)
g.interpolate(lambda x: np.sin(x[0])/np.exp(x[0]))
tdim = Omegah.topology.dim
fdim = tdim - 1
Omegah.topology.create_connectivity(fdim, tdim)
boundary_facets =
    np.flatnonzero(mesh.compute_boundary_facets(Omegah.topology))
boundary_dofs = fem.locate_dofs_topological(V, fdim,
    boundary_facets)
bc = fem.dirichletbc(g, boundary_dofs)
```

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3. Then we need a bit of work to impose **essential boundary conditions**
4. We create **test** and **trial functions**

```
u = ufl.TrialFunction(V)
v = ufl.TestFunction(V)
```

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2. Then we **build** the **mesh** and the **function space**
3. Then we need a bit of work to impose **essential boundary conditions**
4. We create **test** and **trial functions**
5. We build the source and the **variational formulation**

```
f = fem.Function(V)
f.interpolate(lambda x: 2.0*np.cos(x[0])/np.exp(x[0]))
a = ufl.dot(ufl.grad(u), ufl.grad(v)) * ufl.dx
F = f * v * ufl.dx
```

# FEniCSx Code Example

---

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6. Finally we **solve** the **linear system** (directly...it's 1D!)

```
problem = fem.petsc.LinearProblem(a, F, bcs=[bc],  
    ↳ petsc_options={"ksp_type": "preonly", "pc_type": "lu"})  
uh = problem.solve()
```

# FEniCSx Code Example

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7. and **compute the error**: Error\_L2 : 1.14e-04

```
V2 = fem.FunctionSpace(Omegah, ("CG", 2))
uex = fem.Function(V2)
uex.interpolate(lambda x: np.sin(x[0])/np.exp(x[0]))
L2_error = fem.form(ufl.inner(uh - uex, uh - uex) * ufl.dx)
error_local = fem.assemble_scalar(L2_error)
error_L2 = np.sqrt(Omegah.comm.allreduce(error_local, op=MPI.SUM))
```

# FEniCSx Code Example

---

We can implement this simple case in the FEniCSx Library in few lines of code

1. First we need to load some packages
2. Then we **build** the **mesh** and the **function space**
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To **run** the example there is a  Python notebook using FEniCSx shared through  
 [bit.ly/3tTEBfl](https://bit.ly/3tTEBfl) (and executed on **Google Colab**).

# FEM Spaces

We can build many different types of Finite Elements.

## FE Definition (Ciarlet, 1978)

A *finite element* is a triple  $(K, \mathcal{P}, \mathcal{N})$  where

- (i)  $K \subset \mathbb{R}^n$  is a simply connected bounded open set with piecewise smooth boundary (*element domain*);
- (ii)  $\mathcal{P}$  is a finite-dimensional space of functions defined on  $K$  (*space of shape functions*);
- (iii)  $\mathcal{N} = \{N_1, \dots, N_d\}$  is a basis of  $\mathcal{P}^*$  (*degrees of freedom*).

# FEM Spaces

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- (iii)  $\mathcal{N} = \{N_1, \dots, N_d\}$  is a basis of  $\mathcal{P}^*$  (*degrees of freedom*).

## Dual basis definition

Let  $(K, \mathcal{P}, \mathcal{N})$  be a finite element. A basis  $\{\psi_1, \dots, \psi_d\}$  of  $\mathcal{P}$  is called *dual basis* or *nodal basis* to  $\mathcal{N}$  if  $N_i(\psi_j) = \delta_{ij}$ .

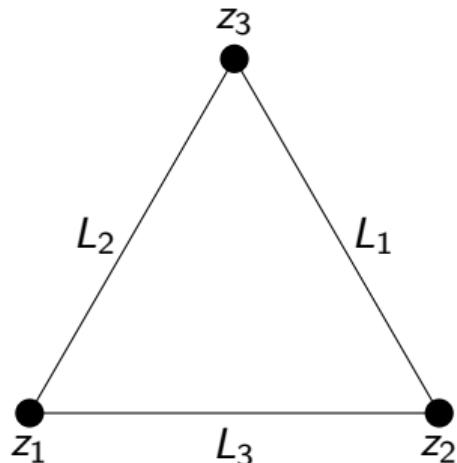
# A lineup of some usual (and unusual) suspects

---

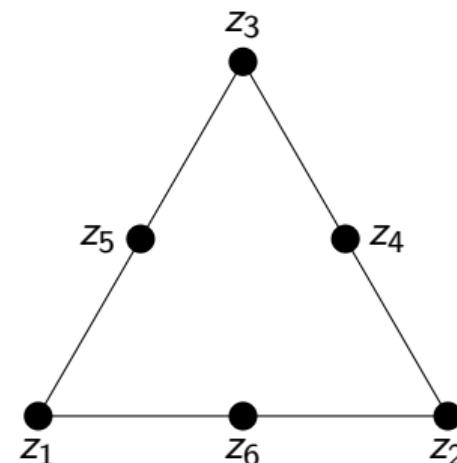


# FEM Spaces: triangular finite elements

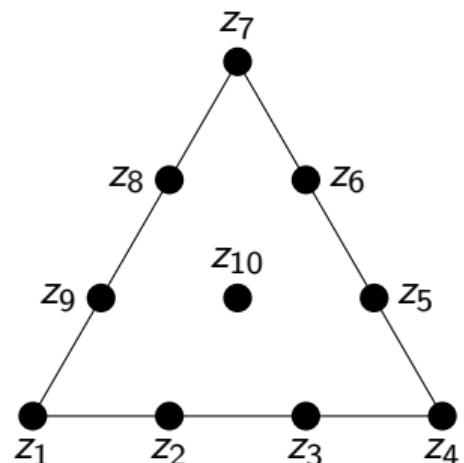
$K$  any triangle, space  $\mathcal{P}_k$  of bivariate polynomials of degree  $\leq k$ ,



Linear Lagrange element



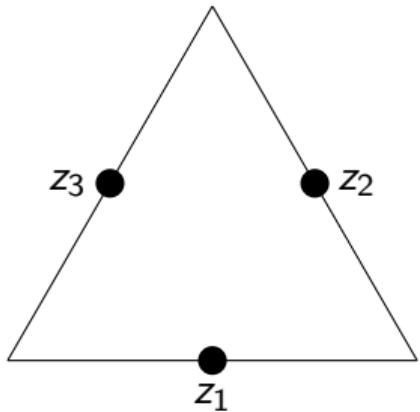
Quadratic Lagrange element



Cubic Lagrange element

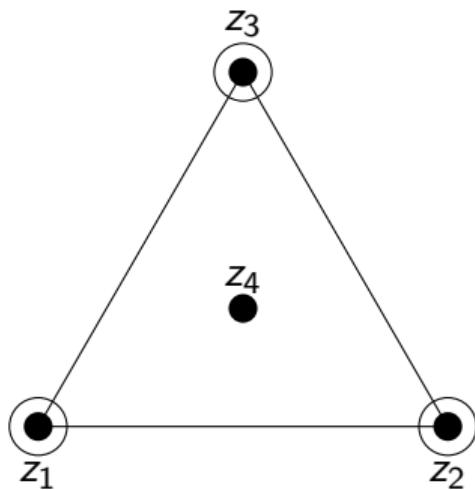
“●” Point evaluations determining the  $\mathcal{N} = \{\mathcal{N}_1, \dots, \mathcal{N}_{\frac{1}{2}(k+1)(k+2)}\}$ .

# FEM Spaces: triangular finite elements

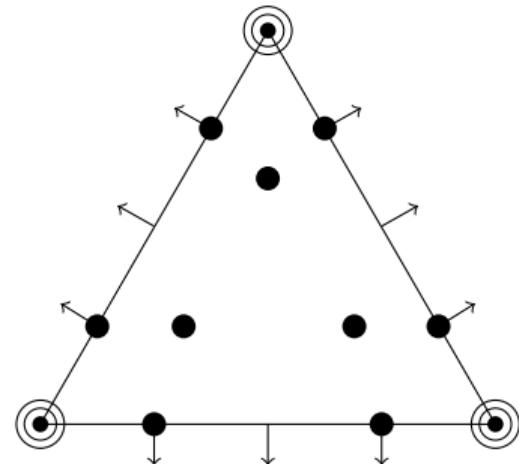


Linear nonconforming  
Crouzeix-Raviart element

- Point evaluations, ○ Gradient evaluations, ⊙ Three second derivative, ↑ Normal derivative.



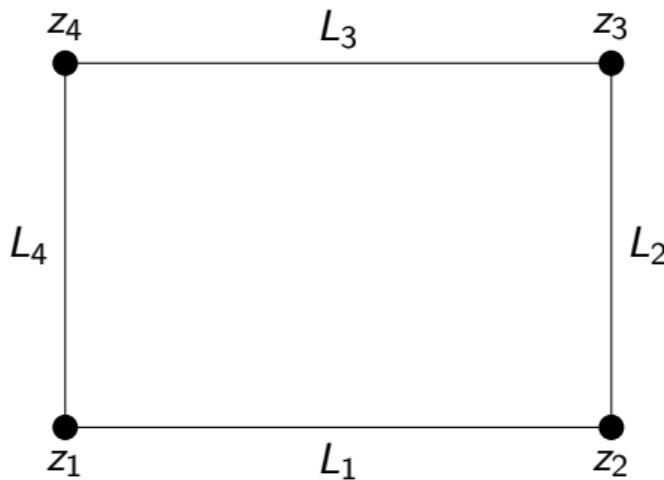
Cubic Hermite element



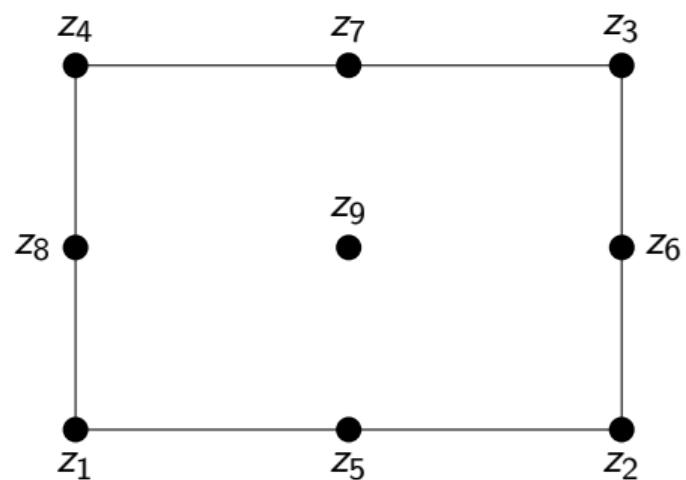
Quintic Argyris element

# FEM Spaces: rectangular finite elements

$K$  any rectangle, space  $\mathcal{Q}_k = \left\{ \sum_j c_j p_j(x) q_j(x), \ p_j, q_j \in \mathbb{P}_{\leq k}[x] \right\}$ ,



Bilinear Lagrange element



Biquadratic Lagrange element

- Point evaluations for  $\mathcal{N} = \{N_1, \dots, N_d\}$ ,  $d = \dim \mathcal{Q}_k = (\dim \mathbb{P}_{\leq k}[x])^2$ .



# FEM Spaces: it's a vast world

---

Much of what we discussed and of what we are going to discuss in the next slides can be applied to *FEM-adjacent* methods, a (obviously not exhaustive) list of ideas:

DG: Discontinuous Galerkin, (Cockburn, Karniadakis, and Shu 2000) for a general overview, linear solvers (Ayuso de Dios et al. 2014; Dobrev et al. 2006)...

IgA: Isogeometric Analysis, (Cottrell, Hughes, and Bazilevs 2009) for a general overview, adaptive meshes (Giannelli, Jüttler, and Speleers 2012; Patrizi and Dokken 2020), linear solvers (Donatelli et al. 2015; Horníková, Vuik, and Egermaier 2021; Sangalli and Tani 2016)...

VEM: Virtual Elements, (Beirão da Veiga et al. 2014, 2016) for a general overview, linear solvers (Antonietti, Mascotto, and Verani 2018; Dassi and Scacchi 2020)...

Another nice source of information is: [defelement.com](http://defelement.com).

# Variational crimes

---



“The crime is now logical and reasonable.”

Murder for Christmas, A. Christie

# The Penal Code

---

- 💣 *Petrov–Galerkin* approaches, where the function  $u$  satisfying  $a(u, v)$  for all  $v \in V$  is an element of  $U \neq V$ ;
- 💣 *non-conforming* approaches, where the discrete spaces  $U_h$  and  $V_h$  are not subspaces of  $U$  and  $V$ , respectively; and
- 💣 *non-consistent* approaches, where the discrete problem involves a bilinear form  $a_h \neq a$  (and  $a_h$  might not be well-defined for all  $u \in U$ ).

We thus need a **more general framework** that covers these cases as well.

- $U, V$  be Banach spaces, with  $V$  reflexive,  $U^*, V^*$  denote their topological duals
- Given  $a : U \times V \rightarrow \mathbb{R}$  bilinear,  $F \in V^*$  continuous we look for  $u \in U$  satisfying

$$a(u, v) = F(v) \quad \text{for all } v \in V. \tag{W}$$

# Existence and uniqueness in a world full of crimes

## Theorem Banach–Nečas–Babuška

Let  $U$  and  $V$  be Banach spaces and  $V$  be reflexive. If  $a : U \times V \rightarrow \mathbb{R}$  and  $F : V \rightarrow \mathbb{R}$  satisfy:

- (i) *Inf-sup condition*: there exists a  $c_1 > 0$  such that

$$\inf_{u \in U} \sup_{v \in V} \frac{a(u, v)}{\|u\|_U \|v\|_V} \geq c_1.$$

- (ii) *Continuity*: there exist  $c_2, c_3$  such that

$$|a(u, v)| \leq c_2 \|u\|_U \|v\|_V, \quad |F(v)| \leq c_3 \|v\|_V, \quad \forall u \in U, \forall v \in V$$

- (iii) *Injectivity*: for any  $v \in V$   $a(u, v) = 0$  for all  $u \in U$  implies  $v = 0$ .

Then there exists a unique solution  $u \in U$  to  $(\mathcal{W})$ , which satisfies

$$\|u\|_U \leq \frac{1}{c_1} \|F\|_{V^*}.$$

# Mixed Methods - The Poisson equation

---

Let us start again from the **Poisson equation** with homogeneous Dirichlet conditions

$$\begin{cases} -\Delta u = -\nabla \cdot \nabla u = -\nabla^2 u = -\operatorname{div} \operatorname{grad} u = f, & \mathbf{x} \in \Omega \subset \mathbb{R}^n \\ u = 0, & \mathbf{x} \in \partial\Omega. \end{cases}$$

We introduce an **auxiliary variable**  $\sigma = \nabla u \in \mathbb{L}^2(\Omega)^n$  and rewrite it as

$$\begin{cases} \nabla u - \sigma = 0, \\ -\nabla \cdot \sigma = f. \end{cases}$$

This system can be formulated in variational form in two different ways:

1. we *formally* integrate by parts in the second equation  $\Rightarrow$  *primal* approach,
2. we *formally* integrate by parts in the first equation  $\Rightarrow$  *dual* approach.

# Mixed Methods - The Poisson equation - Primal

---

We look for  $(\sigma, u) \in \mathbb{L}^2(\Omega)^n \times \mathbb{H}_0^1(\Omega)$  satisfying

$$\begin{cases} (\sigma, \tau) - (\tau, \nabla u) = 0 & \text{for all } \tau \in \mathbb{L}^2(\Omega)^n, \\ -(\sigma, \nabla v) = -(f, v) & \text{for all } v \in \mathbb{H}_0^1(\Omega). \end{cases}$$

that we can restate in **abstract form** as

$$a(\sigma, \tau) = (\sigma, \tau) : V \times V \rightarrow \mathbb{R}, \quad b(v, \mu) = -(v, \nabla \mu) : V \times M \rightarrow \mathbb{R},$$

on the two (reflexive) Banach spaces  $V = \mathbb{L}^2(\Omega)^n$  and  $M = \mathbb{H}_0^1(\Omega)$  for the problem

Find  $u, \lambda$  s.t.  $\begin{cases} a(u, v) + b(v, \lambda) = \langle f, v \rangle_{V^*, V} & \text{for all } v \in V, \\ b(u, \mu) = \langle g, \mu \rangle_{M^*, M} & \text{for all } \mu \in M. \end{cases}$

# Mixed Methods - Abstract Saddle-Point formulation

---

To uncover the connection with the discrete case we are aiming at, let us reformulate the previous in operator form by introducing

$$\begin{aligned} A : V &\rightarrow V^*, \quad \langle Au, v \rangle_{V^*, V} = a(u, v) \quad \text{for all } v \in V, \\ B : V &\rightarrow M^*, \quad \langle Bu, \mu \rangle_{M^*, M} = b(u, \mu) \quad \text{for all } \mu \in M, \\ B^* : M &\rightarrow V^*, \quad \langle B^*\lambda, v \rangle_{V^*, V} = b(v, \lambda) \quad \text{for all } v \in V. \end{aligned}$$

From which we rewrite our problem as

$$\text{Find } u, \lambda \text{ s.t. } \begin{cases} Au + B^*\lambda = f & \text{in } V^*, \\ Bu = g & \text{in } M^*. \end{cases}$$

At this stage, this should be very familiar!

# Mixed Methods - Abstract Saddle-Point formulation

## Abstract Saddle-Point

$$\text{Find } u, \lambda \text{ s.t. } \begin{cases} Au + B^* \lambda = f & \text{in } V^*, \\ Bu = g & \text{in } M^*. \end{cases}$$

- If  $B$  is *invertible*  $\Rightarrow$  **existence** and **uniqueness** first of  $u$  and then of  $\lambda$  follow immediately,

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- Usually, we **are not this lucky**, remember our **starting example**:

$$\langle B\sigma, \mu \rangle_{(\mathbb{H}_0^1)^*, \mathbb{H}_0^1} = b(\sigma, \mu) = -(\sigma, \nabla \mu),$$

that is we have to **require** that  $A$  is **injective** and **coercive** on  $\ker B$  to obtain a **unique**  $u$ ...

# Mixed Methods - Abstract Saddle-Point formulation

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- ...for the **existence** of  $\lambda$  we then need  $B^*$  to be **surjective**.

# Mixed Methods - Banach–Nečas–Babuška

## Theorem (Continuous Brezzi)

We assume that

- (i)  $a : V \times V \rightarrow \mathbb{R}$  satisfies the conditions of the Banach–Nečas–Babuška Theorem for  $U = V = \ker B$
- (ii)  $b : V \times M \rightarrow \mathbb{R}$  is such that the Ladyzhenskaya–Babuška–Brezzi condition holds

$$\exists \beta > 0 : \inf_{\mu \in M} \sup_{v \in V} \frac{b(v, \mu)}{\|v\|_V \|\mu\|_M} \geq \beta.$$

$\Rightarrow \exists! (u, \lambda) \in V \times M$  solving the mixed saddle-point system and satisfying

$$\|u\|_V + \|\lambda\|_M \leq C(\|f\|_{V^*} + \|g\|_{M^*}).$$

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- ☛  $a(u, v)$  has to satisfy the BNB condition only on  $\ker B$ , not on all of  $V$ !
- ☛ LBB condition couples  $V$  and  $M$  spaces, this is going to have repercussions in a moment!

# Mixed Methods - Back to Poisson

## Mixed Continuous Primal Poisson Problem

Find  $(\sigma, u) \in \mathbb{L}^2(\Omega)^n \times \mathbb{H}_0^1(\Omega)$  s.t.

$$\begin{cases} (\sigma, \tau) - (\tau, \nabla u) = 0 & \forall \tau \in \mathbb{L}^2(\Omega)^n, \\ -(\sigma, \nabla v) = -(f, v) & \forall v \in \mathbb{H}_0^1(\Omega). \end{cases}$$

Coercivity:  $a$  is coercive on  $V$  with constant  $\alpha = 1$ ,

LBB: chose  $v \in \mathbb{H}_0^1(\Omega) = M$  and take  $\tau = -\nabla v \in \mathbb{L}^2(\Omega)^n = V$ , then

$$\sup_{\tau \in V} \frac{b(\tau, v)}{\|\tau\|_V} = \sup_{\tau \in V} \frac{-(\tau, \nabla v)}{\|\tau\|_{\mathbb{L}^2(\Omega)^n}} \geq \frac{(\nabla v, \nabla v)}{\|\nabla v\|_{\mathbb{L}^2(\Omega)^n}} = |v|_{\mathbb{H}^1} \geq c_\Omega^{-1} \|v\|_M$$

❶ to get  $C_\Omega^{-1}$  we use **Poincaré inequality**: for  $1 \leq p < \infty$ ,  $\Omega$  an open bounded set  $\Rightarrow$

$\exists c_\Omega : \|f\|_{W_p^1(\Omega)} \leq c_\Omega |f|_{W_p^1(\Omega)}$  depending only on  $p$  and  $\Omega$ .

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# Mixed Methods - Galerkin Approach

## Abstract Saddle-Point

Find  $u, \lambda$  s.t.  $\begin{cases} Au + B^* \lambda = f & \text{in } V^*, \\ Bu = g & \text{in } M^*. \end{cases}$

- $V_h \subset V, M_h \subset M,$
- 👑  $V_h$  and  $M_h$  **cannot be** selected independently!

# Mixed Methods - Galerkin Approach

## Abstract Discrete Saddle-Point

Find  $u_h, \lambda_h$  such that

$$\begin{cases} a(u_h, v_h) + b(v_h, \lambda_h) = \langle f, v_h \rangle_{V^*, V} \quad \forall v_h \in V_h, \\ b(u_h, \mu_h) = \langle g, \mu_h \rangle_{M^*, M} \quad \forall \mu_h \in M_h. \end{cases}$$

- $V_h \subset V, M_h \subset M,$
- $V_h$  and  $M_h$  **cannot be** selected independently!

## Theorem (Discrete Brezzi)

$$\text{If } \exists \alpha_h > 0 : \inf_{u_h \in \ker B_h} \sup_{v_h \in \ker B_h} \frac{a(u_h, v_h)}{\|u_h\|_V \|v_h\|_V} \geq \alpha_h,$$

$$\text{If } \exists \beta_h > 0 : \inf_{\mu_h \in M_h} \sup_{v_h \in V_h} \frac{b(v_h, \mu_h)}{\|v_h\|_V \|\mu_h\|_M} \geq \beta_h.$$

$\Rightarrow \exists! (u_h, \lambda_h) \in V_h \times M_h$  solving the discrete saddle-point and satisfying

$$\|u_h\|_V + \|\lambda_h\|_M \leq C_h(\|f\|_{V^*} + \|g\|_{M^*}).$$

## Mixed Methods - Dual approach

---

We integrate by parts the **first equation**:

$$\int_{\Omega} (\operatorname{div} \tau) w \, dx + \int_{\Omega} \tau \cdot \nabla w \, dx = \int_{\partial\Omega} (\tau \cdot \nu) w \, dx$$

We need to define the **proper Sobolev space**.

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$\mathbb{H}(\operatorname{div})$

We define the space

$$\mathbb{H}(\operatorname{div}) = \{\tau \in \mathbb{L}^2(\Omega)^n : \operatorname{div} \tau \in \mathbb{L}^2(\Omega)\},$$

with the norm

$$\|\tau\|_{\mathbb{H}(\operatorname{div})}^2 := \|\tau\|_{\mathbb{L}^2(\Omega)^n}^2 + \|\operatorname{div} \tau\|^2.$$

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Well posedness of the *normal trace*.

$\mathcal{C}^\infty(\overline{\Omega})^n$  is dense in  $\mathbb{L}^2(\Omega)^n \supset \mathbb{H}(\operatorname{div}) \Rightarrow \tau \in \mathbb{H}(\operatorname{div})$  has  $(\tau|_{\partial\Omega} \cdot \nu) \in \mathbb{H}^{-1/2}(\partial\Omega)$ .

# Mixed Methods - Dual approach

---

The **dual problem** is then

$$\text{find } (\sigma, u) \in \mathbb{H}(\text{div}) \times \mathbb{L}^2 \text{ s.t. } \begin{cases} (\sigma, \tau) + (\text{div } \tau, u) = 0 & \forall \tau \in \mathbb{H}(\text{div}), \\ (\text{div } \sigma, v) = -(f, v) & \forall v \in \mathbb{L}^2. \end{cases}$$

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- we have used that  $u|_{\partial\Omega} = 0$ ,
- This is a **general saddle problem** with  $V = \mathbb{H}(\text{div})$ , and  $M = \mathbb{L}^2$ :

$$a(\sigma, \tau) = (\sigma, \tau), \quad b(\sigma, v) = (\text{div } \sigma, v).$$

and  $a$  and  $b$  bounded by Cauchy–Schwarz inequality.

# Mixed Methods - Dual approach

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- we have used that  $u|_{\partial\Omega} = 0$ ,
- This is a **general saddle problem** with  $V = \mathbb{H}(\text{div})$ , and  $M = \mathbb{L}^2$ :

$$a(\sigma, \tau) = (\sigma, \tau), \quad b(\sigma, v) = (\text{div } \sigma, v).$$

and  $a$  and  $b$  bounded by Cauchy–Schwarz inequality.

- For the **existence of the solution** we need to prove **coercivity** for  $a$ .

$$\ker B = \{\tau \in \mathbb{H}(\text{div}) : (\text{div } \tau, v) = 0 \forall v \in \mathbb{L}^2\}$$

Since  $\text{div } \tau \in \mathbb{L}^2$  we have  $\|\text{div } \tau\|_{\mathbb{L}^2} = 0$  whenever  $\tau \in \ker B \subset \mathbb{H}(\text{div})$ , and therefore

$$a(\tau, \tau) = \|\tau\|_{\mathbb{L}^2(\Omega)^n}^2 = \|\tau\|_{\mathbb{H}(\text{div})}^2 \quad \forall \tau \in \ker B,$$

indeed we have just proved **coercivity with  $\alpha = 1$** .

# Mixed Methods - Dual approach

---

The **dual problem** is then

$$\text{find } (\sigma, u) \in \mathbb{H}(\text{div}) \times \mathbb{L}^2 \text{ s.t. } \begin{cases} (\sigma, \tau) + (\text{div } \tau, u) = 0 & \forall \tau \in \mathbb{H}(\text{div}), \\ (\text{div } \sigma, v) = -(f, v) & \forall v \in \mathbb{L}^2. \end{cases}$$

- we have used that  $u|_{\partial\Omega} = 0$ ,
- This is a **general saddle problem** with  $V = \mathbb{H}(\text{div})$ , and  $M = \mathbb{L}^2$ :

$$a(\sigma, \tau) = (\sigma, \tau), \quad b(\sigma, v) = (\text{div } \sigma, v).$$

and  $a$  and  $b$  bounded by Cauchy–Schwarz inequality.

- The form  $a$  is coercive on  $\ker B$  with  $\alpha = 1$ ,
- We now need to verify the **LBB condition**. This requires some work.

# Mixed Methods - Dual approach - LBB

## Assumption:

We make the simplifying assumption of having  $\partial\Omega$  represented by a  $C^1$  function or, analogously, having  $\Omega$  convex.

## Lemma (Surjectivity)

For any  $f \in \mathbb{L}^2$ , there exists a function  $\tau \in \mathbb{H}(\text{div})$  with  $\text{div } \tau = f$  and  $\|\tau\|_{\mathbb{H}(\text{div})} \leq C\|f\|_{\mathbb{L}^2}$ .

- The space  $\mathbb{H}^1(\Omega)^n \subset \mathbb{H}(\text{div})$ , thus if we take a  $v \in M = \mathbb{L}^2$  and the corresponding  $\tau_v \in \mathbb{H}(\text{div})$  given by the surjectivity lemma (i.e.,  $\text{div } \tau_v = v$ ) we find

$$\sup_{\tau \in V} \frac{b(\tau, v)}{\|\tau\|_V} = \sup_{\tau \in V} \frac{(\text{div } \tau, v)}{\|\tau\|_{\mathbb{H}(\text{div})}} \geq \frac{(\tau_v, v)}{\|\tau_v\|_{\mathbb{H}(\text{div})}} \geq \frac{(v, v)}{C\|v\|_{\mathbb{L}^2(\Omega)}} = \frac{1}{c}\|v\|_{\mathbb{L}^2(\Omega)}.$$

We have then proved the LBB condition for  $\beta = \frac{1}{C}$ .

# Mixed Methods - Dual approach - LBB

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- We have then proved the LBB condition for  $\beta = \frac{1}{C}$ .
- By **Continuous Brezzi** we have  $\exists! (\sigma, u) \in V \times M$  solving the **saddle problem** and such that

$$\|\sigma\|_{\mathbb{H}(\text{div})} + \|u\|_{\mathbb{L}^2(\Omega)} \leq C\|f\|_{\mathbb{L}^2(\Omega)}.$$

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- The solution  $u$  we have obtained seem to be only in  $\mathbb{L}^2$  ...but  $u$  satisfies

$$(\sigma, \tau) + (\text{div } \tau, u) = 0,$$

thus an **integration by parts** shows that  $u$  has a weak derivative and satisfies boundary conditions, that is,  $u$  is where it should be  $u \in \mathbb{H}_0^1(\Omega)$ .

# Mixed Methods - Galerkin Approach for Poisson

Let us build the **discrete problem**.

A property of the form

We can (but won't) show that for any partition  $\Omega_h$  of  $\Omega$

$$\{\tau \in \mathbb{L}^2(\Omega)^n : \tau|_{\Omega_j} \in \mathbb{H}^1(\Omega_j) \text{ and } \tau|_{\Omega_j} \cdot \hat{\mathbf{n}} = \tau_{\Omega_i} \cdot \hat{\mathbf{n}} \forall \overline{\Omega}_i \cap \overline{\Omega}_j \neq \emptyset\} \subset \mathbb{H}^1(\Omega).$$

In layman terms, piecewise differentiable functions with continuous normal traces across elements are in  $\mathbb{H}^1(\Omega)$ .

This observation is crucial for building **conformal FEM spaces** for this problem.

- We could consider are the *Raviart-Thomas elements* (Raviart and Thomas 1977),
- The other usual option are the **Brezzi-Douglas-Marini elements** (Brezzi, Douglas, and Marini 1985),

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- We could consider are the *Raviart-Thomas elements* (Raviart and Thomas 1977),
- The other usual option are the **Brezzi-Douglas-Marini elements** (Brezzi, Douglas, and Marini 1985),
- To build the matrices we will use the code from (Zhang 2015).

## Mixed Methods - Test problem

---

We consider a **more general formulation** of the Poisson problem

$\Omega \subset \mathbb{R}^2$  a polygonal domain, with boundary

$$\partial\Omega = \Gamma_D \cap \Gamma_N \quad (\Gamma_D \cap \Gamma_N = \emptyset, \mu(\Gamma_D) \neq 0)$$

$$\begin{cases} -\nabla \cdot (\alpha(x)\nabla u) = f, & \text{in } \Omega, \\ -\alpha\nabla u \cdot \hat{\mathbf{n}} = g_N, & \text{on } \Gamma_N, \\ u = g_D, & \text{on } \Gamma_D. \end{cases}$$

With

- $f \in \mathbb{L}^2(\Omega)$ ,
- $g_D \in \mathbb{H}^{1/2}(\Gamma_D)$  and  $g_N \in \mathbb{L}^2(\Gamma_N)$ ,
- $\alpha(x)$  positive **piecewise constant**.

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The weak form is then for  
 $(\tau, v) \in \mathbb{H}_N(\text{div}) \times \mathbb{L}^2(\Omega)$

$$\begin{cases} (\alpha^{-1}\boldsymbol{\sigma}, \boldsymbol{\tau}) - (\text{div } \boldsymbol{\tau}, u) = -(\boldsymbol{\tau} \cdot \hat{\mathbf{n}}, g_D)_{\Gamma_D} \\ (\text{div } \boldsymbol{\sigma}, v) = (f, v) \end{cases}$$

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where

- $\boldsymbol{\sigma} = -\alpha(x)\nabla u$  is the **flux**,
- $\mathbb{H}_N(\text{div}) = \{\boldsymbol{\tau} \in \mathbb{H}(\text{div}) : \boldsymbol{\tau} \cdot \hat{\mathbf{n}} = 0 \text{ on } \Gamma_N\}$ .

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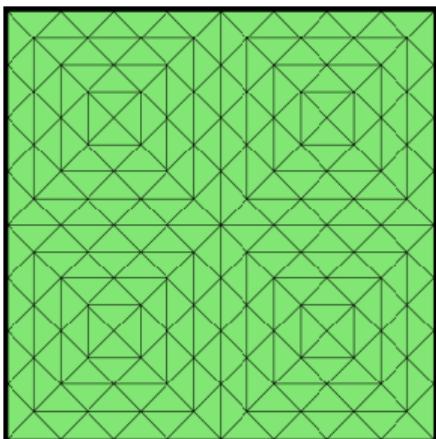
- $f \in \mathbb{L}^2(\Omega)$ ,
- $g_D \in \mathbb{H}^{1/2}(\Gamma_D)$  and  $g_N \in \mathbb{L}^2(\Gamma_N)$ ,
- $\alpha(x)$  positive **piecewise constant**.

- $\boldsymbol{\sigma} = -\alpha(x)\nabla u$  is the **flux**,
- $\mathbb{H}_N(\text{div}) = \{\boldsymbol{\tau} \in \mathbb{H}(\text{div}) : \boldsymbol{\tau} \cdot \hat{\mathbf{n}} = 0 \text{ on } \Gamma_N\}$ .

Existence theory is not substantially different, just longer to write, see (Boffi, Brezzi, and Fortin 2013, Chapter 7).

# Mixed Methods - Galerkin Approach for Poisson

To apply the **discrete version of Brezzi's Theorem**, for which we select the Brezzi-Douglas-Marini and piecewise constant elements to build our mixed space.



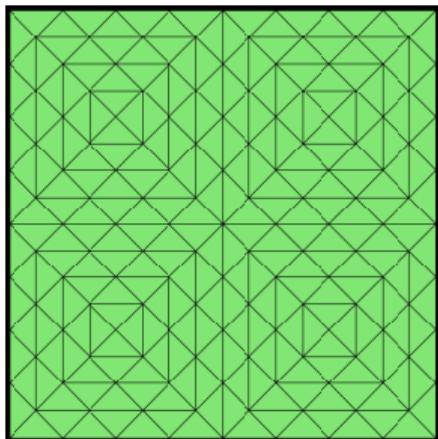
**Mesh:** shape-regular affine triangulation  $\Omega_h$

- **Mesh:**

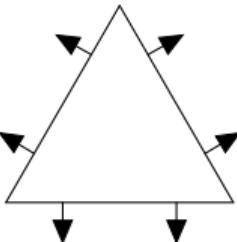
```
RefinementLevels = 2;
node = [-1 1; 0 1; 1 1; -0.5 0.5; 0.5 0.5; -1 0; 0 0; 1 0;
         ↪ -0.5 -0.5; 0.5 -0.5;-1.0 -1.0; 0.0 -1.0;1.0 -1.0];
elem =[4 2 1;4 1 6;4 6 7;4 7 2;5 3 2;5 2 7;5 7 8;5 8 3;9
         ↪ 7 6; 9 6 11;9 11 12;9 12 7;10 8 7;10 7 12;10 12 13;10
         ↪ 13 8];
bdEdge = [2 0 0;1 0 0; 0 0 0;0 0 0;2 0 0;0 0 0;0 0 0;0 0 0;1 0
         ↪ 0; 0 0 0;1 0 0;1 0 0;0 0 0;0 0 0;0 0 0;1 0 0;1 0 0];
for i=1:RefinementLevels
    [node,elem,bdEdge] = uniformrefine(node,elem,bdEdge);
end
```

# Mixed Methods - Galerkin Approach for Poisson

To apply the **discrete version of Brezzi's Theorem**, for which we select the Brezzi-Douglas-Marini and piecewise constant elements to build our mixed space.



**Mesh:** shape-regular affine triangulation  $\Omega_h$



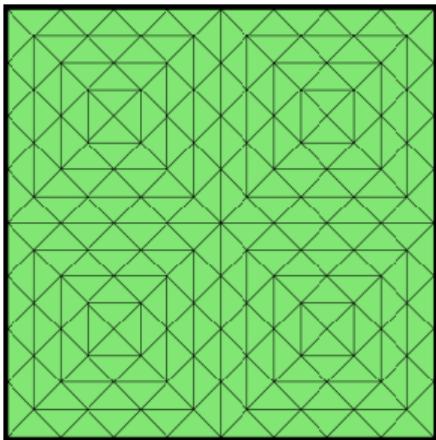
- **Mesh:**  $\Omega_h$
- The BDM<sub>1</sub> elements

In general,  $\mathbf{q} \in \text{BDM}_k = (\mathbb{P}_k)^2$ , thus  $\text{div } \mathbf{q} \in \mathbb{P}_{k-1}$ , and to complete the definition we impose the values on the normal trace  $\phi = \mathbf{q} \cdot \hat{\mathbf{n}}$  on  $\partial K$  belonging to

$$\{\phi \mid \phi \in \mathbb{L}^2(K), \phi|_{\partial\Omega} \in \mathbb{P}_k\}.$$

# Mixed Methods - Galerkin Approach for Poisson

To apply the **discrete version of Brezzi's Theorem**, for which we select the Brezzi-Douglas-Marini and piecewise constant elements to build our mixed space.



- **Mesh:**  $\Omega_h$
- The  $BDM_1$  elements
- The  $P_0$  elements

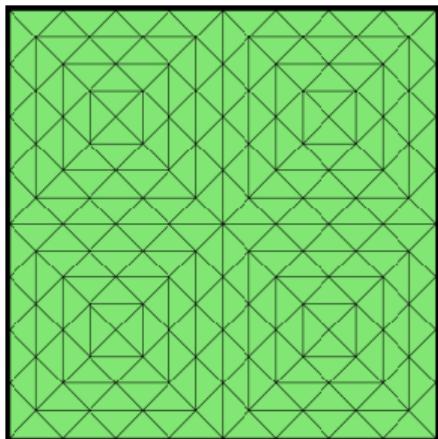
$$P_0 = \{v : v|_K \in \mathbb{P}_0(K), \quad \forall K \in \Omega_h\}.$$

**Mesh:** shape-regular affine triangulation  $\Omega_h$

# Mixed Methods - Galerkin Approach for Poisson

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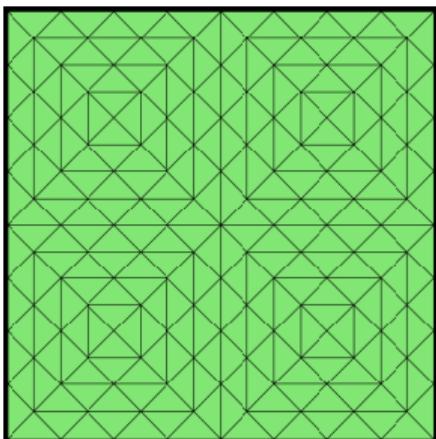


**Mesh:** shape-regular affine triangulation  $\Omega_h$

- **Mesh:**  $\Omega_h$
- The  $BDM_1$  elements  $\rightsquigarrow V_h$
- The  $P_0$  elements  $\rightsquigarrow M_h$
- For the convergence analysis see (Brezzi, Douglas, and Marini 1985, Section 3 and 4).

# Mixed Methods - Galerkin Approach for Poisson

To apply the **discrete version of Brezzi's Theorem**, for which we select the Brezzi-Douglas-Marini and piecewise constant elements to build our mixed space.



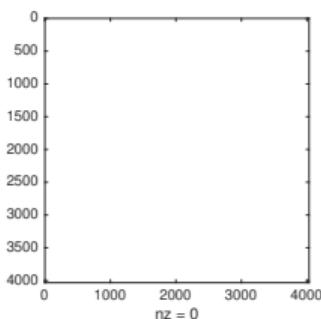
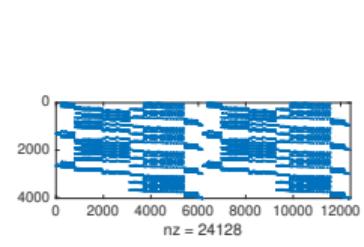
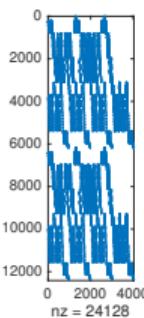
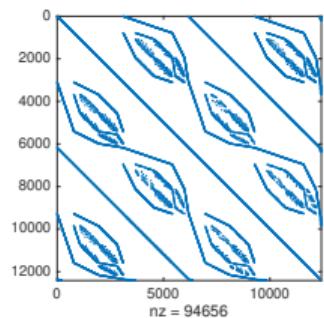
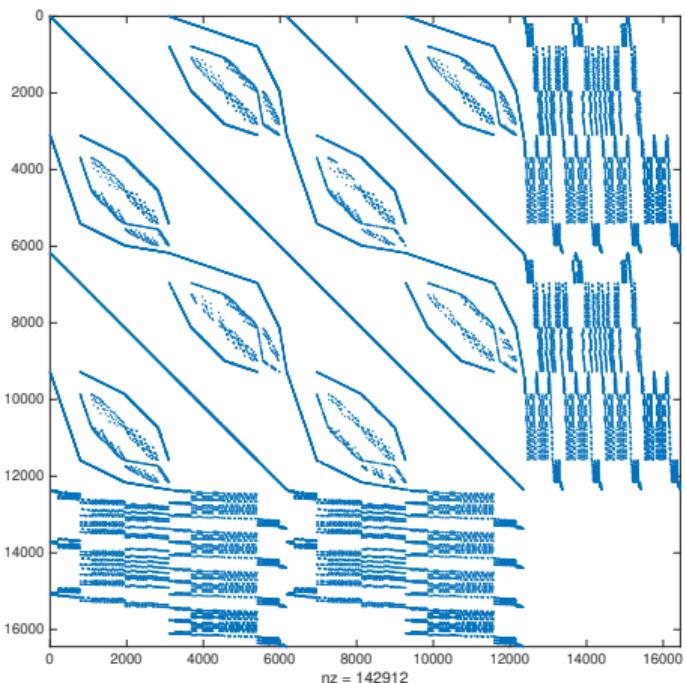
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- **Mesh:**  $\Omega_h$
- The  $BDM_1$  elements
- The  $P_0$  elements
- For the convergence analysis see (Brezzi, Douglas, and Marini 1985, Section 3 and 4).
- And look at the code for assembling the matrix

```
NT = size(elem,1); % Number of triangles  
NE = size(edge,1); % Number of edges  
sol = zeros(2*NE+NT,1); % Space to store the solution  
inva = 1./exactalpha((node(elem(:,1)) + node(elem(:,2)) +  
    ↳ node(elem(:,3)))/3);  
[a,b,area] = gradlambda(node,elem);  
M = assemblebdm(NT,NE,a,b,area,elem2edge,signedge,inva);
```

# Mixed Methods - The Saddle-Point Matrix

We can finally look at our first **saddle-point** matrix for the Poisson problem.



# Mixed Methods - Eigenvalue Bounds

One of the results you have seen in the **morning lectures** concerns eigenvalue bounds for these matrices. Let us look at it numerically.

Theorem (Rusten and Winther 1992)

Let  $\mu_1 \geq \mu_2 \geq \dots \geq \mu_n > 0$  be the eigenvalues of  $A$ ,  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_m > 0$  the singular values of  $B$ . If we denote as  $\sigma(\mathcal{A})$  the spectrum of  $\mathcal{A}$ , then

$$\sigma(\mathcal{A}) \subset I = I^- \cup I^+,$$

where

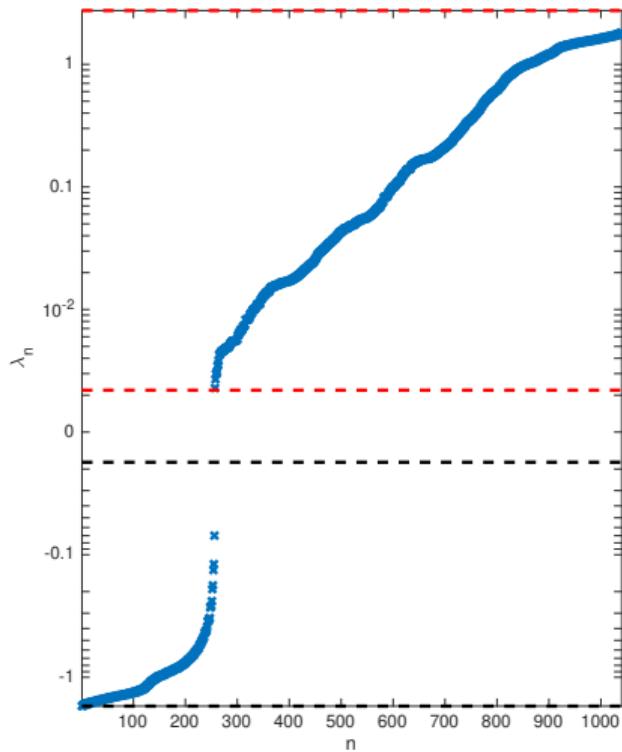
$$I^- = \left[ \frac{1}{2} \left( \mu_n - \sqrt{\mu_n^2 + 4\sigma_1^2} \right), \frac{1}{2} \left( \mu_1 - \sqrt{\mu_1^2 + 4\sigma_m^2} \right) \right],$$

$$I^+ = \left[ \mu_n, \frac{1}{2} \left( \mu_1 + \sqrt{\mu_1^2 + 4\sigma_1^2} \right) \right].$$

# Mixed Methods - Eigenvalue Bounds

We can compute the bounds with few lines of code:

```
lambda = eig(M(freeDof,freeDof));  
mun = eigs(A,1,'smallestabs');  
mu1 = eigs(A,1,'largestabs');  
sigma1 = svds(BT,1,'largest');  
sigmam = svds(BT,1,'smallest');  
  
Iminus(1) = 0.5*(mun - sqrt(mun^2+4*sigma1^2));  
Iminus(2) = 0.5*(mu1 - sqrt(mu1^2+4*sigmam^2));  
Iplus(1) = mun;  
Iplus(2) = 0.5*(mu1 + sqrt(mu1^2 +  
    4*sigma1^2));
```



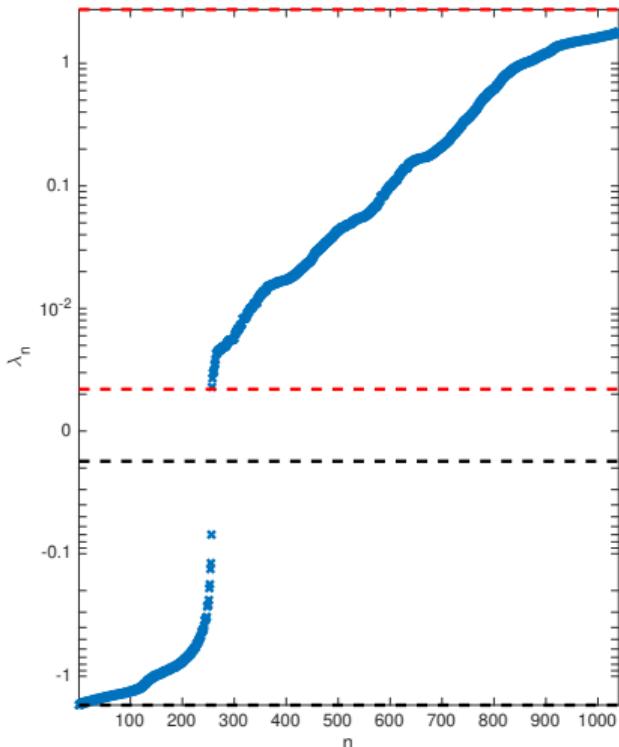
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sigmam = svds(BT,1,'smallest');

Iminus(1) = 0.5*(mun - sqrt(mun^2+4*sigma1^2));
Iminus(2) = 0.5*(mu1 - sqrt(mu1^2+4*sigmam^2));
Iplus(1) = mun;
Iplus(2) = 0.5*(mu1 + sqrt(mu1^2 +
    4*sigma1^2));
```

 **Next week**, after you become familiar with iterative methods, we will focus on **preconditioning**.



# Mixed Methods - The Stokes equation

Let us consider the **Stokes equations** for the *steady flow* of a **very viscous fluid**

$$\begin{cases} -\nabla^2 \mathbf{u} + \nabla p = \mathbf{0}, & \text{Momentum equation,} \\ \nabla \cdot \mathbf{u} = 0, & \text{Incompressibility constraint.} \end{cases}$$

- $\mathbf{u}$  is a *vector-valued function* representing the velocity of the fluid,
- $p$  is a *scalar* function representing the pressure.

## Modeling assumption

The crucial **modeling assumption** is that the flow is “low speed” we **neglect** effects due to **convection**.

# Mixed Methods - The Stokes equation

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## Why do we care?

Stokes equations represent a limiting case of the more general Navier–Stokes equations

# The Stokes equation: weak formulation

---

Let us build the **weak formulation**

$$\begin{aligned}-\nabla^2 \mathbf{u} + \nabla p &= \mathbf{0}, \\ \nabla \cdot \mathbf{u} &= 0,\end{aligned}$$

# The Stokes equation: weak formulation

---

Let us build the **weak formulation**, we select  $(\mathbf{v}, q) \in V \times M$

$$\int_{\Omega} \mathbf{v} \cdot (-\nabla^2 \mathbf{u} + \nabla p) = \mathbf{0},$$

$$\int_{\Omega} q \nabla \cdot \mathbf{u} = 0,$$

# The Stokes equation: weak formulation

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Let us build the **weak formulation**, we select  $(\mathbf{v}, q) \in V \times M$

$$\int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v} - \int_{\Omega} p \nabla \cdot \mathbf{v} - \int_{\partial\Omega} \left( \frac{\partial \mathbf{u}}{\partial n} - p \hat{\mathbf{n}} \right) \cdot \mathbf{v} = 0,$$
$$\int_{\Omega} q \nabla \cdot \mathbf{u} = 0,$$

- Here  $\nabla \mathbf{u} : \nabla \mathbf{v}$  is the **componentwise** scalar product, e.g., in dimension 2, this is  $\nabla u_x \cdot \nabla v_x + \nabla u_y \cdot \nabla v_y$

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- Here  $\nabla \mathbf{u} : \nabla \mathbf{v}$  is the **componentwise** scalar product
- We select boundary conditions  $\partial\Omega = \Gamma_N \cup \Gamma_D$   $\Gamma_D \cap \Gamma_N = \emptyset$ ,  $\mu(\Gamma_D) \neq 0$ :

$$\mathbf{u} = \mathbf{w} \text{ on } \Gamma_D, \quad \frac{\partial \mathbf{u}}{\partial n} - p \hat{\mathbf{n}} = \mathbf{s} \text{ on } \Gamma_N$$

# The Stokes equation: weak formulation

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Let us build the **weak formulation**, we select  $(\mathbf{v}, q) \in \mathbb{H}_{E_0}^1 \times \mathbb{L}^2$

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- We define the spaces

$$\mathbb{H}_E^1 = \{ \mathbf{u} \in \mathbb{H}^1(\Omega)^d \mid \mathbf{u} = \mathbf{w} \text{ on } \Gamma_D \}, \quad \mathbb{H}_{E_0}^1 = \{ \mathbf{v} \in \mathbb{H}^1(\Omega)^d \mid \mathbf{v} = 0 \text{ on } \Gamma_D \}.$$

# The Stokes equation: weak formulation

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$$\begin{aligned}\int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v} - \int_{\Omega} p \nabla \cdot \mathbf{v} - \int_{\partial\Omega} \mathbf{s} \cdot \mathbf{v} &= \mathbf{0}, \\ \int_{\Omega} q \nabla \cdot \mathbf{u} &= 0,\end{aligned}$$

- Here  $\nabla \mathbf{u} : \nabla \mathbf{v}$  is the **componentwise** scalar product
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# The Stokes equation: weak formulation

---

Let us build the **weak formulation**

Find  $(\mathbf{u}, p) \in \mathbb{H}_E^1 \times \mathbb{L}^2(\Omega)$  s.t.  $\begin{cases} \int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v} - \int_{\Omega} p \nabla \cdot \mathbf{v} - \int_{\partial\Omega} \mathbf{s} \cdot \mathbf{v} = \mathbf{0}, & \forall \mathbf{v} \in \mathbb{H}_{E_0}^1 \\ \int_{\Omega} q \nabla \cdot \mathbf{u} = 0, & \forall q \in \mathbb{L}^2. \end{cases}$

- Here  $\nabla \mathbf{u} : \nabla \mathbf{v}$  is the **componentwise** scalar product
- We select boundary conditions  $\partial\Omega = \Gamma_N \cup \Gamma_D$   $\Gamma_D \cap \Gamma_N = \emptyset$ ,  $\mu(\Gamma_D) \neq 0$ :

$$\mathbf{u} = \mathbf{w} \text{ on } \Gamma_D, \quad \frac{\partial \mathbf{u}}{\partial n} - p \hat{\mathbf{n}} = \mathbf{s} \text{ on } \Gamma_N$$

- We define the spaces

$$\mathbb{H}_E^1 = \{\mathbf{u} \in \mathbb{H}^1(\Omega)^d \mid \mathbf{u} = \mathbf{w} \text{ on } \Gamma_D\}, \quad \mathbb{H}_{E_0}^1 = \{\mathbf{v} \in \mathbb{H}^1(\Omega)^d \mid \mathbf{v} = 0 \text{ on } \Gamma_D\}.$$

# The Stokes equation: issues with BCs

$$\text{Find } (\mathbf{u}, p) \in \mathbb{H}_E^1 \times \mathbb{L}^2(\Omega) \text{ s.t. } \begin{cases} \int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v} - \int_{\Omega} p \nabla \cdot \mathbf{v} - \int_{\partial\Omega} \mathbf{s} \cdot \mathbf{v} = \mathbf{0}, & \forall \mathbf{v} \in \mathbb{H}_{E_0}^1, \\ \int_{\Omega} q \nabla \cdot \mathbf{u} = 0, & \forall q \in \mathbb{L}^2. \end{cases}$$

## Words of caution

1. For a **unique velocity solution** the Dirichlet part of the boundary has to be nontrivial,
2. If the velocity is fixed everywhere on the boundary ( $\Gamma_D \equiv \partial\Omega$ ) the pressure solution is only unique up to a constant (*hydrostatic pressure level*) and  $\mathbf{w}$  has to satisfy

$$0 = \int_{\Omega} \nabla \cdot \mathbf{u} = \int_{\partial\Omega} \mathbf{u} \cdot \hat{\mathbf{n}} = \int_{\partial\Omega} \mathbf{w} \cdot \hat{\mathbf{n}},$$

i.e., the **volume of fluid entering** the domain must be **matched** by the **volume of fluid flowing out** of it.

# The Stokes equation: Mixed Elements

As we have done for Poisson, we need to select  $V_h \subset V = \mathbb{H}_{E_0}^1$  and  $M_h \subset M = \mathbb{L}^2(\Omega)$ :

Find  $(\mathbf{u}_h, p_h) \in V_h \times M_h$  s.t.  $\begin{cases} \int_{\Omega} \nabla \mathbf{u}_h : \nabla \mathbf{v}_h - \int_{\Omega} p_h \nabla \cdot \mathbf{v}_h - \int_{\partial\Omega} \mathbf{s} \cdot \mathbf{v}_h = \mathbf{0}, & \forall \mathbf{v}_h \in V_h, \\ \int_{\Omega} q_h \nabla \cdot \mathbf{u}_h = 0, & \forall q_h \in M_h. \end{cases}$

To determine the subspaces  $V_h$  and  $M_h$  we want to apply the **Theorem (Discrete Brezzi)**

$$\min_{q_h \neq \text{const.}} \max_{\mathbf{v}_h \neq \mathbf{0}} \frac{|(q_h, \nabla \cdot \mathbf{v}_h)|}{\|\mathbf{v}_h\|_V \|q_h\|_M} \geq \beta.$$

where

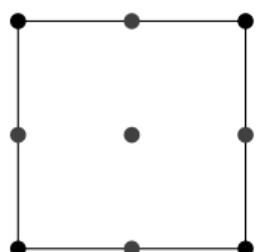
- $\|\mathbf{v}\|_V = (\int_{\Omega} \mathbf{v} \cdot \mathbf{v} + \nabla \mathbf{v} : \nabla \mathbf{v})^{\frac{1}{2}}$ ,
- $\|q\|_M = \|q - \mu(\Omega)^{-1} \int_{\Omega} q\|$ .

# The Stokes equation: Mixed Elements

## Idea for finding inf-sup stable elements

The idea is to consider “local enclosed flow Stokes problems” posed on a subdomain  $\mathcal{M} \subset \Omega$  ( $\mathbf{w} \cdot \hat{\mathbf{n}} = 0$  on  $\partial\mathcal{M}$ ) called a *macroelement* that has a topology that is regular and simple enough (so that we can actually do estimates and computations).

$Q_2$ - $Q_1$  Elements



Two velocity components

We approximate the 2 components of velocity with a single  $Q_2$  FEM space

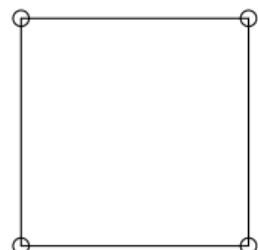
$$\{\Phi_1, \dots, \Phi_{2n}\} = \{(\phi_1, 0)^T, \dots, (\phi_n, 0)^T, (0, \phi_1)^T, \dots, (0, \phi_n)^T\}$$

# The Stokes equation: Mixed Elements

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$Q_2$ - $Q_1$  Elements



Pressure

We approximate the 2 components of velocity with a single  $Q_2$  FEM space  $\{\Phi_j\}_{j=1}^{n_u}$ . And the scalar pressure component with  $Q_1$  FEM space  $\{\psi_j\}_{j=1}^{n_p}$  giving:

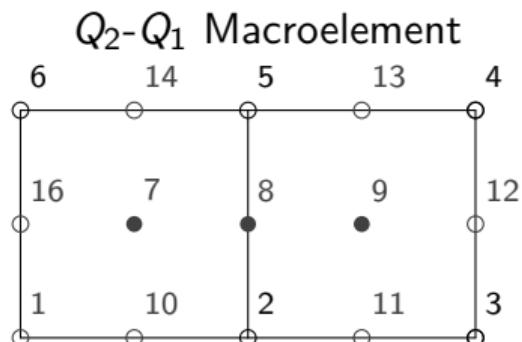
$$\mathcal{A} = \begin{bmatrix} A & O & B_x^T \\ O & A & B_y^T \\ B_x & B_y & O \end{bmatrix} \quad \begin{aligned} a_{i,j} &= \int_{\Omega} \nabla \phi_i \cdot \nabla \phi_j, \\ b_{x,ki} &= - \int_{\Omega} \psi_k \partial_x \phi_i, \\ b_{y,kj} &= - \int_{\Omega} \psi_k \partial_y \phi_j. \end{aligned}$$

Since we have an **enclosed flow**  $\ker B^T = \{1\}$ .

# The Stokes equation: Mixed Elements

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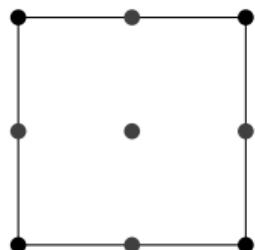
Three interior velocity nodes  
and six pressure nodes.

$B^T$  is a  $6 \times 6$  matrix, with some effort we can compute all the entries and verify that  $\ker B^T = \{\mathbf{1}\}$  (part of the computations are done in (Elman, Silvester, and Wathen 2014, Section 3.3.1)).

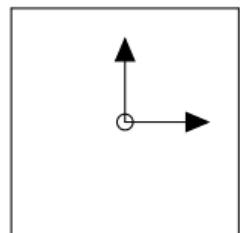
- Then stability holds for all patches of elements with the same topology,
- Any grid made of an even number of cells can be decomposed this way.

# The Stokes equation: other stable elements

$Q_2-P_{-1}$  Elements

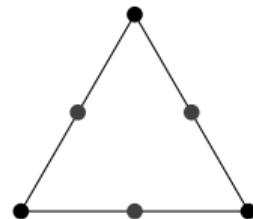


Velocity

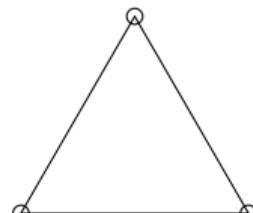


Pressure, and pressure derivatives

$P_2-P_1$  Elements

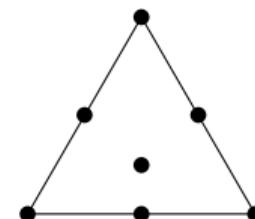


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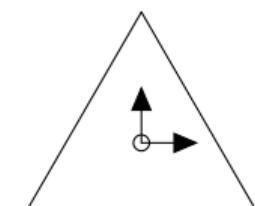


Pressure

$P_{2*}-P_{-1}$  Element



Velocity

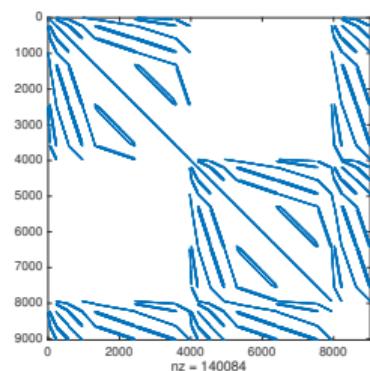


Pressure, and pressure derivatives

# The Stokes equation: the associated saddles

The  $P_2-P_1$  (Taylor-Hood) case for the **colliding flow** test problem.

- $\Omega = [-1, 1] \times [-1, 1]$
- $u_x = 20xy^3, u_y = 5x^4 - 5y^4,$   
 $p = 60x^2y - 20y^3 + c,$
- Dirichlet boundary condition on all the square  
 $\psi(x, y) = 5xy^4 - x^5.$

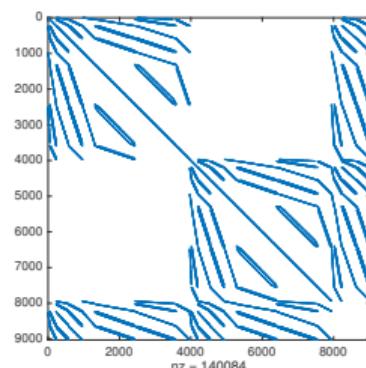


```
% Building the mesh
RefinementLevels = 2;
square = [0,1,0,1];
h = 0.25;
[node,elem] = squaremesh(square,h);
for i=1:RefinementLevels
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end
% Building the test problem: colliding flows
bdFlag = setboundary(node,elem,'Dirichlet');
pde = Stokesdata1;
options.solver='none'; % We just perform the build
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```

Question:

Can we apply (Rosten and Winther 1992)?

# The Stokes equation: properties of the matrix

Theorem (Elman, Silvester, and Wathen 2014, Theorem 3.2.1)

With  $P_1$ ,  $P_2$ ,  $Q_1$  or  $Q_2$  approximation on a *shape-regular, quasi-uniform* subdivision of  $\mathbb{R}^2$ , the matrix  $A$  for the *discrete vector Laplacian* satisfies

$$ch^2 \leq \frac{\langle A\mathbf{v}, \mathbf{v} \rangle}{\langle \mathbf{v}, \mathbf{v} \rangle} \leq C \quad \forall \mathbf{v} \in \mathbb{R}^{n_u},$$

where  $h$  is the length of the longest edge in the mesh, and  $c$  and  $C$  are **constants independent** of  $h$ .

- This gives us information the behavior of the **smallest and largest eigenvalue** of  $A$  (*Rayleigh Principle*)!

$h^2$	$\lambda_{\min}(A)$	$h^2/\lambda_{\min}(A)$	$\lambda_{\max}(A)$
0.0156	0.0768	0.2035	10.5391
0.0039	0.0193	0.2028	10.6346
0.0010	0.0048	0.2027	10.6586
0.0002	0.0012	0.2027	10.6647

# The Stokes equation: properties of the matrix

---

To uncover information on the  $B$  matrices, we need to introduce a **discrete representation of the norm of  $M_h \subset \mathbb{L}^2$** :

$$p_h \in M_h : \|p_h\| = \langle Qp_h, p_h \rangle^{1/2}, \quad Q = (q_{kl}), q_{k,l} = \int_{\Omega} \psi_k \psi_l, \quad k, l = 1, \dots, n_p.$$

- ! The matrix  $Q$  is called **mass matrix** for the pressure space, *in general*, we call mass-matrices all the matrices obtained in this way for the basis of a given FEM space.

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## Generalized singular values

We call **generalized singular values** the **real** numbers  $\sigma$  associated with the following generalized eigenvalue problem

$$\begin{bmatrix} O & B^T \\ B & O \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ \mathbf{q} \end{bmatrix} = \sigma \begin{bmatrix} A & O \\ O & Q \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ \mathbf{q} \end{bmatrix}.$$

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$\sigma \neq 0$  we select vector  $(\mathbf{v}^T, -\mathbf{q}^T)^T$  and obtain

$$\begin{aligned} <\mathbf{v}, B^T \mathbf{q}> - <\mathbf{q}, B\mathbf{v}> &= 0 = \sigma (<\mathbf{v}, A\mathbf{v}> - <\mathbf{q}, Q\mathbf{q}>) \\ \Rightarrow < A\mathbf{v}, \mathbf{v}> &= < Q\mathbf{q}, \mathbf{q}>. \end{aligned}$$

That is

$$\frac{< BA^{-1}B^T \mathbf{q}, \mathbf{q}>}{< Q\mathbf{q}, \mathbf{q}>} = \sigma^2 = \frac{< B^T Q^{-1} B\mathbf{v}, \mathbf{v}>}{< A\mathbf{v}, \mathbf{v}>}.$$

# The Stokes equation: properties of the matrix

---

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- If  $\ker B^T = \mathbf{0}$  then  $B$  has  $n_p$  positive singular values.
- This is linked to the inf-sup condition:

$$\begin{aligned}\beta &\leq \min_{q_h \neq \text{const}} \max_{\mathbf{v}_h \neq \mathbf{0}} \frac{|(q_h, \nabla \cdot \mathbf{v}_h)|}{\|\nabla \mathbf{v}_h\| \|q_h\|} = \min_{\mathbf{q} \neq \mathbf{1}} \max_{\mathbf{v} \neq \mathbf{0}} \frac{|\langle \mathbf{q}, B\mathbf{v} \rangle|}{\langle A\mathbf{v}, \mathbf{v} \rangle^{1/2} \langle Q\mathbf{q}, \mathbf{q} \rangle^{1/2}} \\ &= \min_{\mathbf{q} \neq \mathbf{1}} \frac{1}{\langle Q\mathbf{q}, \mathbf{q} \rangle^{1/2}} \max_{\mathbf{w} = A^{1/2}\mathbf{v} \neq \mathbf{0}} \frac{|\langle \mathbf{q}, BA^{-1/2}\mathbf{w} \rangle|}{\langle \mathbf{w}, \mathbf{w} \rangle^{1/2}} \\ &= \min_{\mathbf{q} \neq \mathbf{1}} \frac{\langle A^{-1/2}B^T \mathbf{q}, A^{-1/2}B^T \mathbf{q} \rangle^{1/2}}{\langle Q\mathbf{q}, \mathbf{q} \rangle^{1/2}} = \min_{\mathbf{q} \neq \mathbf{1}} \frac{\langle BA^{-1}B^T \mathbf{q}, \mathbf{q} \rangle^{1/2}}{\langle Q\mathbf{q}, \mathbf{q} \rangle^{1/2}} = \sigma_{\min}\end{aligned}$$

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# The Stokes equation: properties of the matrix

Theorem (Elman, Silvester, and Wathen 2014, Theorem 3.22)

Let  $\partial\Omega \equiv \Gamma_D$ , the Stokes problem discretized with a uniformly stable mixed approximation on a shape-regular, quasi-uniform subdivision of  $\mathbb{R}^2$ , has a **Schur complement matrix**  $BA^{-1}B^T$  that is spectrally equivalent to the pressure mass matrix  $Q$ :

$$\beta^2 \leq \frac{\langle BA^{-1}B^T \mathbf{q}, \mathbf{q} \rangle}{\langle Q \mathbf{q}, \mathbf{q} \rangle} \leq 1, \quad \forall \mathbf{q} \in \mathbb{R}^{n_p} : \mathbf{q} \neq \mathbf{1}.$$

The inf-sup constant  $\beta$  is bounded away from zero independently of  $h$  and the condition number (discarding the zero eigenvalue)  $\kappa^e(BA^{-1}B^T) \leq C/(c\beta)^2$  for  $c$  and  $C$  given by

$$ch^2 \leq \frac{\langle Q \mathbf{q}, \mathbf{q} \rangle}{\langle \mathbf{q}, \mathbf{q} \rangle} \leq Ch^2, \quad \forall \mathbf{q} \in \mathbb{R}^{n_p}.$$

# The Stokes equation: properties of the matrix

We can **run this test** on the usual test problem  
by running the code in the folder

`</> E3-Stokes/stokesmatrixproperties.m`

This tests both:

1. The bound on the vector Laplacian,
2. The bounds on the Schur complement.

for the  $P_2-P_1$  elements.

$h$	$\lambda_2$	$\lambda_{n_p}$
0.2500	0.1352	0.9932
0.1250	0.1341	0.9996
0.0625	0.1336	1.0000
0.0312	0.1334	1.0000

Generalized eigenvalues for:

$$BA^{-1}B^T \mathbf{x} = \lambda Q \mathbf{x}$$

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## Why do we care?

As you will see in the following, these information are **useful** for the **design of iterative solvers**.

# The Stokes equation: stabilized discretizations

---

We have seen that the matter of obtain a **stable discretization** depends on the null-space of  $B^T$ .

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## Idea behind stabilization

If the discretization is not stable  $\exists \mathbf{p} \neq \mathbf{0}$  such that  $B^T \mathbf{p} = \mathbf{0}$ , that is  $(\mathbf{0}^T, \mathbf{p}^T)^T$  is a null vector for the homogeneous saddle-point system. The **idea** behind stabilization is **relaxing the incompressibility constraint** so that this vector is no longer in the kernel **and** we still obtain a reasonable error bound for the convergence of the method.

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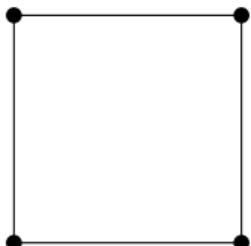
## Technique

The **technique** to devise stabilization is again using *macroelements*.

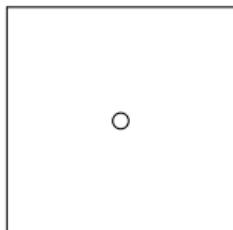
# The Stokes equation: stabilized $Q_1-P_0$

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- This is the simplest *unstable* element,



Velocity

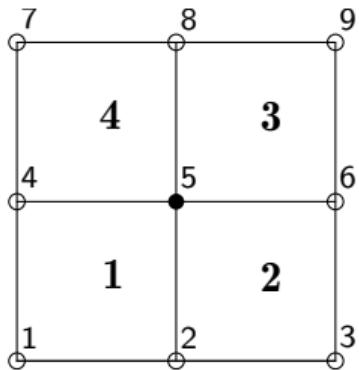


Pressure

# The Stokes equation: stabilized $Q_1-P_0$

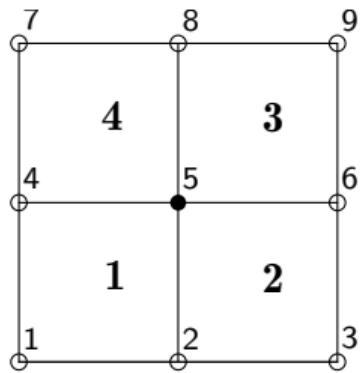
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# The Stokes equation: stabilized $Q_1-P_0$

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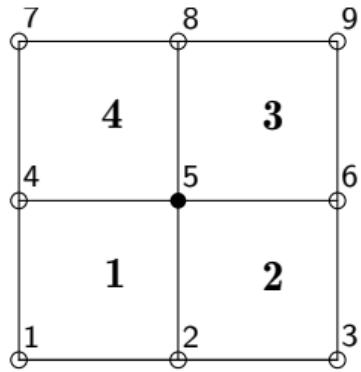
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- The pressure coefficient  $\mathbf{p}$  solves

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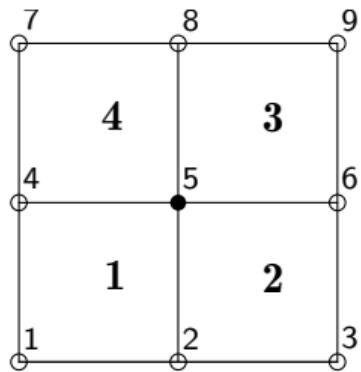
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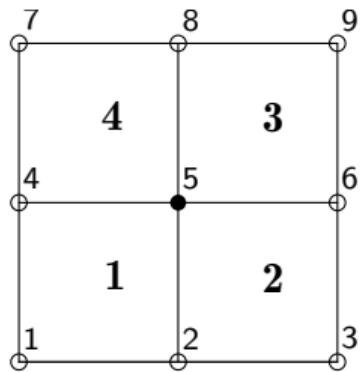
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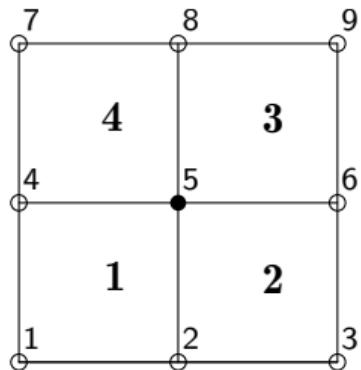
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- What about  $\gamma$ ?

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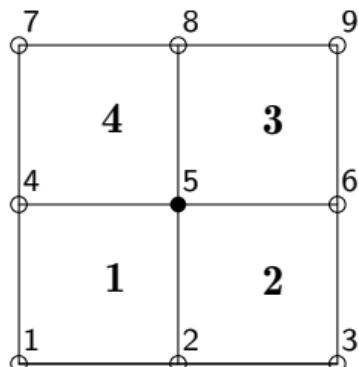


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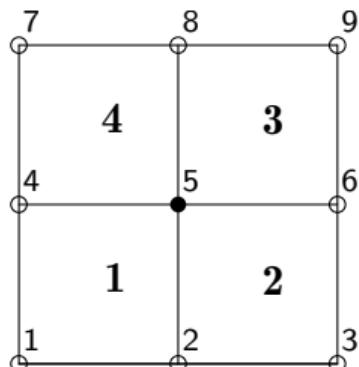
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- Complete stabilization matrix:  $C = \text{blockdiag}(C_*, \dots, C_*)$ .

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Other stabilization are possible

This is not the only possible stabilization matrix, other choices are possible, consider, e.g.,

$$C^* = h_x h_y \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ -1 & 0 & -1 & 2 \end{bmatrix}$$

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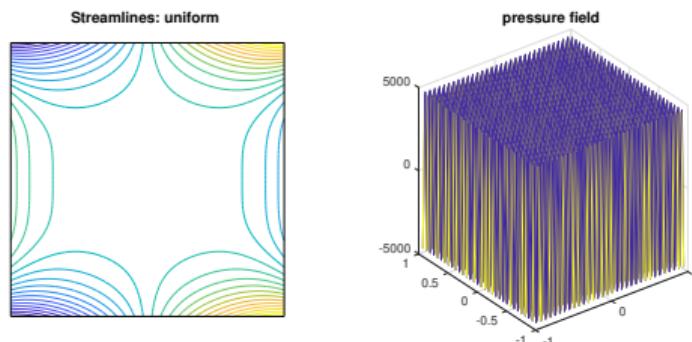
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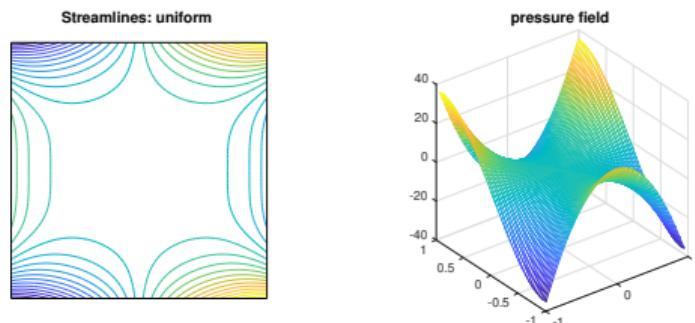
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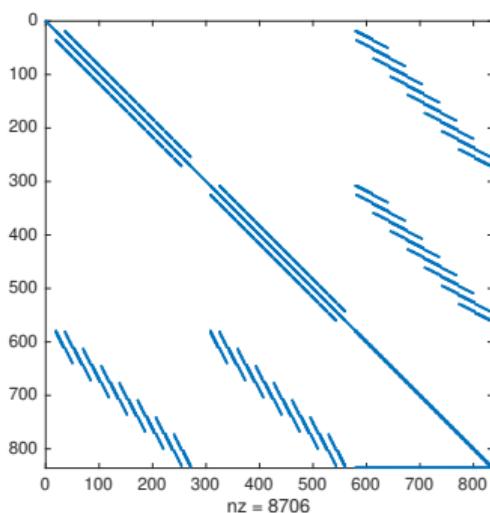
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$$\gamma = -1/4$$



# The Stokes equation: stabilized matrix properties

What can we say about the **spectral properties** of the **stabilized matrices**?



- To substitute the inf-sup condition we introduce the operator:

$$s(q_h) : M_h \rightarrow \mathbb{R}$$

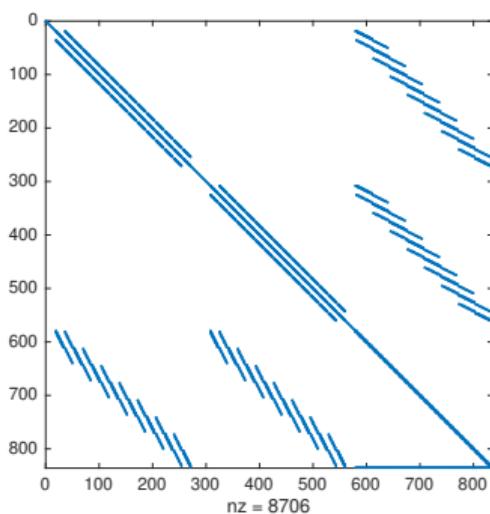
$$q_h \mapsto s(q_h) = \max_{\mathbf{v}_h \neq 0} \frac{(q_h, \nabla \cdot \mathbf{v}_h)}{\|\nabla \mathbf{v}_h\|} + c(q_h, q_h)^{1/2},$$

where  $c(\cdot, \cdot)$  is the stabilization operator that generates the matrix  $C$ .

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## Uniform stabilization

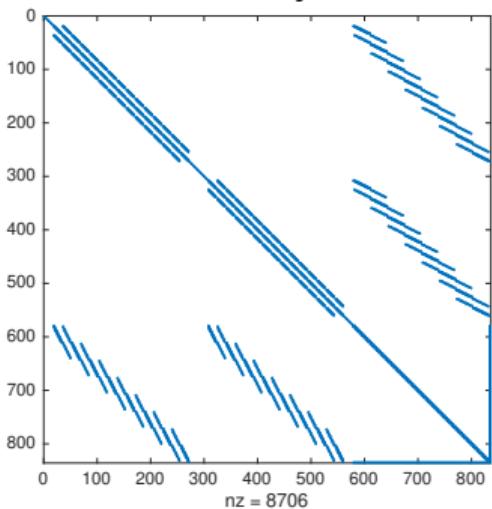
The Stokes problem is said to be **uniformly stabilized** if there exists  $\beta$  **independent of  $h$**  such that

$$\begin{bmatrix} A & B^T \\ B & -C \end{bmatrix} \begin{bmatrix} \mathbf{u} \\ \mathbf{p} \end{bmatrix} = \begin{bmatrix} \mathbf{f} \\ \mathbf{g} \end{bmatrix}$$

$$s(q_h) \geq \beta^2 \|q_h\|, \quad \forall q_h \in M_h.$$

# The Stokes equation: stabilized matrix properties

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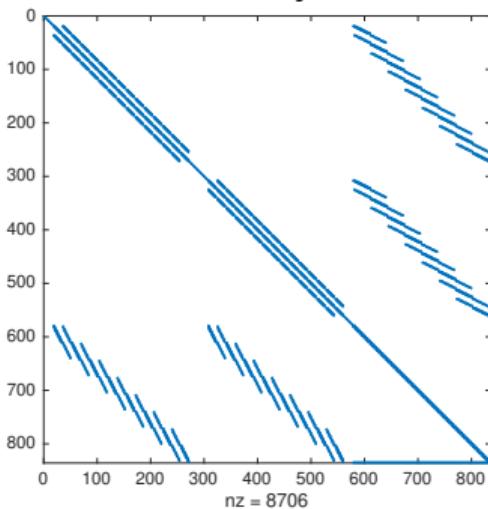
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- Then the generalized inf-sup conditions is

$$\beta^2 = 2 \min_{\mathbf{q} \neq 0} \frac{\langle (BA^{-1}B^T + C)\mathbf{q}, \mathbf{q} \rangle}{\langle Q\mathbf{q}, \mathbf{q} \rangle}.$$

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# The Stokes equation: stabilized matrix properties

Theorem (Elman, Silvester, and Wathen 2014, Theorem 3.29)

Let  $\partial\Omega \equiv \Gamma_D$ , the Stokes problem discretized with an *ideally* stabilized mixed approximation on a shape-regular, quasi-uniform subdivision of  $\mathbb{R}^2$ , has a **Schur complement matrix**  $BA^{-1}B^T + C$  that is spectrally equivalent to the pressure mass matrix  $Q$ :

$$\beta^2 \leq \frac{\langle (BA^{-1}B^T + C)\mathbf{q}, \mathbf{q} \rangle}{\langle Q\mathbf{q}, \mathbf{q} \rangle} \leq 2, \quad \forall \mathbf{q} \in \mathbb{R}^{n_p} : \mathbf{q} \neq \mathbf{1}.$$

The *generalized* inf-sup constant  $\beta$  is bounded away from zero independently of  $h$ .

$\ell(h = 2^{-\ell})$	$\beta^2$	$\lambda_{n_p}$
3	0.280929	1.7238
4	0.252201	1.74406
5	0.233876	1.74859
6	0.221837	1.74965

The **colliding flow** problem can be tested with

«/» E3-Stokes/stokesmatrixpropertiesstab.m

# The Navier-Stokes Equation

---

We add to the Stokes problem a **forcing term** and a **convection term** obtaining

$$\begin{aligned}-\nu \nabla^2 \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} + \nabla p &= \mathbf{f}, \\ \nabla \cdot \mathbf{u} &= 0,\end{aligned}$$

where

- $\nu > 0$  is the *kinematic viscosity*,
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$$\partial\Omega \equiv \Gamma_D$$

If the velocity is specified everywhere on the boundary, then the pressure solution to the Navier-Stokes problem is only unique up to a *hydrostatic constant*.

# The Navier-Stokes Equation: normalization

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We normalize the system

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to better highlight if the system is **diffusion dominated** or **advection dominated**.

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$$\begin{aligned} -\frac{1}{\mathcal{R}} \nabla^2 \mathbf{u}_* + \mathbf{u}_* \cdot \nabla \mathbf{u}_* + \nabla p_* &= \frac{L}{U^2} \mathbf{f}, \quad \mathcal{R} = UL/\nu \\ \nabla \cdot \mathbf{u}_* &= 0, \end{aligned}$$

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## Reynolds number

We call  $\mathcal{R} = UL/\nu$  the Reynolds number. If  $\mathcal{R} \leq 1$  then the problem is diffusion dominated, for increasing values of  $\mathcal{R}$  we get instead convection dominated problems.

# The Navier-Stokes Equation: weak formulation

Can be written *similarly to the Stokes problem*

$$\text{Find } (\mathbf{u}, p) \in V \times M : \begin{cases} \nu \int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v} + \int_{\Omega} (\mathbf{u} \cdot \nabla \mathbf{u}) \cdot \mathbf{v} - \int_{\Omega} p(\nabla \cdot \mathbf{v}) = \int_{\Omega} \mathbf{f} \cdot \mathbf{v}, \\ \int_{\Omega} q(\nabla \cdot \mathbf{u}) = 0. \end{cases}$$

We need again the **suitable spaces**

- $\mathbf{u} \in \mathbb{H}_E^1 = \{\mathbf{u} \in \mathbb{H}^1(\Omega)^d \mid \mathbf{u} = \mathbf{w} \text{ on } \Gamma_D\} \equiv V, p \in \mathbb{L}^2(\Omega) \equiv M,$
- $\mathbf{v} \in \mathbb{H}_{E_0}^1 = \{\mathbf{v} \in \mathbb{H}^1(\Omega)^d \mid \mathbf{v} = 0 \text{ on } \Gamma_D\},$
- New addition is a **trilinear form** for the **velocity term**:

$$c : \mathbb{H}_{E_0}^1 \times \mathbb{H}_{E_0}^1 \times \mathbb{H}_{E_0}^1 \rightarrow \mathbb{R}$$
$$(\mathbf{z}, \mathbf{u}, \mathbf{v}) \mapsto c(\mathbf{z}, \mathbf{u}, \mathbf{v}) = \int_{\Omega} (\mathbf{z} \cdot \nabla \mathbf{u}) \cdot \mathbf{v}.$$

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To **simplify the proof** we restrict to the case  $\partial\Omega \equiv \Gamma_D$  and  $w = 0$ , that is, a fluid confined into a fixed domain  $\Omega$ , by this choice  $V = \mathbb{H}_E^1 \equiv \mathbb{H}_{E_0}^1 \equiv \mathbb{H}_0^1(\Omega)^d$ .

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- We restate the problem

$$\text{Find } (u, p) \in V \times M : \begin{cases} a(\mathbf{u}, \mathbf{v}) + c(\mathbf{u}, \mathbf{u}, \mathbf{v}) + b(\mathbf{v}, p) = (\mathbf{f}, \mathbf{v}), & \forall \mathbf{v} \in V, \\ b(\mathbf{u}, q) = 0, & \forall q \in M, \end{cases} \quad (\text{NS})$$

with

$$a(\mathbf{w}, \mathbf{v}) = \nu(\nabla \mathbf{w}, \nabla \mathbf{v}) : V \times V \rightarrow \mathbb{R}, \quad b : V \times Q \rightarrow \mathbb{R} = -(q, \nabla \cdot \mathbf{v}),$$

$$c(\mathbf{u}, \mathbf{u}, \mathbf{v}) = \sum_{i,j=1}^d \left( u_j \frac{\partial u_i}{\partial x_j}, v_i \right) : V \times V \times V \rightarrow \mathbb{R}$$

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The existence proof then follows in few steps.

1. We consider the problem on the space  $V_{\text{div}} = \{\mathbf{v} \in \mathbb{H}^1(\Omega)^d : \operatorname{div} \mathbf{v} = 0\}$ , then a solution of (NS) is a solution also of the problem on this space

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Lemma (Quarteroni and Valli 1994, Lemma 10.1.1)

Let  $\mathbf{u}$  be a solution of  $(\text{NS}_{\text{div}})$ . Then there exists a unique  $p \in M$  such that  $(\mathbf{u}, p)$  is a solution of problem (NS).

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# The Navier-Stokes Equation: existence

Theorem (Quarteroni and Valli 1994, Theorem 10.1.1)

Let  $\mathbf{f} \in \mathbb{H}_{\text{div}} = \{\mathbf{v} \in \mathbb{L}^2(\Omega)^d \mid \operatorname{div} \mathbf{v} = 0 \text{ in } \Omega, \mathbf{v} \cdot \hat{\mathbf{n}} = 0 \text{ on } \partial\Omega\}$ , with

$$\|\mathbf{f}\| < \frac{\nu^2}{\hat{C} C_\Omega^{1/2}},$$

where  $\hat{C} > 0$  is the *continuity constant* for the trilinear form  $c$ , i.e.,

$$|c(\mathbf{w}, \mathbf{z}, \mathbf{v})| \leq \hat{C} |\mathbf{w}|_1 |\mathbf{z}|_1 |\mathbf{v}|_1 \quad \forall \mathbf{w}, \mathbf{z}, \mathbf{v} \in \mathbb{H}_0^1(\Omega)^d,$$

and  $C_\Omega$  is Poincaré constant for the domain under consideration. Then, there exist a unique solution  $\mathbf{u} \in V_{\text{div}}$  to  $(\text{NS}_{\text{div}})$ .

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and  $C_{\Omega}$  is Poincaré constant for the domain under consideration. Then, there exist a unique solution  $\mathbf{u} \in V_{\text{div}}$  to  $(\text{NS}_{\text{div}})$ .

## Idea of the proof.

1. Use Lax-Milgram for problem  $\mathcal{A}_{\mathbf{w}}(\mathbf{z}, \mathbf{v}) = (\mathbf{f}, \mathbf{v})$ ,  $\forall, \mathbf{v} \in V_{\text{div}}$  and  $\mathcal{A}_{\mathbf{w}}(\mathbf{z}, \mathbf{v}) = a(\mathbf{z}, \mathbf{v}) + c(\mathbf{w}, \mathbf{z}, \mathbf{v})$  to prove existence for every  $\mathbf{w}$ .

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and  $C_{\Omega}$  is Poincaré constant for the domain under consideration. Then, there exist a unique solution  $\mathbf{u} \in V_{\text{div}}$  to  $(\text{NS}_{\text{div}})$ .

## Idea of the proof.

2. The solution we look for is then a fixed point of the map  $\Phi : \mathbf{w} \rightarrow \mathbf{z}$ . First we prove that such solution is in a ball in  $V_{\text{div}}$ .

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and  $C_{\Omega}$  is Poincaré constant for the domain under consideration. Then, there exist a unique solution  $\mathbf{u} \in V_{\text{div}}$  to  $(\text{NS}_{\text{div}})$ .

## Idea of the proof.

3. Finally, we apply *Banach contraction Theorem* (using the hypothesis on  $\mathbf{f}$ ) to prove that there exist a unique fixed point for the problem.

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where  $\hat{C} > 0$  is the *continuity constant* for the trilinear form  $c$  and  $C_\Omega$  is Poincaré constant for the domain under consideration. Then, there exist a unique solution  $\mathbf{u} \in V_{\text{div}}$  to  $(\text{NS}_{\text{div}})$ .

## Conditions - 1

It is not restrictive to assume  $\mathbf{f} \in \mathbb{H}_{\text{div}}$ , any  $\mathbf{f} \in \mathbb{L}^2(\Omega)^d$  can be decomposed as the sum of a function in  $\mathbb{H}_{\text{div}}$  and a function that is a gradient of an  $\mathbb{H}^1(\Omega)$  function. The gradient component of the external force field  $\mathbf{f}$  doesn't play a role,  $(\mathbf{v}, \nabla q) = 0 \forall q \in \mathbb{H}^1$  and  $\mathbf{v} \in \mathbb{H}_{\text{div}}$ .

# The Navier-Stokes Equation: existence

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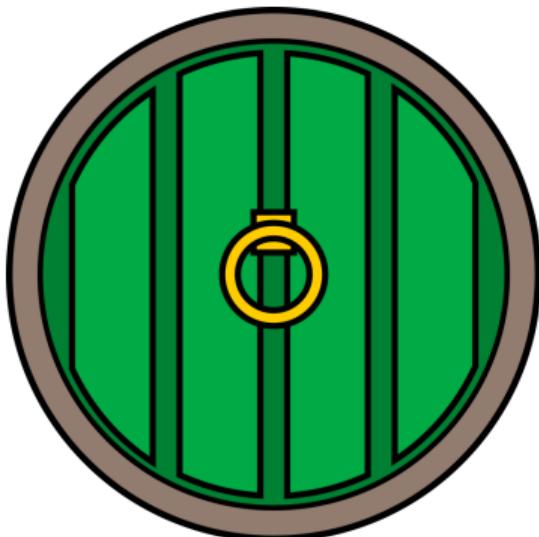
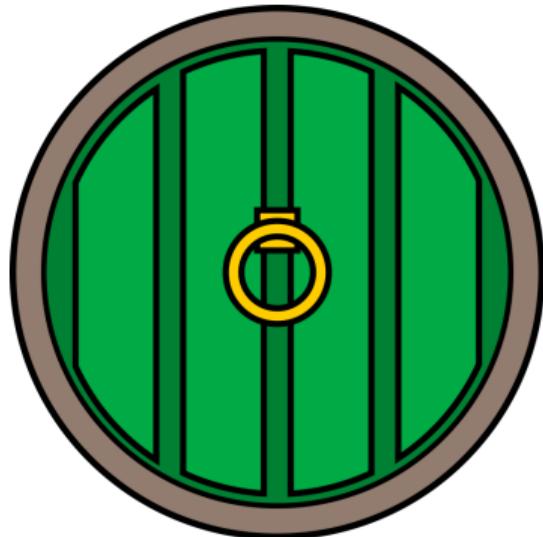
## Conditions - 2

The smallness condition on the viscosity  $\nu$  is necessary for proving uniqueness, and is restrictive. The solution may not be unique when  $\nu$  is small w.r.t.  $\mathbf{f}$ , even for reasonable  $\mathbf{f}$ .

# The Navier-Stokes Equation: linearizations

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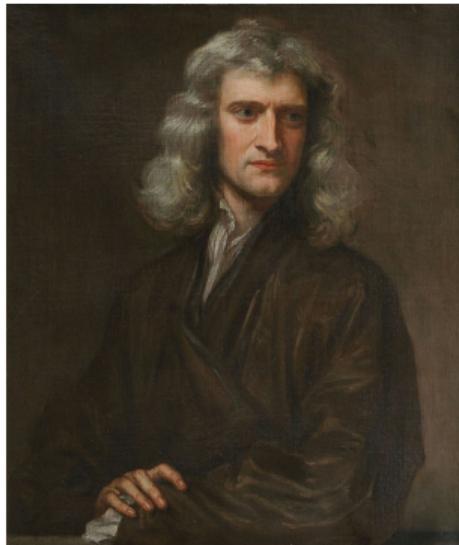
Since we only now how to solve linear problems, to face (NS) we discuss two types of **nonlinear iteration** with a **linearized problem being solved at every step**.



# The Navier-Stokes Equation: linearizations

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Newton Method



Picard's Iteration

# The Navier-Stokes Equation: linearizations

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Since we only now how to solve linear problems, to face (NS) we discuss two types of **nonlinear iteration** with a **linearized problem being solved at every step**.



Newton Method

We introduce both method first in the continuous context.



Picard's Iteration

# The Navier-Stokes Equation: Newton method

1. We have a guess  $\{\mathbf{u}_k, p_k\}$  for the solution,
2. We compute the residual pairs

$$\begin{bmatrix} R_k \\ r_k \end{bmatrix} = \begin{bmatrix} \int_{\Omega} \mathbf{f} \cdot \mathbf{v} - c(\mathbf{u}_k, \mathbf{u}_k, \mathbf{v}) - \nu \int_{\Omega} \nabla \mathbf{u}_k : \nabla \mathbf{v} + \int_{\Omega} p_k (\nabla \cdot \mathbf{v}) \\ - \int_{\Omega} q (\nabla \cdot \mathbf{u}_k) \end{bmatrix} \quad \begin{aligned} \mathbf{v} &\in \mathbb{H}_{E_0}^1, \\ q &\in \mathbb{L}^2(\Omega). \end{aligned}$$

3. Then update the solution as

$$\mathbf{u}_{k+1} = \mathbf{u}_k + \delta \mathbf{u}_k, \quad p_{k+1} = p_k + \delta p_k,$$

for  $\delta \mathbf{u}_k \in \mathbb{H}_{E_0}^1$  and  $\delta p_k \in \mathbb{L}^2(\Omega)$  the solution of

$$\begin{cases} c(\delta \mathbf{u}_k, \mathbf{u}_k, \mathbf{v}) + c(\mathbf{u}_k, \delta \mathbf{u}_k, \mathbf{v}) + \nu \int_{\Omega} \nabla \delta \mathbf{u}_k : \nabla \mathbf{v} - \int_{\Omega} \delta p_k (\nabla \cdot \mathbf{v}) = R_k, & \forall \mathbf{v} \in \mathbb{H}_{E_0}^1, \\ \int_{\Omega} q (\nabla \cdot \delta \mathbf{u}_k) = r_k, & \forall q \in \mathbb{L}^2 \end{cases}$$

# The Navier-Stokes Equation: Discrete Newton

As we have done for the **Stokes problem** we select  $V_h \subset \mathbb{H}_{E_0}^1$  and  $M_h \subset \mathbb{L}^2(\Omega)$ ,

- The Newton **updates** are then computed by solving  $\forall \mathbf{v} \in V_h, \forall q_h \in M_h$

$$\begin{cases} c(\delta \mathbf{u}_h^{(k)}, \mathbf{u}_h^{(k)}, \mathbf{v}_h) + c(\mathbf{u}_h^{(k)}, \delta \mathbf{u}_h^{(k)}, \mathbf{v}) + \nu \int_{\Omega} \nabla \delta \mathbf{u}_h^{(k)} : \nabla \mathbf{v}_h - \int_{\Omega} \delta p_h^{(k)} (\nabla \cdot \mathbf{v}_h) = R_h^{(k)} \\ \int_{\Omega} q_h (\nabla \cdot \delta \mathbf{u}_h^{(k)}) = r_h^{(k)} \end{cases}$$

where  $R_k(\mathbf{v}_h)$ , and  $r_k(q_h)$  are the nonlinear residuals w.r.t. discrete formulation.

- Selecting basis  $V_h = \text{Span}\{\phi_j\}$ ,  $M_h = \text{Span}\{\psi_j\}$  and representing (dropping the  $k$ )

$$\mathbf{u}_h = \sum_{j=1}^{n_u} \mathbf{u}_j \phi_j + \sum_{n_u+1}^{n_u+n_\partial} \mathbf{u}_j \phi_j, \quad p_h = \sum_{k=1}^{n_p} \mathbf{p}_k \psi_k,$$

and

$$\delta \mathbf{u}_h \sum_{j=1}^{n_u} + \delta \mathbf{u}_j \phi_j, \quad \delta p_h = \sum_{k=1}^{n_p} \delta \mathbf{p}_k \psi_k,$$

we get the corresponding discrete system

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- we get the corresponding discrete system

$$\mathcal{A}\delta = \begin{bmatrix} \nu \mathbf{A} + \mathbf{N} + \mathbf{W} & \mathbf{B}^T \\ \mathbf{B} & \mathbf{O} \end{bmatrix} \begin{bmatrix} \delta \mathbf{u} \\ \delta \mathbf{p} \end{bmatrix} = \begin{bmatrix} \mathbf{f} \\ \mathbf{g} \end{bmatrix}$$

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$$\mathcal{A}\delta = \begin{bmatrix} \nu \mathbf{A} + \mathbf{N} + \mathbf{W} & \mathbf{B}^T \\ \mathbf{B} & -\nu^{-1} \mathbf{C} \end{bmatrix} \begin{bmatrix} \delta \mathbf{u} \\ \delta \mathbf{p} \end{bmatrix} = \begin{bmatrix} \mathbf{f} \\ \mathbf{g} \end{bmatrix}$$

- If we use *unstable* elements, we need a stabilization matrix.

# The Navier-Stokes Equation: Picard's Iteration

The second approach for linearization is **Picard's iteration**, we start again from

1. We have a guess  $\{\mathbf{u}_k, p_k\}$  for the solution,
2. We compute the residual pairs

$$\begin{bmatrix} R_k \\ r_k \end{bmatrix} = \begin{bmatrix} \int_{\Omega} \mathbf{f} \cdot \mathbf{v} - c(\mathbf{u}_k, \mathbf{u}_k, \mathbf{v}) - \nu \int_{\Omega} \nabla \mathbf{u}_k : \nabla \mathbf{v} + \int_{\Omega} p_k (\nabla \cdot \mathbf{v}) \\ - \int_{\Omega} q (\nabla \cdot \mathbf{u}_k) \end{bmatrix} \quad \begin{array}{l} \mathbf{v} \in \mathbb{H}_{E_0}^1, \\ q \in \mathbb{L}^2(\Omega). \end{array}$$

3. Then update the solution as

$$\mathbf{u}_{k+1} = \mathbf{u}_k + \delta \mathbf{u}_k, \quad p_{k+1} = p_k + \delta p_k,$$

for  $\delta \mathbf{u}_k \in \mathbb{H}_{E_0}^1$  and  $\delta p_k \in \mathbb{L}^2(\Omega)$  the solution of

$$\begin{cases} c(\delta \mathbf{u}_k, \mathbf{u}_k, \mathbf{v}) + c(\mathbf{u}_k, \delta \mathbf{u}_k, \mathbf{v}) + \nu \int_{\Omega} \nabla \delta \mathbf{u}_k : \nabla \mathbf{v} - \int_{\Omega} \delta p_k (\nabla \cdot \mathbf{v}) = R_k, & \forall \mathbf{v} \in \mathbb{H}_{E_0}^1, \\ \int_{\Omega} q (\nabla \cdot \delta \mathbf{u}_k) = r_k, & \forall q \in \mathbb{L}^2 \end{cases}$$

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for  $\delta \mathbf{u}_k \in \mathbb{H}_{E_0}^1$  and  $\delta p_k \in \mathbb{L}^2(\Omega)$  the solution of the **Oseen system**

$$\begin{cases} c(\mathbf{u}_k, \delta \mathbf{u}_k, \mathbf{v}) + \nu \int_{\Omega} \nabla \delta \mathbf{u}_k : \nabla \mathbf{v} - \int_{\Omega} \delta p_k (\nabla \cdot \mathbf{v}) = R_k, & \forall \mathbf{v} \in \mathbb{H}_{E_0}^1, \\ \int_{\Omega} q (\nabla \cdot \delta \mathbf{u}_k) = r_k, & \forall q \in \mathbb{L}^2 \end{cases}$$

# The Navier-Stokes Equation: Discrete Picard

---

The **discrete system** is the same of the Newton method **without** the Newton matrix  $\mathbf{W}$ :

$$\mathcal{A}\delta = \begin{bmatrix} \nu\mathbf{A} + \mathbf{N} & \mathbf{B}^T \\ \mathbf{B} & \mathbf{O} \end{bmatrix} \begin{bmatrix} \delta\mathbf{u} \\ \delta\mathbf{p} \end{bmatrix} = \begin{bmatrix} \mathbf{f} \\ \mathbf{g} \end{bmatrix}$$

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- If we use *unstable* elements, we need a stabilization matrix.

## Theorem

Consider the generic saddle-point system

$$\mathcal{A} = \begin{bmatrix} \mathbf{F} & \mathbf{B}^T \\ \mathbf{B} & -\mathbf{C} \end{bmatrix},$$

where  $\mathbf{C}$  is symmetric and positive-semidefinite matrix. If  $\langle \mathbf{F}\mathbf{u}, \mathbf{u} \rangle > 0 \ \forall \mathbf{u} \neq \mathbf{0}$ , then

$$\ker \mathcal{A} = \left\{ \begin{bmatrix} \mathbf{0} \\ \mathbf{p} \end{bmatrix} \mid \mathbf{p} \in \ker(B\mathbf{F}^{-1}\mathbf{B}^T + \mathbf{C}) \right\}.$$

# The Navier-Stokes Equation: Newton and Picard

---

Newton

$$\mathcal{A} = \begin{bmatrix} \nu A + N + W_{xx} & W_{xy} & B_x^T \\ W_{yx} & \nu A + N + W_{yy} & B_y^T \\ B_x & B_y & O \end{bmatrix}$$

Picard

$$\mathcal{A} = \begin{bmatrix} \nu A + N & O & B_x^T \\ O & \nu A + N & B_y^T \\ B_x & B_y & O \end{bmatrix}$$

- Coupled  $\mathcal{A}_{1,1}$  block,
- Quadratic convergence,
- Locally convergent for “large enough”  $\nu$ , and “close enough” initial guess.

- Decoupled  $\mathcal{A}_{1,1}$  block,
- Linear convergence,
- Converges under the existence condition:  $\|\mathbf{f}\| < \nu^2 / \hat{C} C_\Omega^{1/2}$ .

# The Navier-Stokes Equation: Newton and Picard

Newton

$$\mathcal{A} = \begin{bmatrix} \nu A + N + W_{xx} & W_{xy} & B_x^T \\ W_{yx} & \nu A + N + W_{yy} & B_y^T \\ B_x & B_y & O \end{bmatrix}$$

Picard

$$\mathcal{A} = \begin{bmatrix} \nu A + N & O & B_x^T \\ O & \nu A + N & B_y^T \\ B_x & B_y & O \end{bmatrix}$$

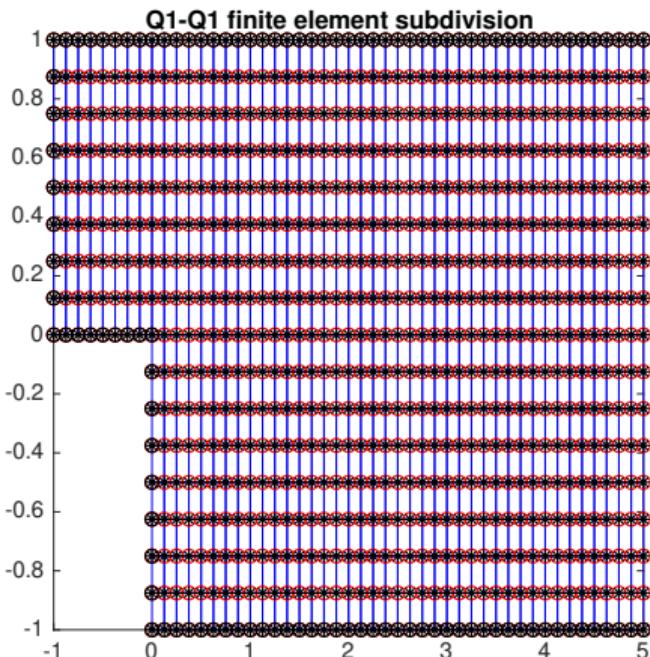
- Coupled  $\mathcal{A}_{1,1}$  block,
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- Locally convergent for “large enough”  $\nu$ , and “close enough” initial guess.
- Decoupled  $\mathcal{A}_{1,1}$  block,
- Linear convergence,
- Converges under the existence condition:  $\|\mathbf{f}\| < \nu^2 / \hat{C} C_\Omega^{1/2}$ .

Next week we will delve into some **numerical experiments**, and try several **preconditioners** discussed in the morning lectures.

# Navier-Stokes: backward facing step

Test problem:

- $L$ -shaped domain  $\Omega$ , parabolic inflow boundary condition, natural outflow boundary condition,



You can run the example as `E4-NavierStokes/navierstokes_solution.m`.

# Navier-Stokes: backward facing step

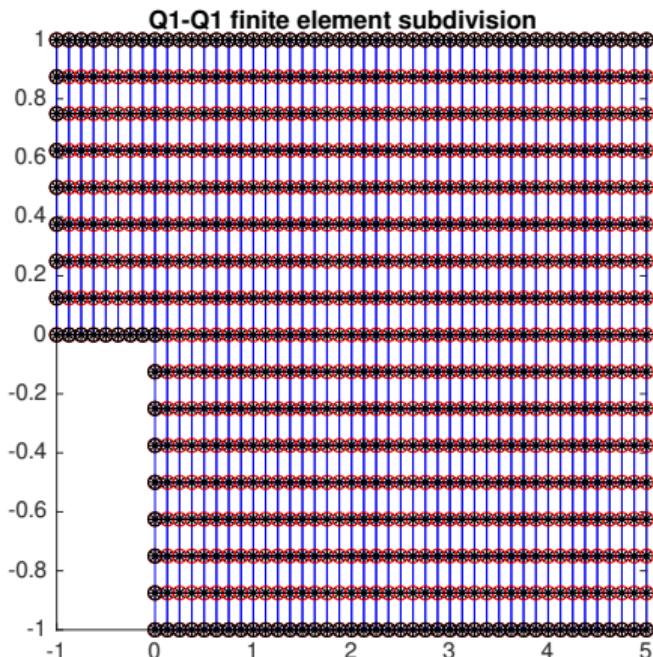
Test problem:

- L-shaped domain  $\Omega$ , **parabolic inflow boundary condition**, natural outflow boundary condition,

## Poiseuille flow

It is a steady horizontal flow in a channel driven by a pressure difference between the two ends

$$u_x = 1 - y^2, \quad u_y = 0, \quad p = -2\gamma x + \text{constant}.$$

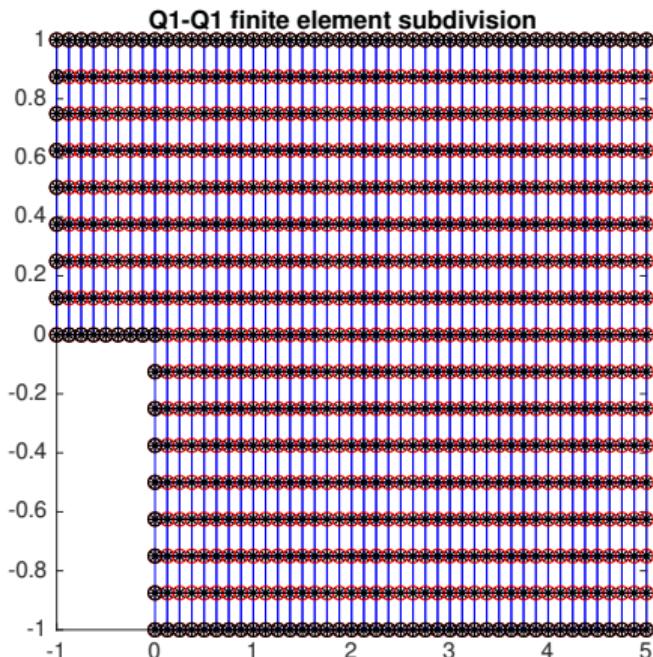


You can run the example as `E4-NavierStokes/navierstokes_solution.m`.

# Navier-Stokes: backward facing step

Test problem:

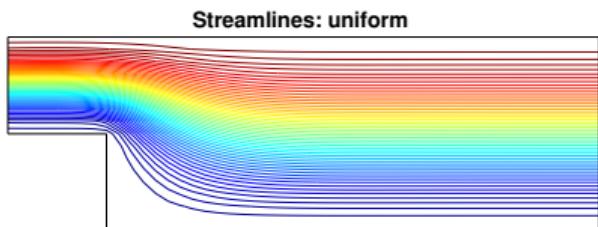
- $L$ -shaped domain  $\Omega$ , parabolic inflow boundary condition, natural outflow boundary condition,
- **Inflow**  $x = -1, 0 \leq y \leq 1$ ,  
**No flow** on the boundary,  
**Neumann condition** at the outflow  $x = L$ ,  
 $-1 < y < 1$ .
- Discretized with (unstable) Q1-Q1 elements.



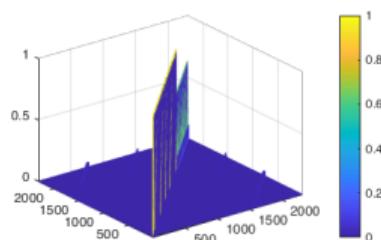
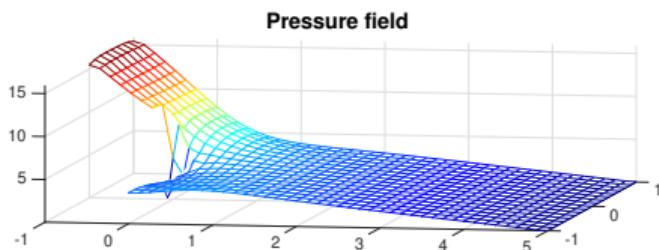
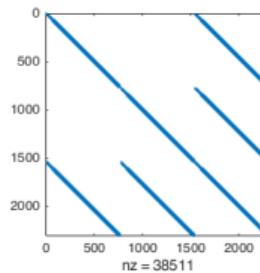
You can run the example as `E4-NavierStokes/navierstokes_solution.m`.

# Navier-Stokes: backward facing step

Initial guess:



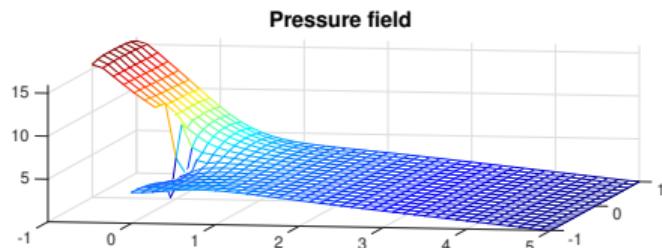
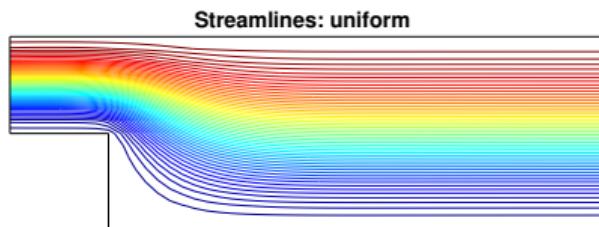
Picard's Iteration



Solution of the associated Stokes problem.

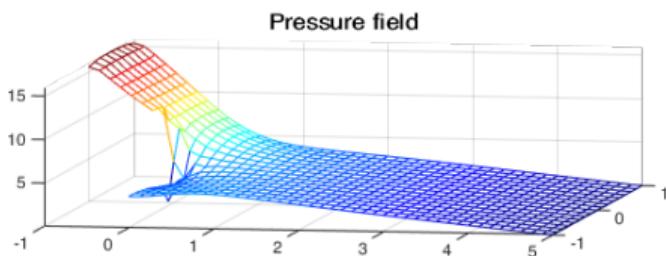
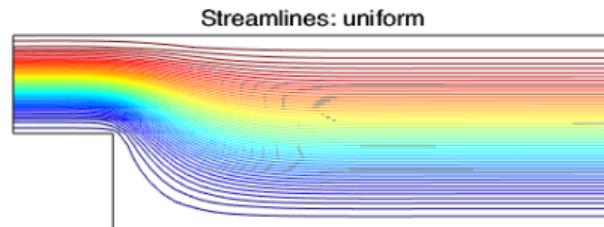
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

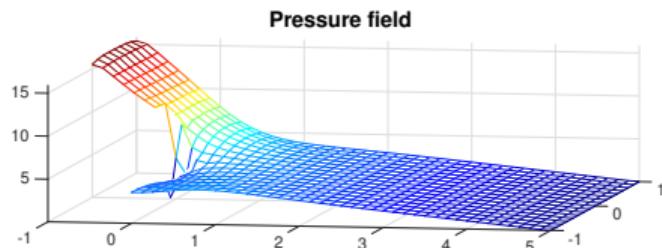
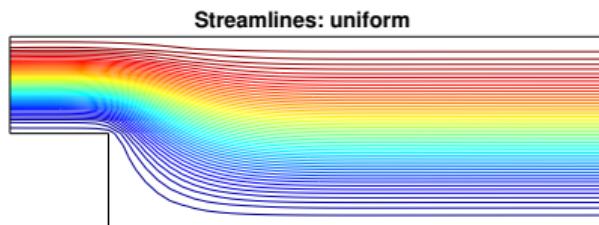
Picard's Iteration



Iteration 1

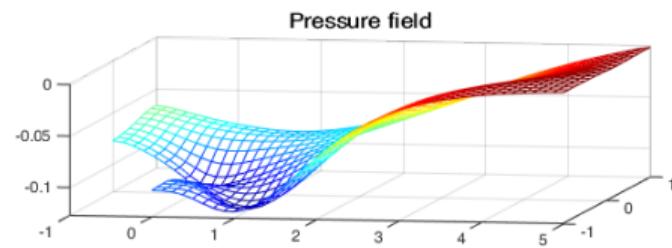
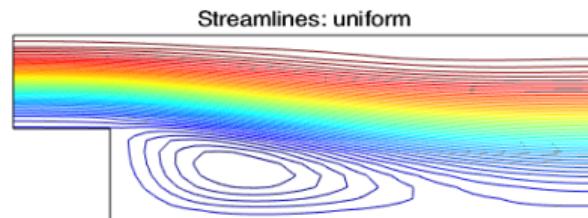
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

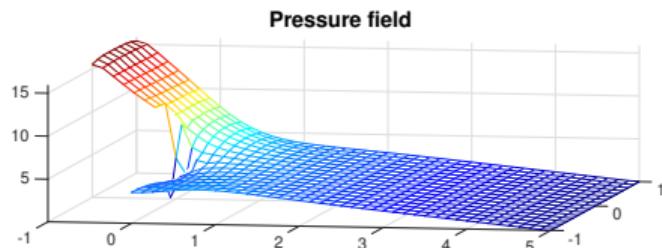
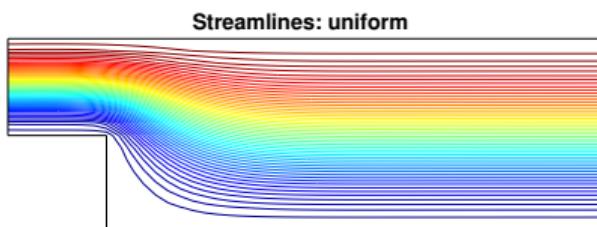
Picard's Iteration



Iteration 2

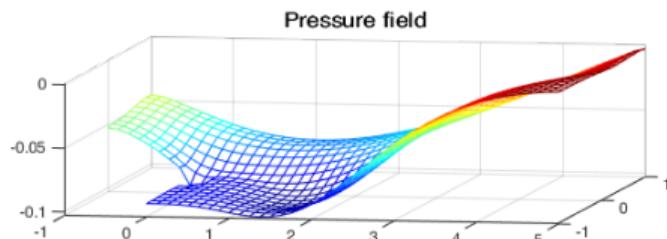
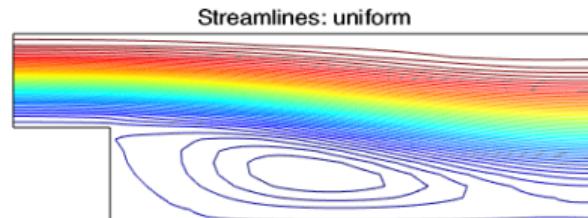
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

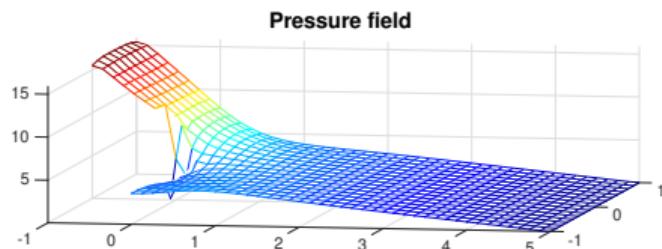
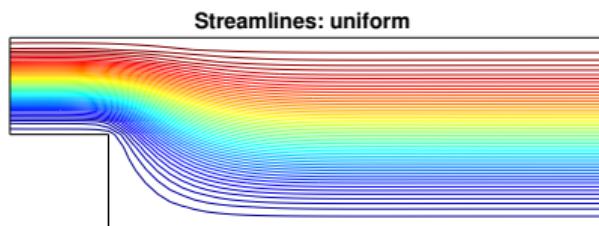
Picard's Iteration



Iteration 3

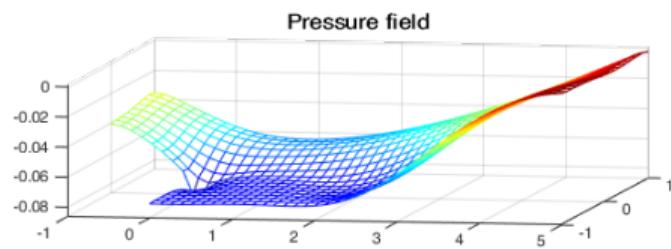
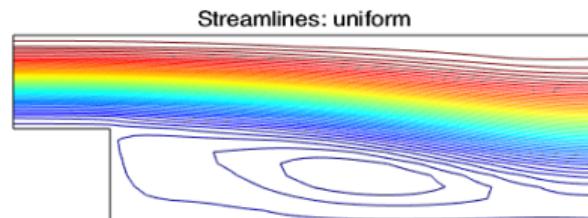
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

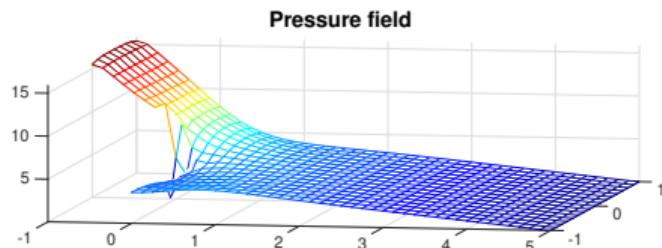
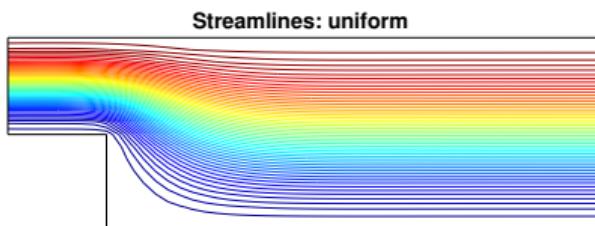
Picard's Iteration



Iteration 4

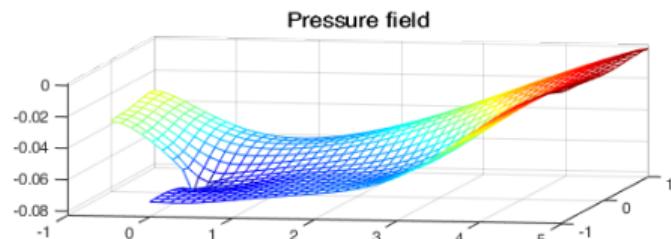
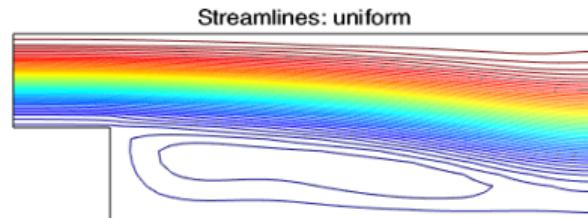
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

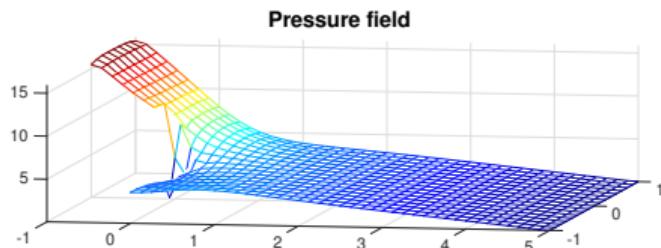
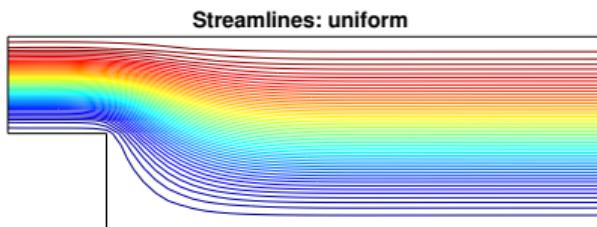
Picard's Iteration



Iteration 5

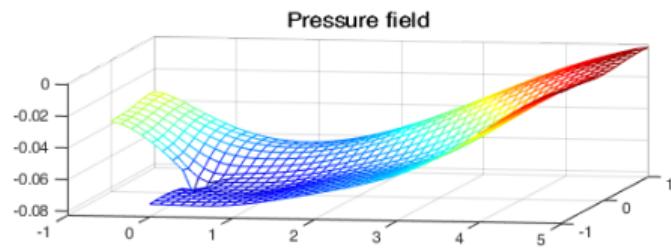
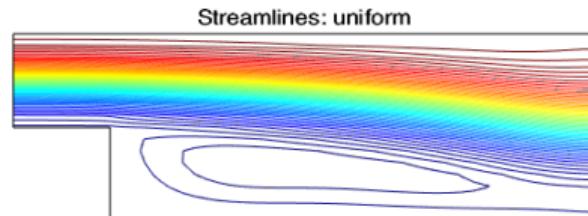
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

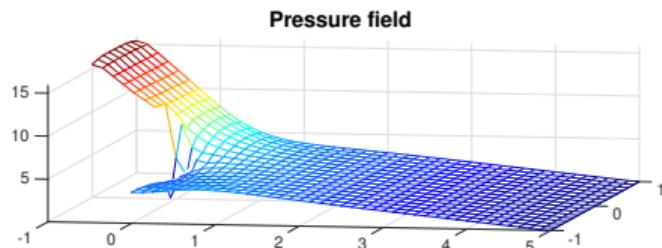
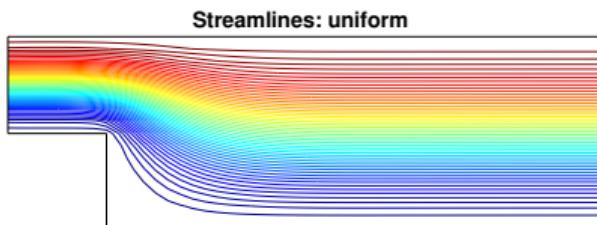
Picard's Iteration



Iteration 6

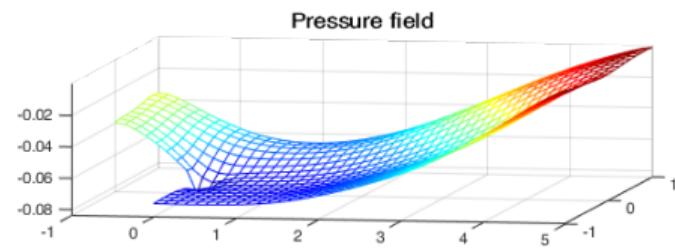
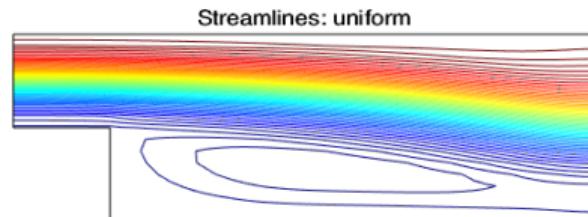
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

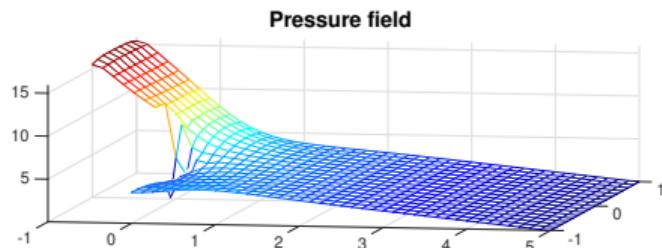
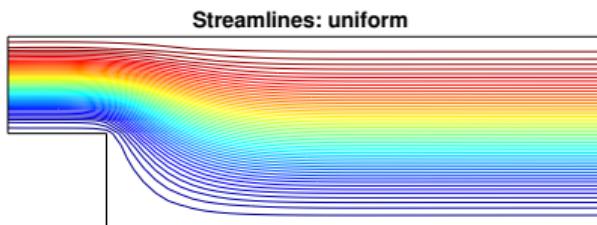
Picard's Iteration



Iteration 7

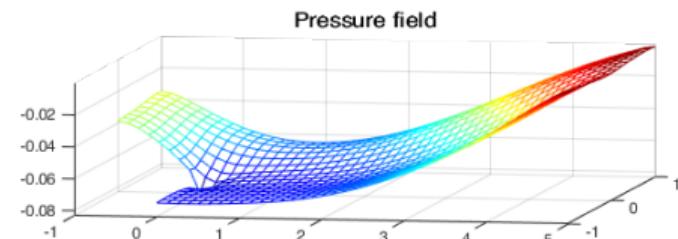
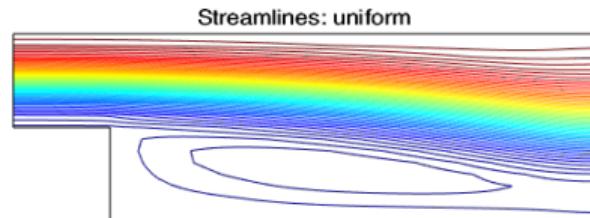
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

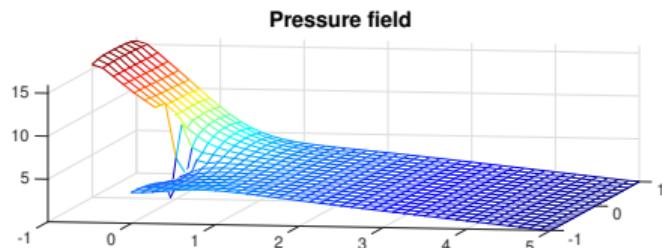
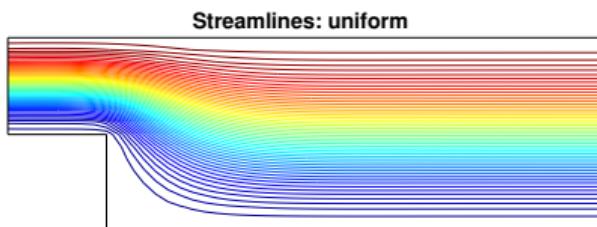
Picard's Iteration



Iteration 8

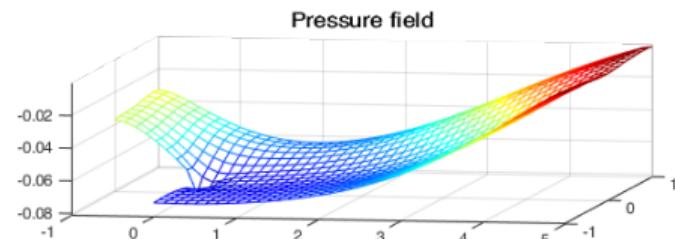
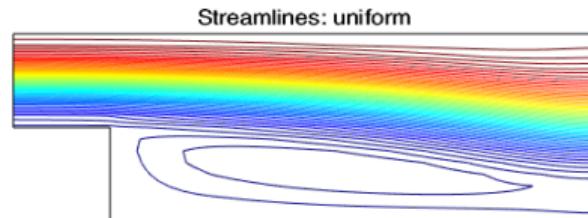
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

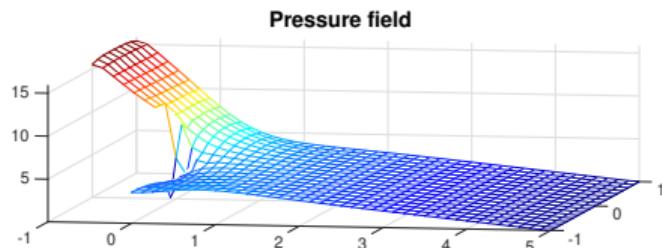
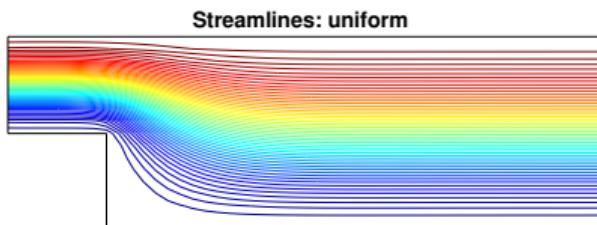
Picard's Iteration



Iteration 9

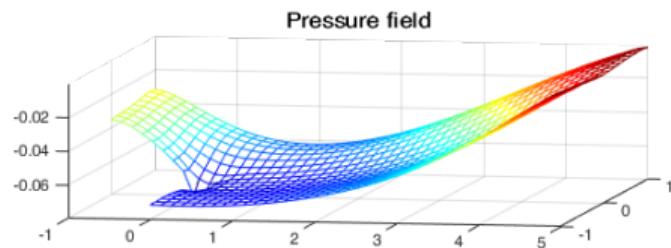
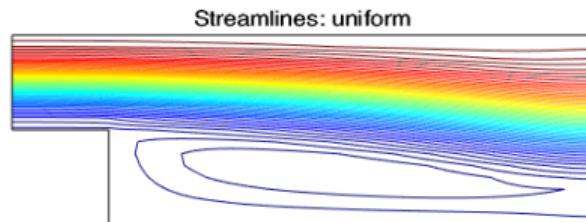
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

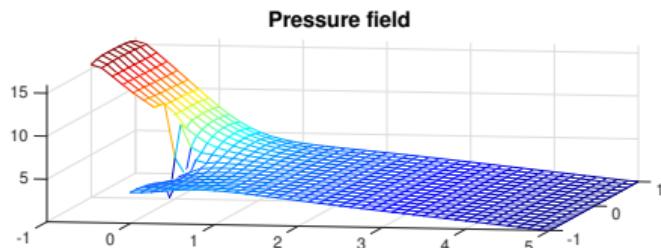
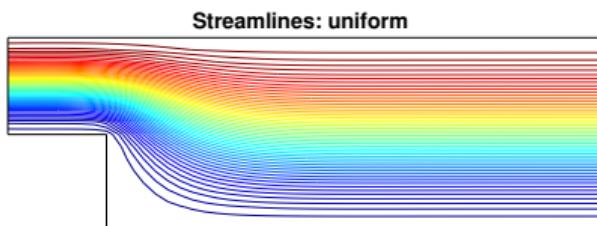
Picard's Iteration



Iteration 10

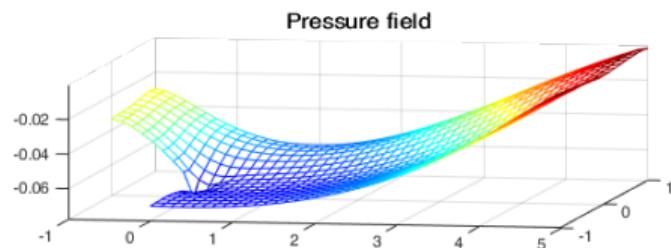
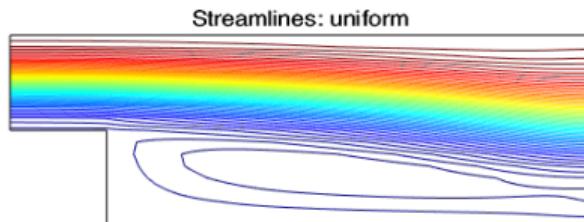
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

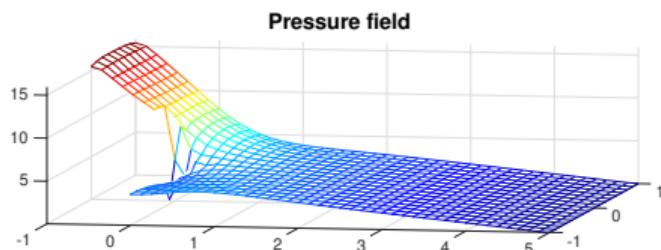
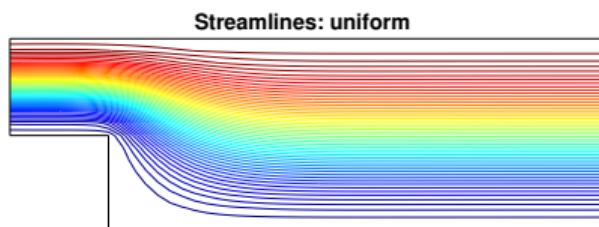
Picard's Iteration



Iteration 11

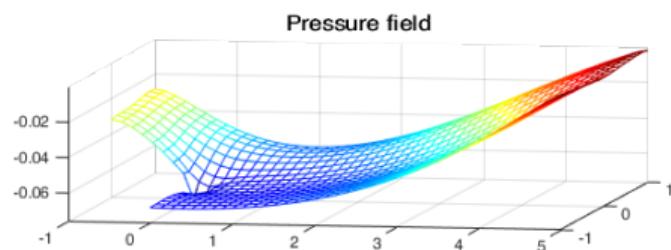
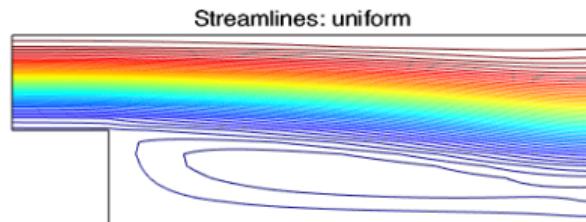
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

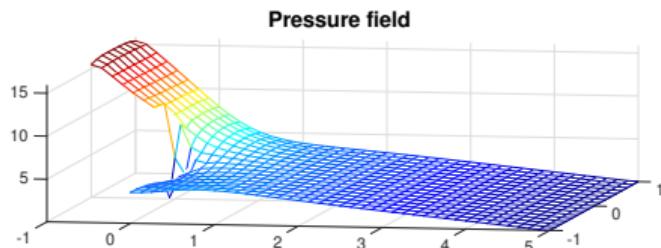
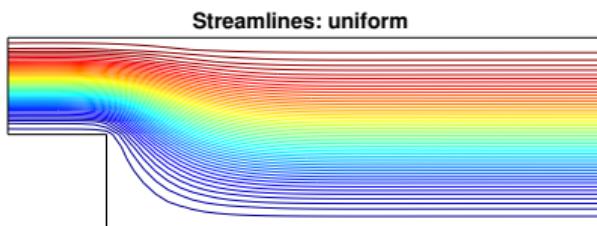
Picard's Iteration



Iteration 12

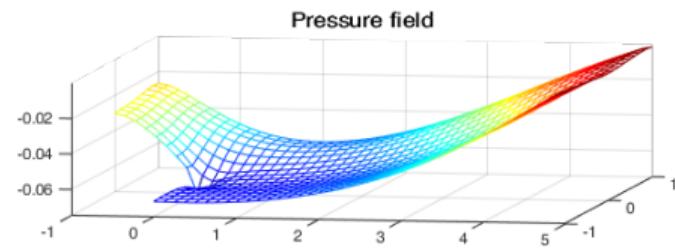
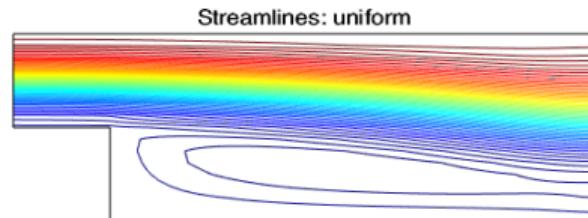
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

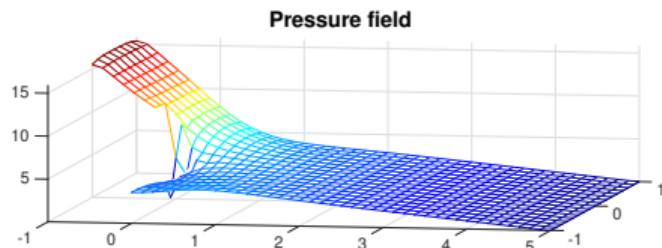
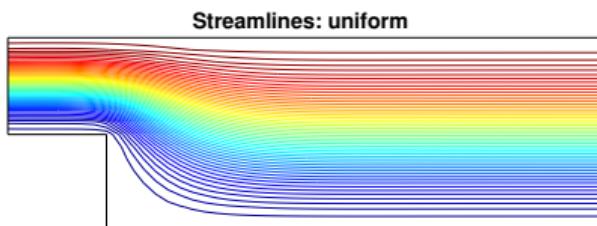
Picard's Iteration



Iteration 13

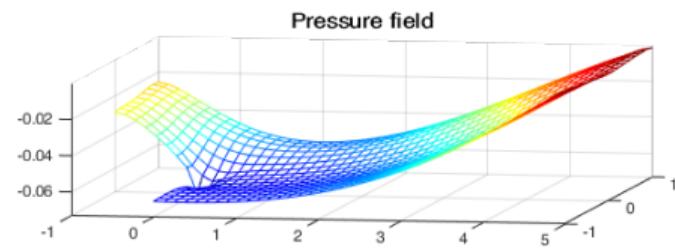
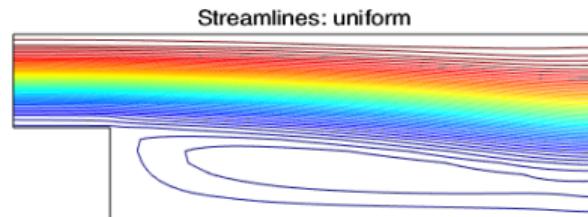
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

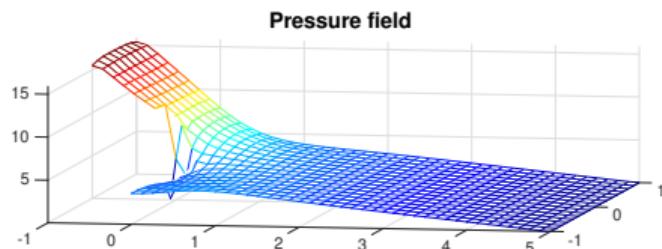
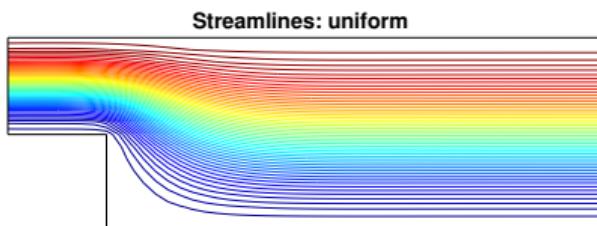
Picard's Iteration



Iteration 14

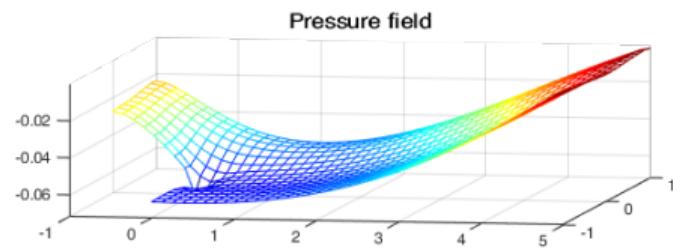
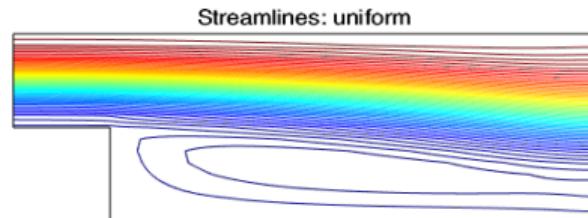
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

Picard's Iteration

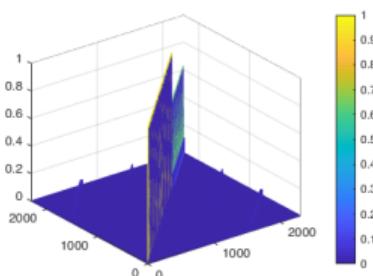
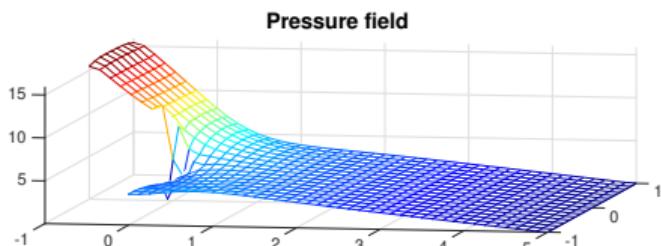
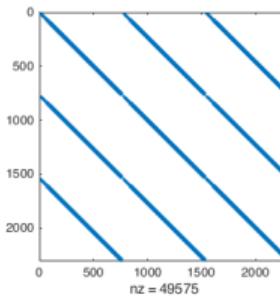
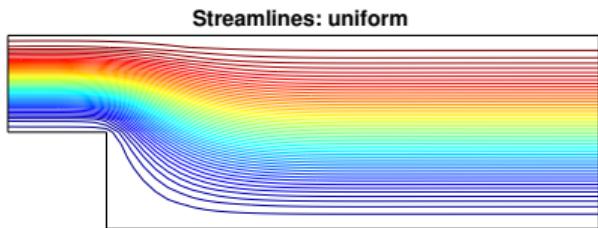


Iteration 15

# Navier-Stokes: backward facing step

Newton Method

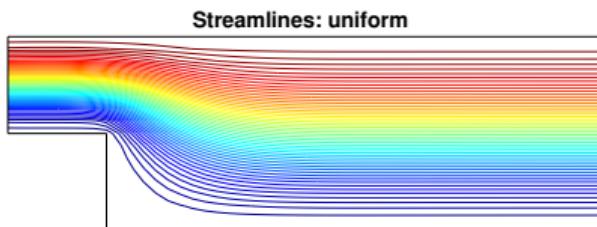
Initial guess:



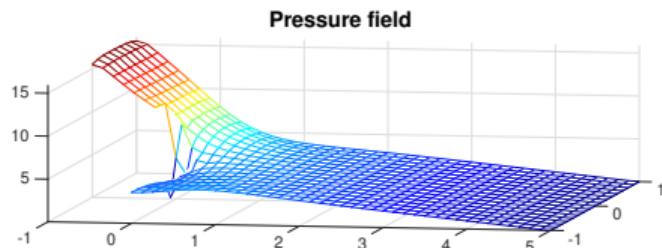
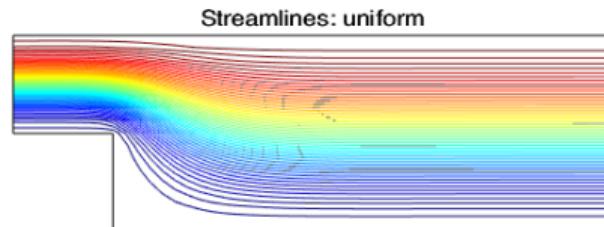
Solution of the associated Stokes problem.

# Navier-Stokes: backward facing step

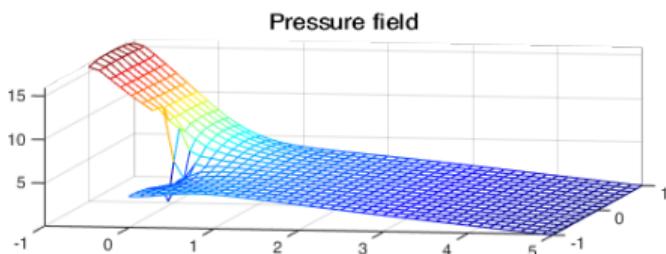
Initial guess:



Newton Method



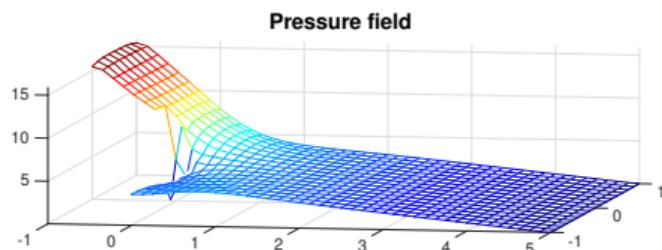
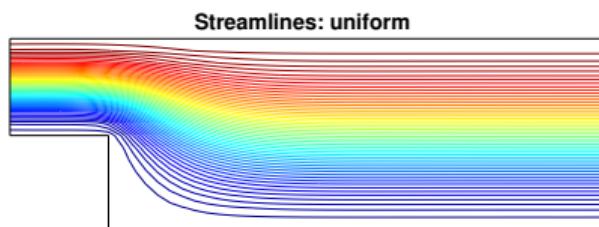
Solution of the associated Stokes problem.



Iteration 1

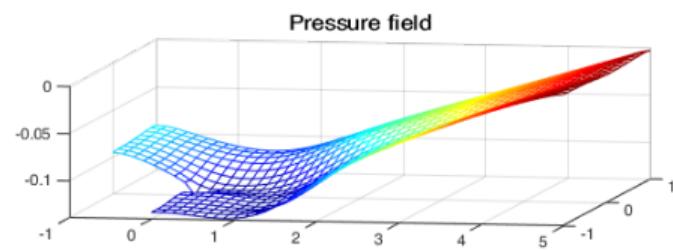
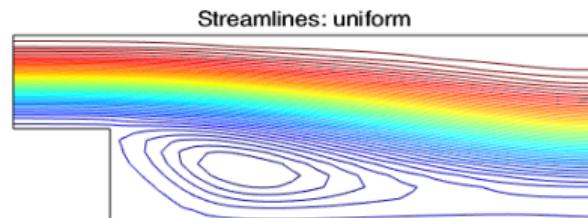
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

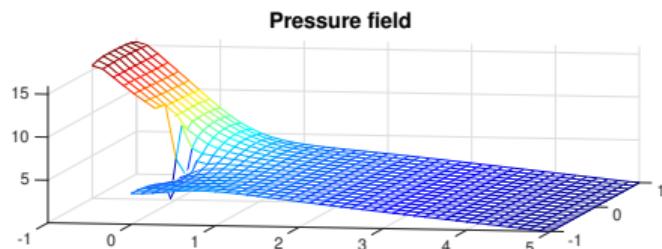
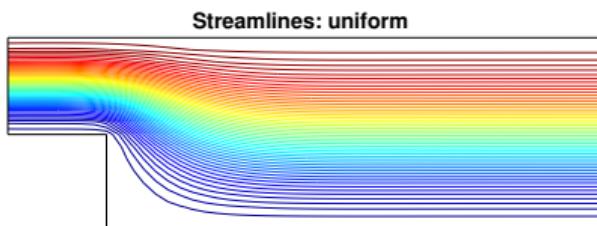
Newton Method



Iteration 2

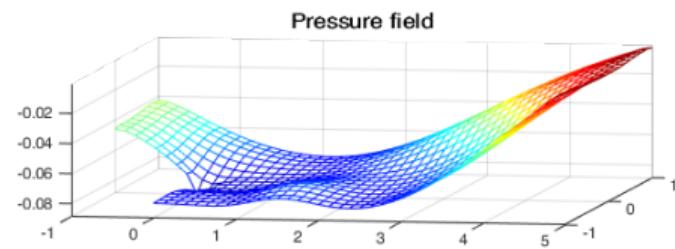
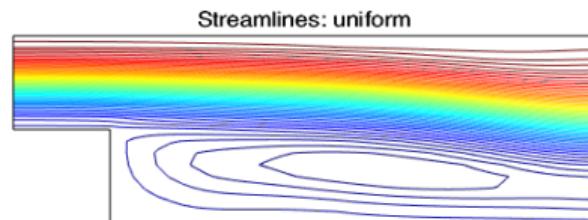
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

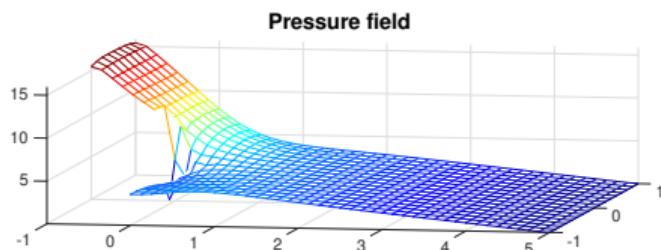
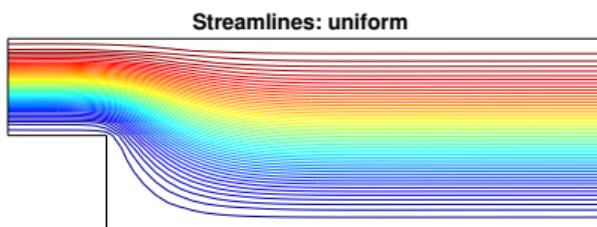
Newton Method



Iteration 3

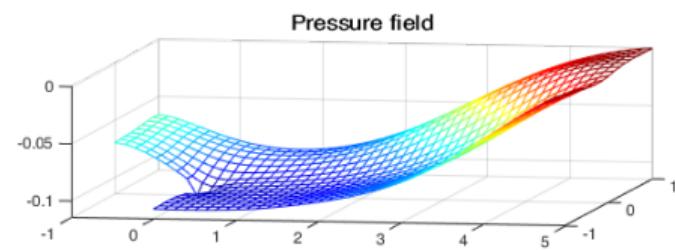
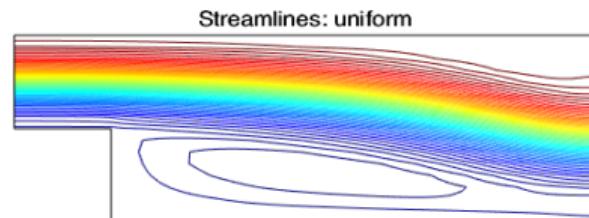
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

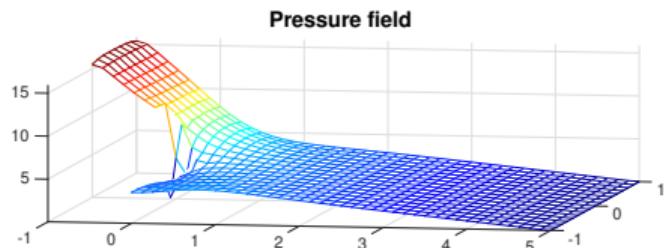
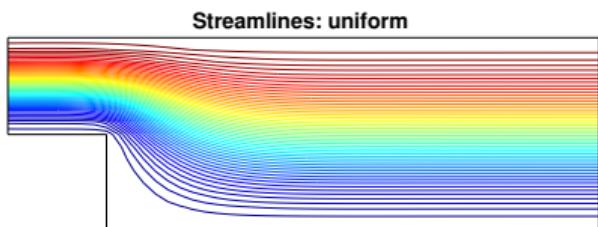
Newton Method



Iteration 4

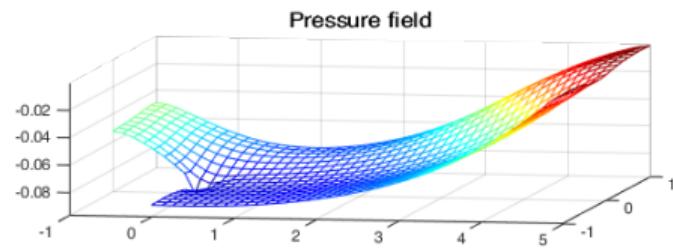
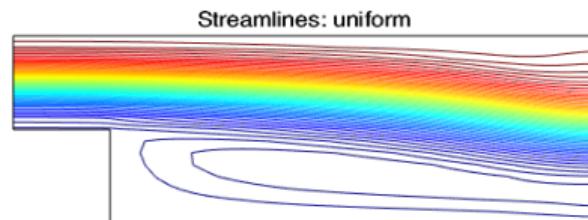
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

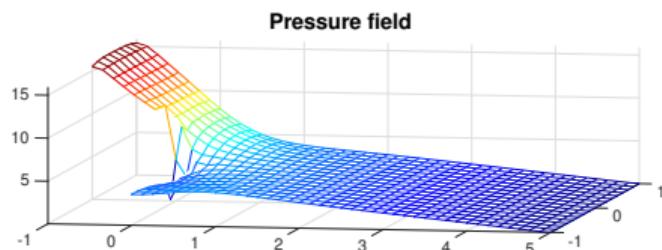
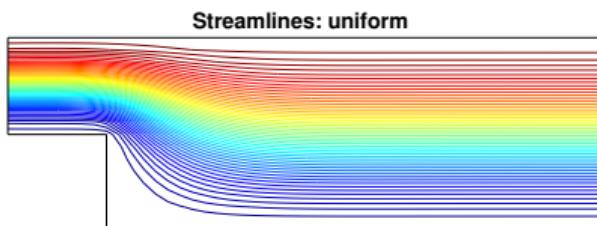
Newton Method



Iteration 5

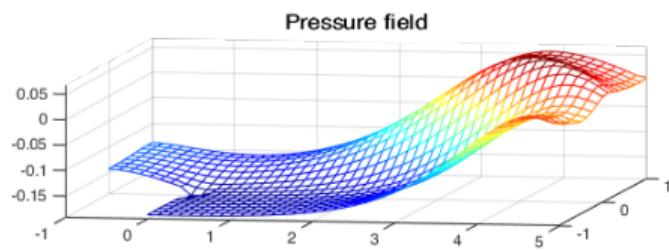
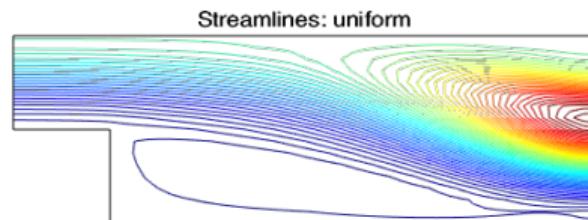
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

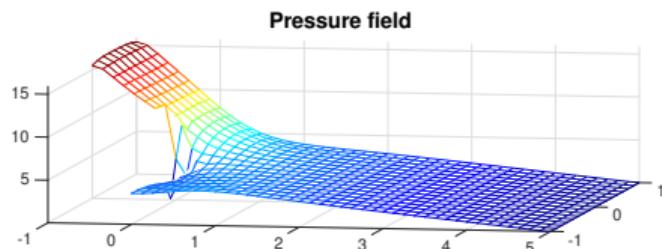
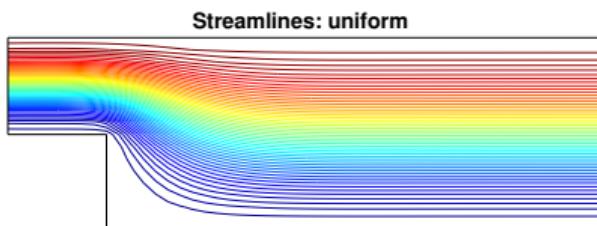
Newton Method



Iteration 6

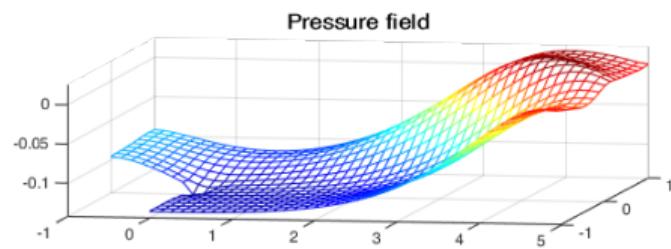
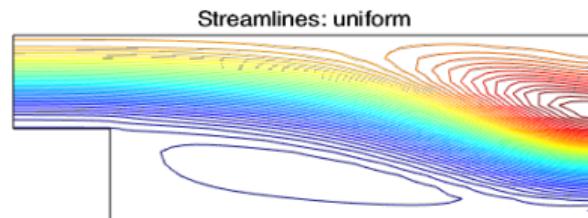
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

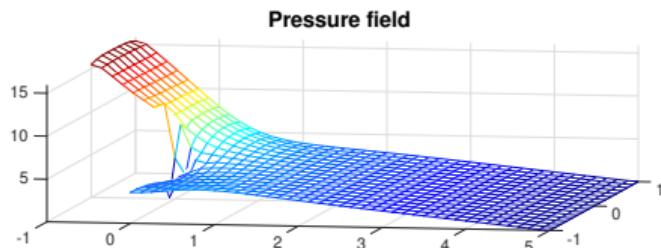
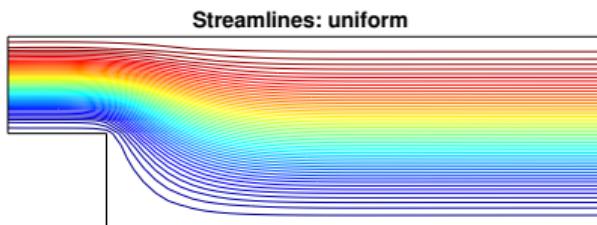
Newton Method



Iteration 7

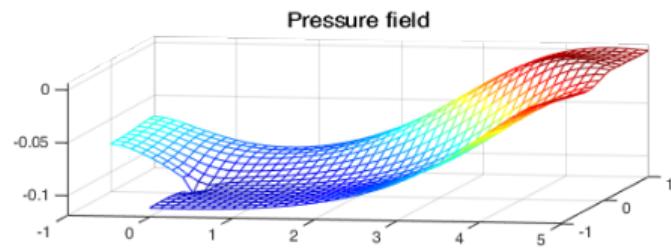
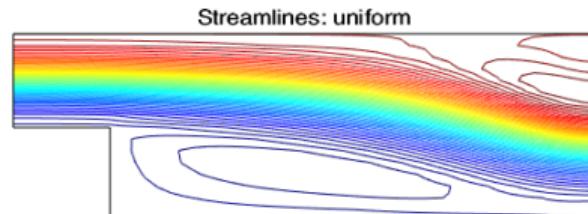
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

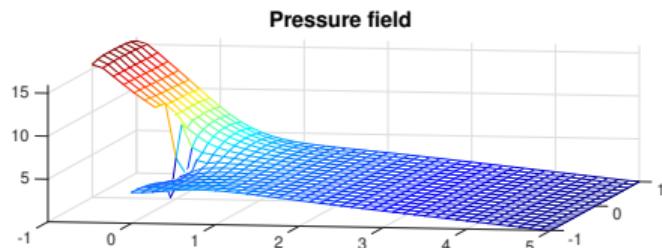
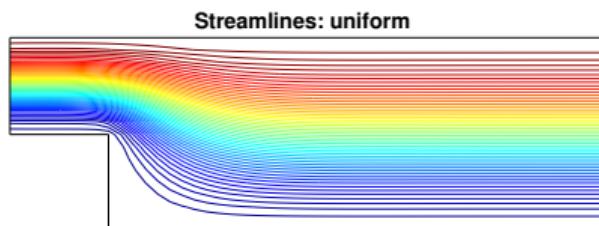
Newton Method



Iteration 8

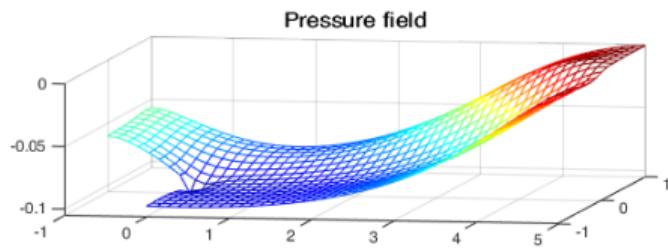
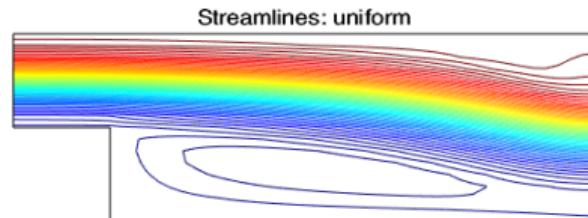
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

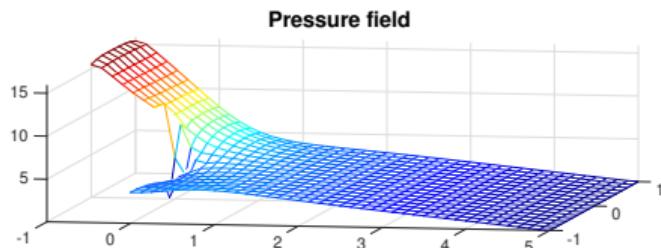
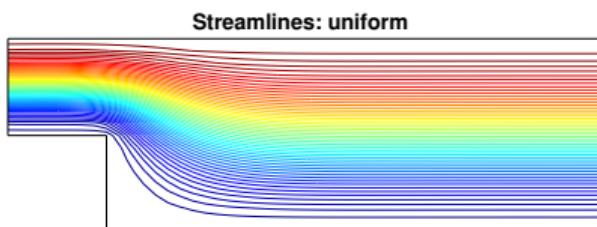
Newton Method



Iteration 9

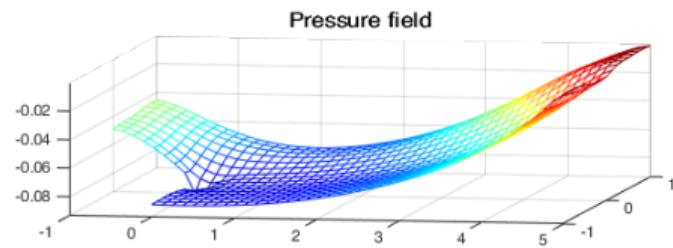
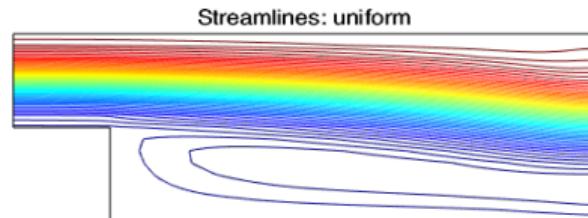
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

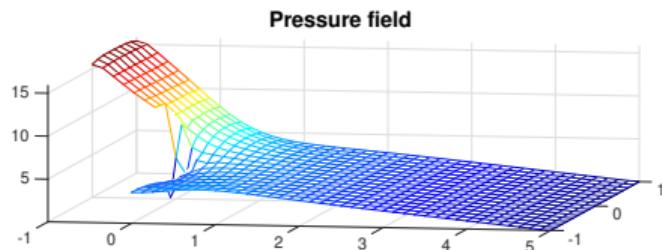
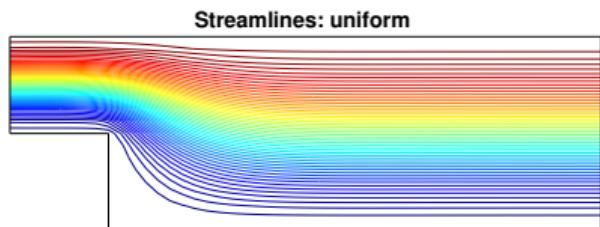
Newton Method



Iteration 10

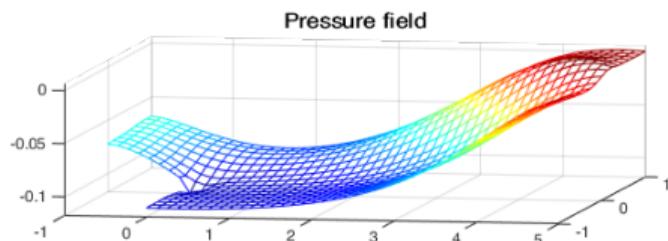
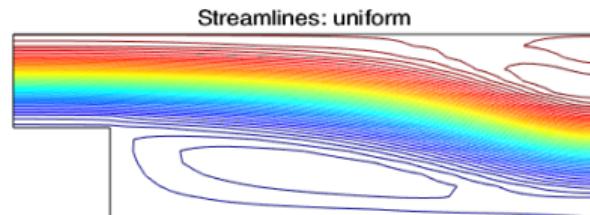
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

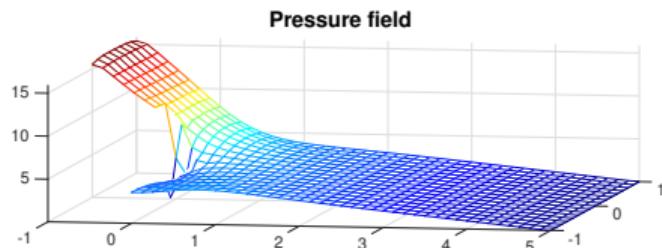
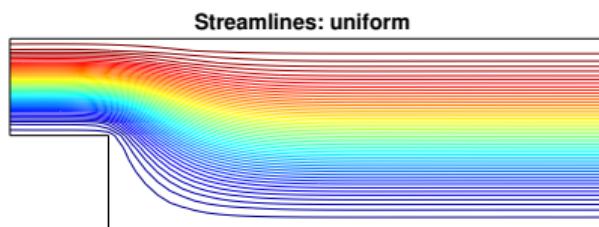
Newton Method



Iteration 11

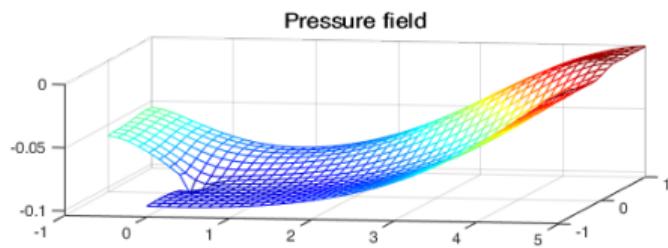
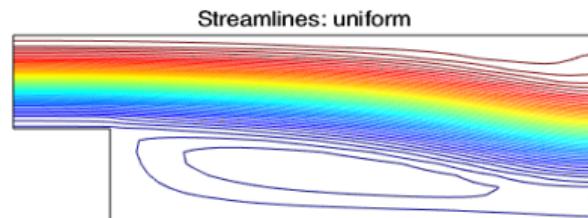
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

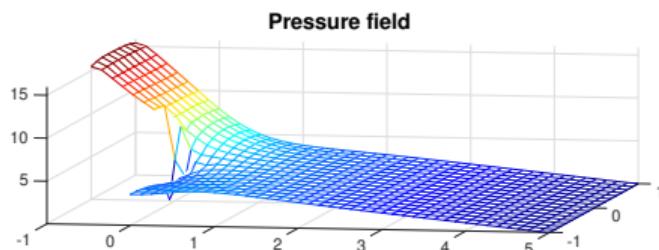
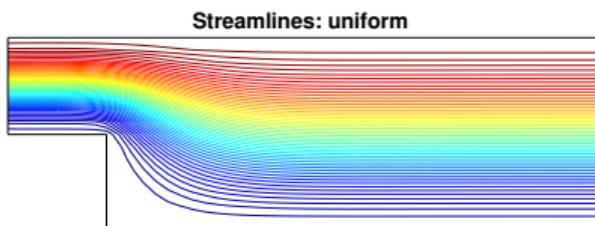
Newton Method



Iteration 12

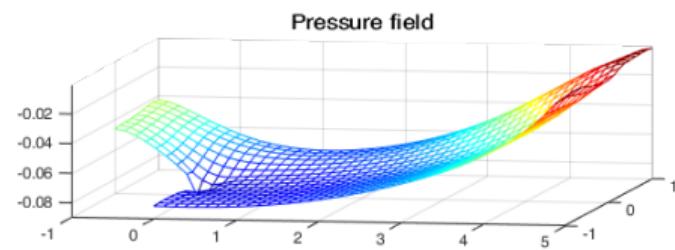
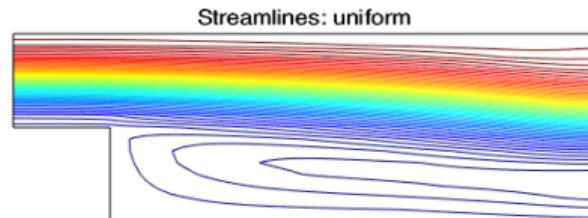
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

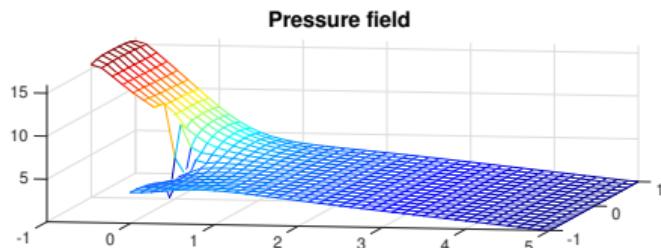
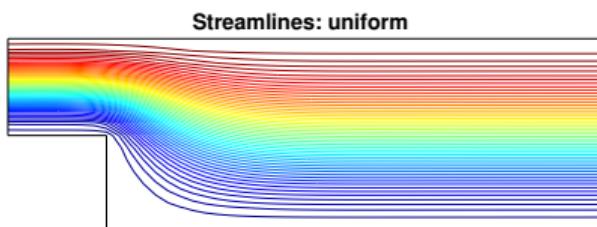
Newton Method



Iteration 13

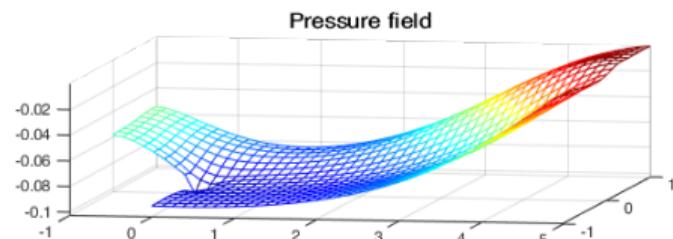
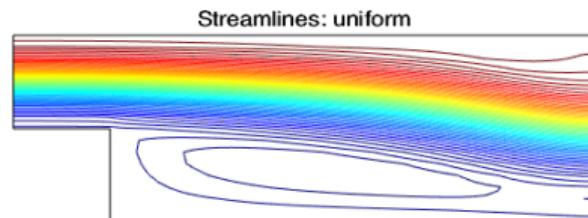
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

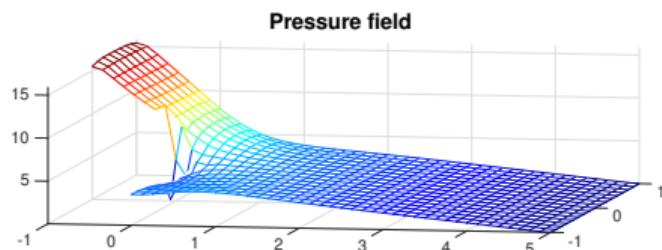
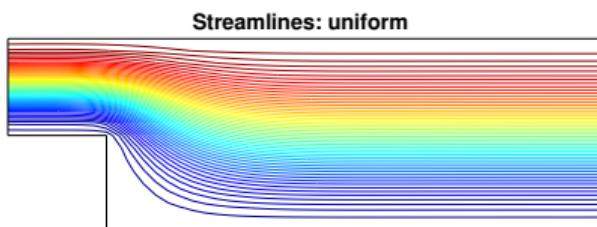
Newton Method



Iteration 14

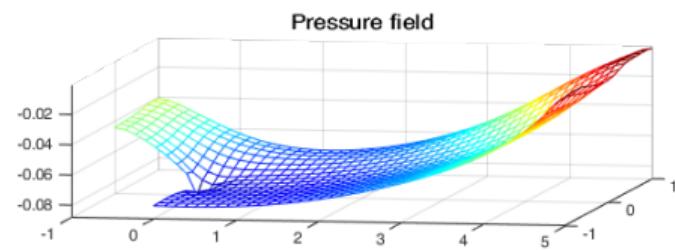
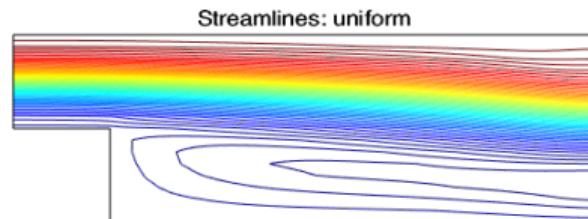
# Navier-Stokes: backward facing step

Initial guess:



Solution of the associated Stokes problem.

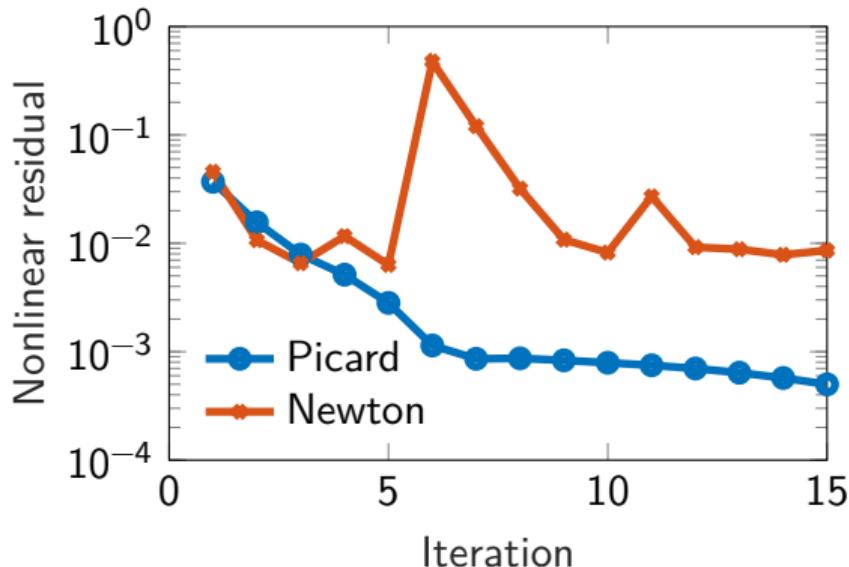
Newton Method



Iteration 15

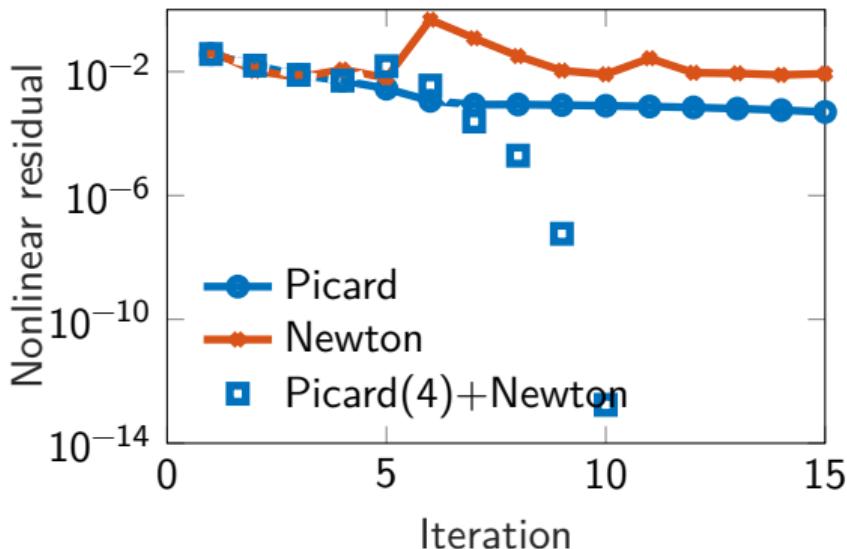
# Navier-Stokes: backward facing step

- For this test problem convergence of the Newton method from the Stokes initial data is quite poor, what can we do?



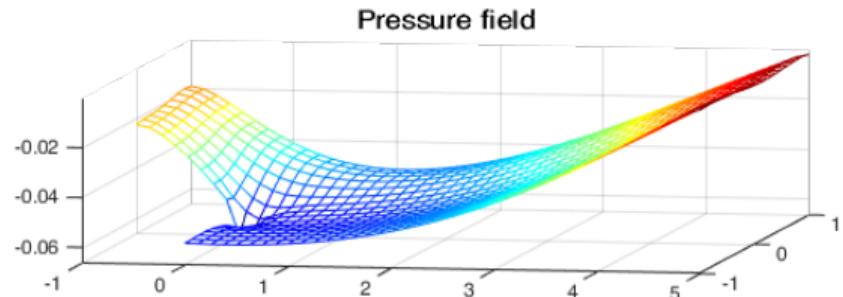
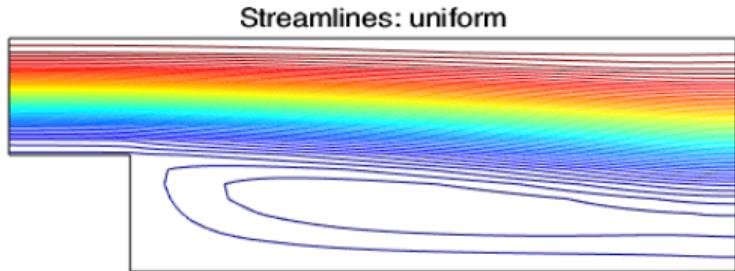
# Navier-Stokes: backward facing step

- For this test problem convergence of the Newton method from the Stokes initial data is quite poor, what can we do?
- We start from Stokes, then perform few steps of Picard's iteration, and finally accelerate with Newton.



# Navier-Stokes: backward facing step

- For this test problem convergence of the Newton method from the Stokes initial data is quite poor, what can we do?
- We start from Stokes, then perform few steps of Picard's iteration, and finally accelerate with Newton.
- The “mess” doesn't end here – unfortunately or fortunately, I'm not yet sure...*boundary layers, bifurcations, absence of stable flows,...*



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