2.5 Mixed formulation for second order equations

Motivation:

- exact flux conservation
- · useful for a posteriori error estimates
- ullet model problem for 4^{th} order problems, Stokes, ...

We consider the diffusion equation

$$-\operatorname{div} \lambda
abla u = f \quad ext{in } \Omega$$
 $u = u_D \quad ext{on } \Gamma_D$
 $\lambda rac{\partial u}{\partial n} = g \quad ext{on } \Gamma_N$

Primal variational formulation

Find $u \in H^1, u = u_D$ on Γ_D such that

$$\int_{\Omega} \lambda
abla u
abla v = \int_{\Omega} f v + \int_{\Gamma_N} g v \quad orall v \in H^1, v = 0 ext{ on } \Gamma_D$$

First order system

Find scalar u and the flux σ such that

$$\lambda^{-1}\sigma=
abla u,\quad {
m div}\,\sigma=-f$$

with boundary conditions

$$\sigma \cdot n = g \text{ on } \Gamma_N, \quad \text{ and } \quad u = u_D \text{ on } \Gamma_D$$

Mixed variational formulation

importing NGSolve-6.2.2204

Find $(\sigma,u)\in H(\mathrm{div}) imes L_2$ such that $\sigma_n=g$ on Γ_N and

$$\int_{\Omega} \lambda^{-1} \sigma au + \operatorname{div} \sigma v + \operatorname{div} au u = - \int_{\Omega} f v + \int_{\Gamma_D} u_D au_n$$

for all test-functions $(au,v)\in H(\mathrm{div}) imes L_2$ with $au_n=0$.

Here σ_n is the normal trace operator $\sigma \cdot n|_{\Gamma_N}$, which is meaningful in $H(\mathrm{div})$.

```
from netgen.geom2d import unit_square
from ngsolve import *
from ngsolve.webgui import Draw

mesh = Mesh(unit_square.GenerateMesh(maxh=0.1))
```

Setup and solve primal problem:

```
In [3]:
         fesp = H1(mesh, order=4, dirichlet="bottom")
         up, vp = fesp.TnT()
         ap = BilinearForm(fesp)
         ap += lam*grad(up)*grad(vp)*dx
         ap.Assemble()
         fp = LinearForm(fesp)
         fp += source*vp*dx + g*vp * ds
         fp.Assemble()
         gfup = GridFunction(fesp, "u-primal")
         gfup.Set(ud, BND)
         r = fp.vec.CreateVector()
         r.data = fp.vec - ap.mat * gfup.vec
         gfup.vec.data += ap.mat.Inverse(freedofs=fesp.FreeDofs()) * r
         Draw (gfup)
         Draw (lam * grad(gfup), mesh, "flux-primal")
        WebGuiWidget(layout=Layout(height='50vh', width='100%'), value={'qui settin
        gs': {}, 'ngsolve version': '6.2.22...
        WebGuiWidget(layout=Layout(height='50vh', width='100%'), value={'gui settin
        gs': {}, 'ngsolve_version': '6.2.22...
        BaseWebGuiScene
Out[31:
```

Solving the mixed problem

Define spaces for mixed problem. Note that essential boundary conditions are set to the $H(\operatorname{div})$ -component on the opposite boundary. Creating a space from a list of spaces generates a product space:

```
In [4]: order_flux=1
V = HDiv(mesh, order=order_flux, dirichlet="right|top|left")
Q = L2(mesh, order=order_flux-1)
fesm = V*Q
```

The space provides now a list of trial and test-functions:

```
In [5]:
    sigma, u = fesm.TrialFunction()
    tau, v = fesm.TestFunction()
```

The normal vector is provided as a *special* coefficient function (which may be used only at the boundary). The orientation depends on the definition of the geometry. In 2D, it is the tangential vector rotated to the right, and is the outer vector in our case. Since every CoefficientFunction must know its vector-dimension, we have to provide the spatial dimension:

```
In [6]:
```

```
normal = specialcf.normal(mesh.dim)
print (normal)
```

coef normal vector, real, dim=2

Define the forms on the product space. For the boundary term, we have to use the Trace operator, which provides the projection to normal direction.

```
In [7]:
    am = BilinearForm(fesm)
    am += (1/lam*sigma*tau + div(sigma)*v + div(tau)*u)*dx
    am.Assemble()
    fm = LinearForm(fesm)
    fm += -source*v*dx + ud*(tau.Trace()*normal)*ds
    fm.Assemble()
    gfm = GridFunction(fesm, name="gfmixed")
```

The proxy-functions used for the forms know to which component of the product space they belong to. To access components of the solution, we have to unpack its components. They don't have own coefficient vectors, but refer to ranges of the big coefficient vector.

```
In [8]: gfsigma, gfu = gfm.components
```

Just to try something:

Now set the essential boundary conditions for the flux part:

```
In [10]: gfsigma.Set(g*normal, BND)

In [12]: # fm.vec.data -= am.mat * gfm.vec
    # gfm.vec.data += am.mat.Inverse(freedofs=fesm.FreeDofs(), inverse="umfpacts solvers.BVP(bf=am, lf=fm, gf=gfm)
    Draw (gfsigma, mesh, "flux-mixed")
    Draw (gfu, mesh, "u-mixed")
```

NgException: SparseMatrix::InverseMatrix: UmfpackInverse not available Calculate the difference:

```
In [ ]:
    print ("err-u: ", sqrt(Integrate( (gfup-gfu)**2, mesh)))
    errflux = lam * grad(gfup) - gfsigma
    print ("err-flux:", sqrt(Integrate(errflux*errflux, mesh)))
```

Post-processing for the scalar variable

The scalar variable is approximated one order lower than the vector variable, what is its gradient. Knowing the gradient of a function more accurate, and knowing its mean value, one can recover the function itself. For this post-processing trick we refer to [Arnold+Brezzi 85]

find $\widehat{u} \in P^{k+1,dc}$ and $\widehat{\lambda} \in P^{0,dc}$ such that

$$\begin{array}{ccccc} \int \lambda \nabla \widehat{u} \nabla \widehat{v} & + & \int \widehat{\lambda} \widehat{v} & = & \int \sigma \nabla \widehat{v} & \forall \, \widehat{v} \\ & \int \widehat{u} \widehat{\mu} & = & \int u \widehat{\mu} & \forall \, \widehat{\mu} \end{array}$$

```
In [ ]:
         fespost u = L2(mesh, order=order flux+1)
         fespost_lam = L2(mesh, order=0)
         fes_post = fespost_u*fespost_lam
         u,la = fes post.TrialFunction()
         v,mu = fes post.TestFunction()
         a = BilinearForm(fes post)
         a += (lam*grad(u)*grad(v)+la*v+mu*u)*dx
         a.Assemble()
         f = LinearForm(fes post)
         f += (gfsigma*grad(v)+gfu*mu)*dx
         f.Assemble()
         gfpost = GridFunction(fes post)
         gfpost.vec.data = a.mat.Inverse() * f.vec
         Draw (gfpost.components[0], mesh, "upost")
         print ("err-upost: ", sqrt(Integrate( (gfup-gfpost.components[0])**2, mes
```

```
In []:
```