

PDEs, modeling & simulation

Achim Schroll



- ▶ potential theory

$$-\Delta\phi = \rho \quad \rightsquigarrow \quad F = m\nabla\phi$$

- ▶ heat transfer

$$T_t + u \cdot \nabla T - \frac{1}{Pe} \Delta T = s$$

- ▶ fluid flow

$$u_t + u \nabla u + \nabla p - \frac{1}{Re} \Delta u = f$$
$$\nabla \cdot u = 0$$

- ▶ continuum / quantum mechanics

$$u_t + \nabla \cdot F(u) = s \quad , \quad i\hbar\psi_t = \left[\frac{-\hbar^2}{2\mu} \Delta + V \right] \psi$$

$$\nabla = (\partial_{x_1}, \partial_{x_2}, \dots, \partial_{x_n})^T$$

$$\nabla f = \operatorname{grad} f$$

$$\Delta f = \nabla \cdot \nabla f = \langle \nabla, \nabla f \rangle = \nabla^T \nabla f = \nabla^2 f$$

$$\nabla \cdot \vec{u} = \langle \nabla, \vec{u} \rangle = \nabla^T \vec{u} = \operatorname{div} \vec{u}$$

$$\nabla \times \vec{u} = \operatorname{rot} \vec{u}$$

product rule

$$\nabla \cdot (g\vec{f}) = \nabla g \cdot \vec{f} + g(\nabla \cdot \vec{f})$$

div-theorem (Gauss's theorem, fundamental theorem of calculus)

$$\int_{\Omega} \nabla \cdot \vec{f} \, dx = \int_{\partial\Omega} \vec{f} \cdot \vec{n} \, dS$$

Green's first identity (partial integration)

$$\int_{\Omega} g \Delta \phi \, dx = \int_{\partial\Omega} g \nabla \phi \cdot \vec{n} \, dS - \int_{\Omega} \nabla g \cdot \nabla \phi \, dx$$

classification of linear, second order PDEs

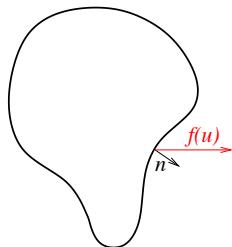
$$Au_{tt} + 2Bu_{tx} + Cu_{xx} + \text{l.o.t.} = 0$$

$$\begin{cases} B^2 - AC < 0 & \text{elliptic} \\ B^2 - AC = 0 & \text{parabolic} \\ B^2 - AC > 0 & \text{hyperbolic} \end{cases}$$

prototypes:

$$\begin{cases} \pm \Delta u = f & \text{Poisson eqn, elliptic} \\ u_t = \nabla \cdot (d \nabla u) & \text{heat eqn, parabolic} \\ u_{tt} = c^2 u_{xx} & \text{wave eqn, hyperbolic} \end{cases}$$

conservation laws



$$\frac{d}{dt} \int_{\Omega} u(t, x) \, dx + \int_{\partial\Omega} f(u(t, x)) \cdot n \, dS = 0$$

divergence theorem: $\int_{\partial\Omega} f \cdot n \, dS = \int_{\Omega} \nabla \cdot f \, dx$

$$u_t + \nabla \cdot f(u) = 0$$

$$\frac{d}{dt} \int_{\Omega} u(t, x) \, dx + \int_{\partial\Omega} f(u(t, x)) \cdot n \, dS = 0$$

Fick's first law (1855): $f(u) = -D\nabla u$

divergence theorem

$$\int_{\Omega} u_t(t, x) - \nabla \cdot (D\nabla u) \, dx = 0$$

diffusion / heat equation

$$u_t = \nabla \cdot (D\nabla u) \, , \quad x \in \Omega \, , \quad t > 0$$

the Poisson equation

$$\begin{cases} -\Delta u = f & x \in \Omega \\ u(x) = u_0(x) & x \in \partial\Omega \end{cases}$$

variational formulation

$$-\int_{\Omega} \Delta u v \, dx = \int_{\Omega} f v \, dx$$

Green's first identity

$$\int_{\Omega} v \Delta w \, dx = \int_{\partial\Omega} v \frac{\partial w}{\partial n} \, dS - \int_{\Omega} \nabla v \cdot \nabla w \, dx$$

test function $v = 0$ on $\partial\Omega \rightsquigarrow \nabla$ is anti-symmetric:

$$\int_{\Omega} v \nabla^2 w \, dx = - \int_{\Omega} \nabla v \cdot \nabla w \, dx$$

the Poisson equation

$$-\int_{\Omega} \nabla \cdot \nabla u v \, dx = \int_{\Omega} f v \, dx$$

weak, variational formulation

$$\int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx, \quad \forall v \in \hat{V}$$

solution space

$$V = \{u \in H^1(\Omega) : u = u_0 \text{ on } \partial\Omega\}$$

test space

$$\hat{V} = \{v \in H^1(\Omega) : v = 0 \text{ on } \partial\Omega\}$$

Hilbert (Sobolev) spaces

$$(f, g)_{L^2} = \int_{\Omega} f(x)g(x) \, dx \quad , \quad \|f\|_{L^2}^2 = (f, f)_{L^2} = \int_{\Omega} f(x)^2 \, dx$$

$$L^2(\Omega) = \{f : \|f\|_{L^2} < \infty\}$$

$$(f, g)_{H^1} = \int_{\Omega} fg + \nabla f \cdot \nabla g \, dx \quad , \quad \|f\|_{H^1}^2 = (f, f)_{H^1} = \|f\|_{L^2}^2 + \|\nabla f\|_{L^2}^2$$

$$H^1(\Omega) = \{f \in L^2(\Omega) : \nabla f \in L^2(\Omega)\}$$

Hilbert spaces are complete!

weak derivatives in L^2

for $w \in C^1(\Omega)$ recall

$$\int_{\Omega} \nabla w(x) \phi(x) \, dx = - \int_{\Omega} w(x) \nabla \phi(x) \, dx \quad , \quad \forall \phi \in C_0^\infty(\Omega) \quad (1)$$

for $w \in L^2(\Omega)$ define the **weak gradient** by (1).

Galerkin formulation / method

find $u \in V$ such that

$$\underbrace{\int_{\Omega} \nabla u \cdot \nabla v \, dx}_{a(u,v)} = \underbrace{\int_{\Omega} f v \, dx}_{l(v)}, \quad \forall v \in \hat{V}$$

Galerkin's method

Find $u_{\Delta} \in V_{\Delta} \subset V$ such that

$$a(u_{\Delta}, v) = l(v), \quad \forall v \in \hat{V}_{\Delta} \subset \hat{V},$$

where $\dim(V_{\Delta}) = \dim(\hat{V}_{\Delta}) < \infty$.

Galerkin method

Choose a basis $\{\phi_j\}_{j=1}^n$ for V_Δ and $\{\varphi_j\}_{j=1}^n$ for \hat{V}_Δ

$$u_\Delta = \sum_j \xi_j \phi_j(x)$$

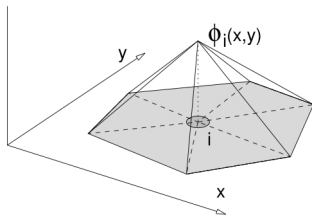
$$a(\sum_j \xi_j \phi_j, v) = l(v) \quad , \quad \forall v \in \hat{V}_\Delta$$

$$\sum_j a(\phi_j, \varphi_i) \xi_j = l(\varphi_i) \quad , \quad i = 1, 2, \dots, n$$

$$A \cdot \xi = b$$

A is symmetric, regular and positive definite.

Finite Elements



triangulation $\Omega_{\Delta} = \cup \Delta_k \subset \Omega$,

h_k longest side of Δ_k , $h = \max_k h_k$

interior vertices p_i , pyramid functions ϕ_j : $\phi_j(p_i) = \delta_{ij}$

$V_{\Delta} = \text{span} \{ \phi_1, \phi_2, \dots, \phi_N \} \subset H^1(\Omega)$

$v \in V_{\Delta}$: $v(x, y) = \sum_{j=1}^N v(p_j) \phi_j(x, y)$

$$u_{\Delta} = \sum_j \xi_j \phi_j(x)$$

$$\int_{\Omega_{\Delta}} \nabla u_{\Delta} \cdot \nabla v \, dx = \int_{\Omega_{\Delta}} f v \, dx, \quad \forall v \in \hat{V}_{\Delta}$$

$$A\xi = b$$

$$a_{ij} = a(\phi_j, \phi_i) = \int_{\Omega_{\Delta}} \nabla \phi_j \cdot \nabla \phi_i \, dx$$

$$b_i = l(\phi_i) = \int_{\Omega_{\Delta}} f \phi_i \, dx$$

automated finite elements: FEniCS

$$\underbrace{\int_{\Omega} \nabla u \cdot \nabla v \, dx}_{a(u,v)} = \underbrace{\int_{\Omega} f v \, dx}_{l(v)}, \quad \forall v \in \hat{V}$$

```
a = inner(grad(u), grad(v))*dx  
l = f*v*dx  
solve(a == l, u, bc)
```



FENICS
PROJECT



$$-\Delta u = f \quad , \quad u(0, y) = uD \quad , \quad \partial u / \partial n = g \quad \text{on } \Gamma$$

$$\int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx + \int_{\Gamma} g v \, dS \quad , \quad \forall v \in \hat{V}$$

```
bc = DirichletBC(V, uD, "near(x[0],0)")
a = inner(grad(u), grad(v))*dx
l = f*v*dx + g*v*ds
solve(a == l, u, bc)
```

the trace operator

Lemma 1: $C^1(\bar{\Omega})$ is dense in $H^1(\Omega)$.

Definition: Let $\Gamma \subset \partial\Omega$.

$$\text{Tr}: H^1(\Omega) \rightarrow L^2(\Gamma)$$

$$u \mapsto u|_{\Gamma}$$

$$u|_{\Gamma} := \lim_{n \rightarrow \infty} u_n|_{\Gamma}, \quad u_n \in C^1(\bar{\Omega})$$

Lemma 2: $\|u\|_{L^2(\Gamma)} \leq C\|u\|_{H^1(\Omega)}$

Corrolary: By Lemma 2, the trace is a unique extension of boundary values in $C^1(\bar{\Omega})$.

$$-\Delta u = f, \quad x \in \Omega$$

$$\partial u / \partial n = 0, \quad x \in \partial\Omega, \quad \int_{\Omega} u \, dx = 0$$

variational formulation: $-\int_{\Omega} \Delta u v \, dx = \int_{\Omega} f v \, dx$

Green's identity: $-\int_{\partial\Omega} v \frac{\partial u}{\partial n} \, dx + \underbrace{\int_{\Omega} \nabla u \cdot \nabla v \, dx}_{a(u,v)} = \underbrace{\int_{\Omega} f v \, dx}_{l(v)},$

weak formulation: $a(u, v) = l(v), \quad \forall v \in H^1(\Omega)$

Galerkin weak formulation

Find $u \in V$ such that $a(u, v) = l(v)$, $\forall v \in V$

where V is a Hilbert space, $l : V \mapsto \mathbb{R}$ is linear and $a : V \times V \mapsto \mathbb{R}$ is bi-linear.

Assumptions:

- ▶ **V-ellipticity**: $\exists k_1 > 0$ s.t. $a(v, v) \geq k_1 \|v\|_V^2$
- ▶ **continuity**:
 $\exists k_2$ s.t. $|a(v, w)| \leq k_2 \|v\|_V \|w\|_V$
 $\exists k_3$ s.t. $|l(v)| \leq k_3 \|v\|_V$
- ▶ and often: **symmetry** $a(u, v) = a(v, u)$

Lax–Milgram theorem (1954): Under these assumptions, the variational problem has a unique solution $u \in V$

$$\|u\|_V \leq k_3/k_1 \ .$$

Conclusion: The Poisson equation with Neumann boundary conditions has a unique, weak solution $u \in H^1(\Omega)$, $\int_{\Omega} u \, dx = 0$ satisfying the bound

$$\|u\|_{H^1} \leq (1 + C)\|f\|_{L^2} \ .$$

Cauchy–Schwarz inequality:

$$|(u, v)| \leq \|u\| \|v\|$$

Poincaré inequality:

If either $v \in H^1(\Omega)$ and $\int_{\Omega} v \, dx = 0$, or $v \in H_0^1(\Omega)$

$$\|v\|_{L^2} \leq C \|\nabla v\|_{L^2}$$

by Poincaré, Poisson is H^1 –elliptic ...

$a(u, v) = l(v)$, $\forall v \in V$, (bi-)linear, V -elliptic, and continuous

inner product: $a(\cdot, \cdot)$

(energy)-norm $\|v\|_a := \sqrt{a(v, v)}$

norm equivalence: $k_1 \|v\|_V^2 \leq a(v, v) \leq k_2 \|v\|_V^2$

Galerkin orthogonality: $a(u - u_\Delta, v) = 0$, $\forall v \in V_\Delta$

best approximation: $\|u - u_\Delta\|_a \leq \|u - v\|_a$, $\forall v \in V_\Delta$

Céa-Lemma (error estimate in the V -norm):

$$\|u - u_\Delta\|_V \leq \sqrt{k_2/k_1} \|u - v\|_V, \quad \forall v \in V_\Delta.$$

consider Poisson

$$\begin{aligned} -\Delta u &= f, & x \in \Omega \subset \mathbb{R}^n \\ u &= 0, & x \in \partial\Omega \end{aligned}$$

$\Omega \subset \mathbb{R}^n$ convex, smoothly bounded and $f \in L^2(\Omega)$

$$a(u, v) = \int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx = l(v), \quad \forall v \in H_0^1(\Omega)$$

by Lax–Milgram, there is a unique solution $u \in H_0^1(\Omega)$.

In fact $u \in H^2(\Omega) \cap H_0^1(\Omega)$ and $\|u\|_{H^2} \leq (1 + C)\|f\|_{L^2}$

recall piecewise linear interpolation $|f(t) - p_1(t)| \leq Ch^2 \|f''\|_\infty$

analogous:

$$I_h v(x) = \sum_j v(p_j) \phi_j(x) \in V_\Delta$$

$$\|v - I_h v\|_{L^2} \leq Ch^2 \|v\|_{H^2}$$

$$\|\nabla v - \nabla I_h v\|_{L^2} \leq Ch \|v\|_{H^2}$$

Theorem 1:

$$\|\nabla u - \nabla u_\Delta\|_{L^2} \leq \|\nabla u - \nabla(I_h u)\|_{L^2} \leq Ch \|u\|_{H^2}$$

dual problem

$$\begin{aligned}-\Delta w &= u - u_\Delta = e, & x \in \Omega \\ w &= 0, & x \in \partial\Omega\end{aligned}$$

weak form:

$$a(v, w) = l(v) = \int_{\Omega} e(x)v(x)dx, \quad \forall v \in H_0^1(\Omega)$$

pick $v = e$ and find $\|e\|_{L^2}^2 \leq \|\nabla e\|_{L^2} Ch \|w\|_{H^2}$

in fact $w \in H^2$ and $\|w\|_{H^2} \leq C \|e\|_{L^2}$

Theorem 2:

$$\|e\|_{L^2} \leq Ch \|\nabla u - \nabla u_\Delta\|_{L^2} = Ch^2 \|u\|_{H^2}$$

the FE system

$$-(p(x)u'(x))' + r(x)u(x) = f(x) \quad , \quad x \in (a, b)$$

$$u(a) = A \quad , \quad u(b) = B$$

$$p(x) \geq c_0 > 0 \quad , \quad r(x) \geq 0$$

weak formulation

$$\underbrace{\int_a^b p u' v' + r u v \, dx}_{a(u,v)} = \underbrace{\int_a^b f v \, dx}_{l(v)}$$

FE ansatz

$$u_\Delta(x) = A\phi_0(x) + \sum_{j=1}^{N-1} u_j \phi_j(x) + B\phi_N(x)$$

the FE system

$$\sum_{j=1}^N a(\phi_j, \phi_i) u_j = l(\phi_i) - a(\phi_0, \phi_i) A - a(\phi_N, \phi_i) B$$

$$a(\phi_j, \phi_i) = \int_a^b p \phi_j' \phi_i' + r \phi_j \phi_i \, dx \quad , \quad l(\phi_i) = \int_a^b f \phi_i \, dx$$

$$\begin{aligned} \frac{1}{h^2} \begin{pmatrix} p_{1/2} + p_{3/2} & -p_{3/2} & & \\ -p_{i-1/2} & p_{i-1/2} + p_{i+1/2} & -p_{i+1/2} & \\ & -p_{N-3/2} & p_{N-3/2} + p_{N-1/2} & \end{pmatrix} \begin{pmatrix} u_1 \\ u_i \\ u_{N-1} \end{pmatrix} + \\ + \begin{pmatrix} r_1 u_1 \\ r_i u_i \\ r_{N-1} u_{N-1} \end{pmatrix} = \begin{pmatrix} f_1 + p_{1/2} A / h^2 \\ f_i \\ f_{N-1} + p_{N-1/2} B / h^2 \end{pmatrix} \end{aligned}$$

Condition number

$$\kappa(A) = \|A\| \cdot \|A^{-1}\|$$

$$\frac{\|\delta x\|}{\|x\|} \leq \kappa(A) \frac{\|\delta b\|}{\|b\|} \quad , \quad \frac{\|\delta x\|}{\|x\|} \lesssim \kappa(A) \frac{\|\delta A\|}{\|A\|}$$

When solving a linear system $Ax = b$,
a loss of $k = \log_{10} \kappa(A)$ significant digits may be expected.

Dirichlet, Neumann and Robin conditions

$$-\Delta u = f \text{ in } \Omega$$

$$u = u_0 \text{ on } \Gamma_D$$

$$\nabla u \cdot n = g \text{ on } \Gamma_N$$

$$\nabla u \cdot n = p(q - u) \text{ on } \Gamma_R$$

$$\int_{\Omega} \nabla u \cdot \nabla v \, dx - \int_{\partial\Omega} \nabla u \cdot n v \, dS = \int_{\Omega} f v \, dx, \quad v = 0 \text{ on } \Gamma_D$$

$$\int_{\Omega} \nabla u \cdot \nabla v \, dx + \int_{\Gamma_R} p u v \, dS = \int_{\Omega} f v \, dx + \int_{\Gamma_N} g v \, dS + \int_{\Gamma_R} p q v \, dS$$

subdomains

let $\Omega = (0, 1)^2$ and $\Omega_0 = (0, 1) \times (0, 1/2)$

define Ω_0 :

```
class Omega0(SubDomain):  
    def inside(self, x, on_boundary):  
        return x[1] <= 0.5 + DOLFIN_EPS
```

create an instance:

```
subdomain0 = Omega0()
```

mark cells in subdomain:

```
subdomains = MeshFunction('size_t', mesh,  
    mesh.topology().dim())  
subdomain0.mark(subdomains, 0)
```

subdomains

```
# define new measures associated with the subdomains
```

```
dx = Measure('dx', domain=mesh,  
subdomain_data=subdomains)
```

```
# bilinear form:
```

```
a = k0*inner(grad(u), grad(v))*dx(0)  
    + k1*inner(grad(u), grad(v))*dx(1)
```


nonlinear Poisson

$$-\nabla \cdot (q(u) \nabla u) = 0 \rightsquigarrow \int_{\Omega} q(u) \nabla u \cdot \nabla v \, dx = 0 \quad \forall v$$

```
u = Function(V)    !!!
```

```
F = inner(q(u)*grad(u), grad(v))*dx
```

```
# Newton needs a derivative;
```

```
du = TrialFunction(V)
```

```
DF = derivative(F, u, du)
```

```
# compute solution:
```

```
problem = NonlinearVariationalProblem(F, u, bc, DF)
```

```
solver = NonlinearVariationalSolver(problem)
```

```
solver.solve()
```

Gâteaux (directional) derivative:

$$\begin{aligned} DF(u, \delta u, v) &= \lim_{\epsilon \rightarrow 0} \frac{F(u + \epsilon \delta u, v) - F(u, v)}{\epsilon} \\ &= \left. \frac{d}{d\epsilon} F(u + \epsilon \delta u, v) \right|_{\epsilon=0} \end{aligned}$$

Example: $F(u, v) = \int_{\Omega} q(u) \nabla u \cdot \nabla v \, dx$

concepts of error quantification

$$Ax = b, \quad A_h x_h = b_h$$

approximation error: $e = x - x_h$

truncation error: $\tau = A - h x - b_h = A_h e$

residual: $\rho = b - A x_h = A e$

a-priori analysis:

$$e = A_h^{-1} \tau, \quad \|e\| \leq \|A_h^{-1}\| \|\tau\|$$

a-posteriori analysis:

$$e = A^{-1} \rho, \quad \|e\| \leq \|A^{-1}\| \|\rho\|$$

concepts of error quantification

duality based analysis:

linear functional: $J : \mathbb{R}^n \rightarrow \mathbb{C}, x \mapsto J(x) = \langle x, j \rangle = j^* x$

dual problem:

$$A^* z = j$$

evaluation:

$$J(x) = \langle x, A^* z \rangle = \langle Ax, z \rangle = \langle b, z \rangle$$

$$J(x) - J(x_h) = J(e) = \langle Ae, z \rangle = \langle \rho, z \rangle$$

$$-\Delta u = f \quad \text{in } \Omega$$

$$u = 0 \quad \text{on } \Gamma_D$$

$$\nabla u \cdot n = g \quad \text{on } \Gamma_N$$

$$\int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx + \int_{\Gamma_N} g v \, dS$$

goal: $|M(u) - M(u_{\Delta})| \leq Tol$

goal oriented adaptivity

$$\begin{aligned}a(u, v) &= l(v) \ , \quad \forall v \in \widehat{V} \\ a(u_\Delta, v) &= l(v) \ , \quad \forall v \in \widehat{V}_\Delta\end{aligned}$$

adaptive problem: find V_Δ and u_Δ s.t. $|M(u) - M(u_\Delta)| \leq Tol$

dual problem: find $w \in \widehat{V}$ s.t. $a(v, w) = M(v)$, $\forall v \in V$

$$M(u) - M(u_\Delta) = a(u, w) - a(u_\Delta, w) = l(w) - a(u_\Delta, w) = r(w)$$

compute w and evaluate the **dual residual** $r(w)$

C. Johnson (1990), R. Rannacher (1995), ...

adaptivity in FEniCS

```
a = inner(grad(u), grad(v))*dx  
l = f*v*dx + g*v*ds  
solve(a == l, u, bc, tol=1e-5, M=u*dx)
```

THE WILKINSON PRIZE FOR NUMERICAL SOFTWARE 2015



FEniCS
PROJECT



conservation of mass:

$$\frac{d}{dt} \int_{\Omega} u(t, x) \, dx + \int_{\partial\Omega} f(u) n \, dS = 0$$

Fick's first law (1855): $f(u) = -D\nabla u$

divergence theorem

$$\int_{\Omega} u_t(t, x) - \nabla \cdot (D\nabla u) \, dx = 0$$

diffusion / heat equation

$$u_t = \nabla \cdot (D\nabla u) \, , \quad x \in \Omega \, , \quad t > 0$$

$$u_t = \Delta u + f$$

implicit time stepping

$$\frac{u^k - u^{k-1}}{\Delta t} = \Delta u^k + f^k$$

weak variational form

$$\int_{\Omega} u^k v + \Delta t \nabla u^k \cdot \nabla v \, dx = \int_{\Omega} \left(u^{k-1} + \Delta t f^k \right) v \, dx \quad , \quad \forall v \in H_0^1(\Omega)$$

diffusion

naive code:

```
while t <= T:  
    a = u*v*dx + dt*inner(grad(u), grad(v))*dx  
    l = (u_1 + dt*f)*v*dx  
    solve(a == l, u, bc)
```

naive code:

```
while t <= T:  
    a = u*v*dx + dt*inner(grad(u), grad(v))*dx  
    l = (u_1 + dt*f)*v*dx  
    solve(a == l, u, bc)
```

pre-assembled:

```
A = assemble(a)  
while t <= T:  
    b = assemble(l, tensor=b)  
    bc.apply(A, b)  
    solve(A, u.vector(), b)
```



time stepping

$$y' = f(y) \quad , \quad y(0) = y_0$$

explicit Euler

$$\frac{y^{k+1} - y^k}{\Delta t} = f(u^k)$$

implicit Euler

$$\frac{y^{k+1} - y^k}{\Delta t} = f(u^{k+1})$$

trapezoidal method (leap-frog)

$$\frac{y^{k+1} - y^k}{\Delta t} = \frac{1}{2} \left(f(u^{k+1}) + f(u^k) \right)$$

time stepping and stability

test equation: $y' = \lambda y$, $y(0) = 1$, $y(t) = \exp(\lambda t)$

time stepping: $y^{k+1} = y^k + \Delta t b_1 \lambda y^{k+1} + \Delta t b_0 \lambda y^k$

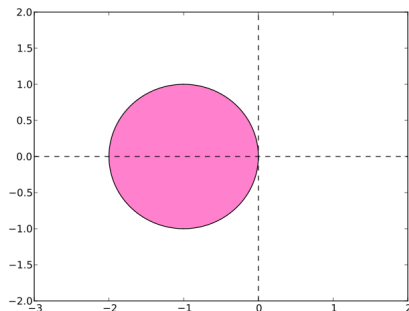
$$y^{k+1} = R(\Delta t \lambda) y^k , \quad R(z) = \frac{1 + b_0 z}{1 - b_1 z}$$

stability condition: $|R(z)| \leq 1$ for all $z \in \mathbb{C}^-$.

time stepping and stability

explicit Euler: $b_0 = 1$, $b_1 = 0$, $R(z) = 1 + z$

stability area: $S = \{z \in \mathbb{C} : |1 + z| \leq 1\}$

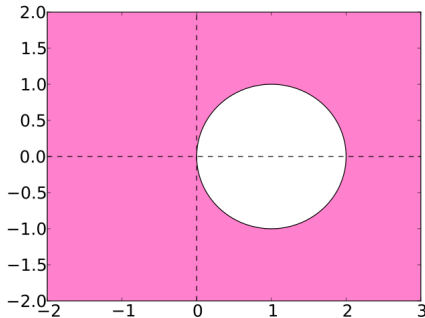


conditional stability: Δt small enough!

time stepping and stability

implicit Euler: $b_0 = 0$, $b_1 = 1$, $R(z) = \frac{1}{1-z}$

stability area: $S = \{z \in \mathbb{C} : 1 \leq |1 - z|\}$

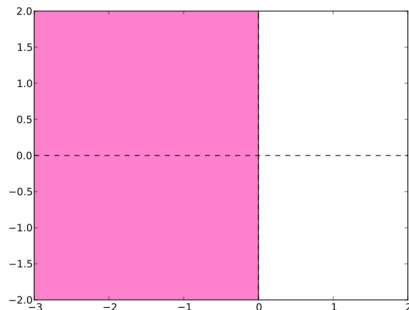


A-stability: $\mathbb{C}^- \subset S$

time stepping and stability

trapezoidal method: $b_0 = 1/2$, $b_1 = 1/2$, $R(z) = \frac{1+z/2}{1-z/2}$

stability area: $S = \{z \in \mathbb{C} : |1 + z/2| \leq |1 - z/2|\}$



A-stability: $S = \mathbb{C}^-$

time stepping and stability

test equation: $y' = \lambda y$, $y(0) = 1$, $y(t) = \exp(\lambda t)$

if $\operatorname{Re}(\lambda) \rightarrow -\infty$, then $y(t) = \exp(\lambda t) \rightarrow 0$.

discrete: $y^{k+1} = R(\Delta t \lambda) y^k$

An A-stable method with $\lim_{|z| \rightarrow \infty} R(z) = 0$ is called **L-stable**.

Remark: If $R(z) = \frac{P(z)}{Q(z)}$ and $\lim_{|z| \rightarrow \infty} R(z) < \infty$, then

$\deg(P) \leq \deg(Q)$ and

$$\lim_{x \rightarrow -\infty} R(x) = \lim_{x \rightarrow +\infty} R(x) = \lim_{x \rightarrow \pm\infty} R(ix) , \quad x \in \mathbb{R}$$

.

Remark: implicit Euler is L-stable.

Remark: explicit Euler and trapezoidal method are not.

Remark: any method with "perfect" stability area $S = \mathbb{C}^-$ is not L-stable.

s-stage RK-method

$$k_i = f(t_n + c_i h, y_n + h \sum_{j=1}^s a_{ij} k_j)$$

$$y_{n+1} = y_n + h \sum_{i=1}^s b_i k_i$$

Butcher tableau:
$$\begin{array}{c|c} c & A \\ \hline & b^T \end{array}$$

Singly Diagonal Implicit RK-methods:

$$\begin{array}{c|cc} \gamma & & \\ 1-\gamma & 1-2\gamma & \gamma \\ \hline & 1/2 & 1/2 \end{array}$$

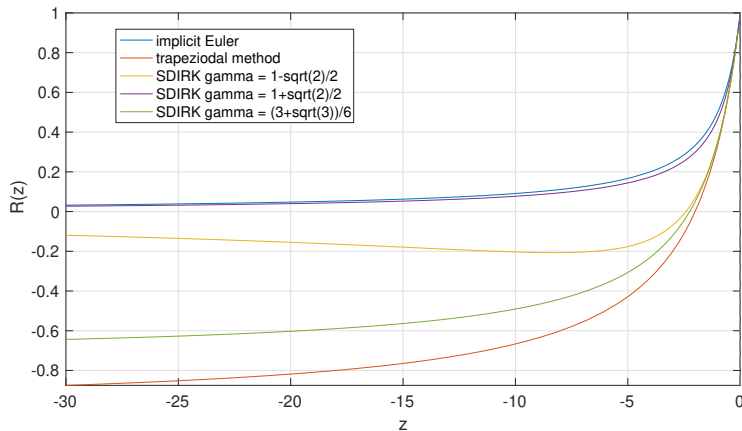
order $p \geq 2$ by construction; order $p = 3$: $\gamma = (3 \pm \sqrt{3})/6$

stability function:

$$R(z) = \frac{1 + z(1 - 2\gamma) + z^2(1/2 - 2\gamma + \gamma^2)}{(1 - \gamma z)^2}$$

SDIRK-method

stability functions along \mathbb{R}^-



A-stable: $\gamma \geq 1/4$

A-stable & order $p = 3$: $\gamma = (3 + \sqrt{3})/6$

L-stable: $\gamma = (2 \pm \sqrt{2})/2$

L-stable, monotone & order $p = 2$: $\gamma = (2 + \sqrt{2})/2$

wave equation as a system

$$u_{tt} = c^2 \Delta u, \quad \begin{pmatrix} u_t \\ v_t \end{pmatrix} = \begin{pmatrix} v \\ c^2 \Delta u \end{pmatrix}$$



FENICS
PROJECT

```
W = VectorFunctionSpace(mesh, "CG", 1)
(u,v) = TrialFunctions(W)
w = Function(W)
(u,v) = w.split()
```

conservation of mass:

$$\frac{d}{dt} \int_{\Omega} \rho(t, x) \, dx + \int_{\partial\Omega} f(\rho) n \, dS = 0$$

convective transport $f(\rho) = v\rho$, divergence theorem ...

$$\rho_t + \nabla \cdot (\rho v) = 0$$

incompressible flow: $\rho \equiv 1 \rightsquigarrow \nabla \cdot v = 0$

conservation of momentum:

$$\rho(v_t + v \cdot \nabla v) = -\nabla p + \nu \Delta v + f$$

incompressible Navier–Stokes (1822)

$$\begin{aligned} v_t + v \cdot \nabla v + \nabla p &= \nu \Delta v + f \\ \nabla \cdot v &= 0 \end{aligned}$$

remark: $v \cdot \nabla v = v \cdot (\nabla v) = (v \cdot \nabla)v$

Navier–Stokes

$$\frac{v^n - v^{n-1}}{\Delta t} + (v^{n-1} \cdot \nabla) v^{n-1} = -\nabla p^n + \nu \Delta v^{n-1} + f^{n-1}$$

Chorin's projection method (1968):

1. find a tentative velocity v^*

$$\frac{v^* - v^{n-1}}{\Delta t} + (v^{n-1} \cdot \nabla) v^{n-1} = \nu \Delta v^{n-1} + f^{n-1}$$

thus, $v^n - v^* = -\Delta t \nabla p^n$ and $\nabla \cdot v^n = \nabla \cdot v^* - \Delta t \Delta p^n = 0$

2. compute the pressure

$$\Delta t \Delta p^n = \nabla \cdot v^*$$

3. update the velocity

$$v^n = v^* - \Delta t \nabla p^n$$

Helmholtz–Hodge decomposition

Helmholtz decomposition (1858): Ω simply connected, $u : \Omega \rightarrow \mathbb{R}^n$

$$u = u_{div} + u_{rot} \quad , \quad \nabla \cdot u_{div} = 0 \quad , \quad \nabla \times u_{rot} = 0 \quad .$$

A conservative field $v = \nabla\phi$ is irrotational $\nabla \times \nabla\phi = 0$.

A irrotational field is conservative $u_{rot} = \nabla\phi$.

A divergence-free field has a vector potential $u_{div} = \nabla \times A$.

$$-\Delta\phi = \nabla \cdot u \quad , \quad u_{rot} = -\nabla\phi \quad , \quad u_{div} = u + \nabla\phi$$

scalar / vector potential: $u = -\nabla\phi + \nabla \times A$

$$\left(\frac{v^n - v^{n-1}}{\Delta t}, \phi \right) + (v^{n-1} \cdot \nabla v^{n-1}, \phi) + (\nabla p^n, \phi) + (\nu \nabla v^{n-1}, \nabla \phi) = (f^{n-1}, \phi)$$

Chorin in weak form:

1. prediction

$$\left(\frac{v^* - v^{n-1}}{\Delta t}, \phi \right) + (v^{n-1} \cdot \nabla v^{n-1}, \phi) + (\nu \nabla v^*, \nabla \phi) = (f^{n-1}, \phi)$$

2. pressure $\Delta t (\nabla p^n, \nabla \psi) = -(\nabla \cdot v^*, \psi)$

3. correction $(v^n, \phi) = (v^*, \phi) - \Delta t (\nabla p^n, \phi)$

demos: NS_Lpipe.py, NS_dolfin.py

$$\rho(v_t + v \cdot \nabla v) + \nabla p = \nu \Delta v + f, \quad \nabla \cdot v = 0$$

$$\Delta u = \nabla \cdot \nabla u = \nabla \cdot \left(\nabla u + (\nabla u)^T \right), \quad \nabla \cdot u = 0$$

symmetric gradient:

$$\epsilon(u) = \frac{1}{2} \left(\nabla u + (\nabla u)^T \right)$$

Cauchy stress:

$$\sigma(v, p) = 2\nu \epsilon(v) - pI$$

$$\rho(v_t + v \cdot \nabla v) = \nabla \cdot \sigma(v, p) + f, \quad \nabla \cdot v = 0$$

Navier–Stokes

$$\rho \left(\frac{v^n - v^{n-1}}{\Delta t} + v^{n-1} \cdot \nabla v^{n-1} \right) = \nabla \cdot \sigma(v^{n-1/2}, p^{n-1/2}) + f^{n-1/2}$$

Incremental Pressure Correction (1979):

1. tentative velocity v^*

$$\rho \left(\frac{v^* - v^{n-1}}{\Delta t} + v^{n-1} \cdot \nabla v^{n-1} \right) = \nabla \cdot \sigma(v^{n-1/2}, p^{n-3/2}) + f^{n-1/2}$$

for computing: $v^{n-1/2} = \frac{1}{2}(v^* + v^{n-1})$

2. pressure update

$$-\Delta t \Delta p^{n-1/2} = -\Delta t \Delta p^{n-3/2} - \rho \nabla \cdot v^*$$

3. velocity correction

$$\rho v^n = \rho v^* - \Delta t \nabla (p^{n-1/2} - p^{n-3/2})$$

partial integration

$$- \int_{\Omega} \nabla \cdot \sigma(v, p) \phi \, dx = \int_{\Omega} \sigma(v, p) : \epsilon(\phi) \, dx - \int_{\partial\Omega} \sigma(v, p) \cdot n \phi \, dS$$

tensor scalar product

$$A : B = \sum_i \sum_j a_{ij} b_{ij}$$

tentative velocity in weak form:

$$\begin{aligned} & \frac{\rho}{\Delta t} (v^* - v^{n-1}, \phi) + \rho (v^{n-1} \cdot \nabla v^{n-1}, \phi) \\ & + \left(\sigma(v^{n-1/2}, p^{n-3/2}) : \epsilon(\phi) \right) - \left(\sigma(v^{n-1/2}, p^{n-3/2}) \cdot n, \phi \right)_{\partial\Omega} \\ & = (f^{n-1/2}, \phi) \end{aligned}$$

free outflow boundary $\nabla v \cdot n = 0$:

$$\sigma(v, p) \cdot n = (2\nu\epsilon(v) - pl) \cdot n = \nu(\nabla v)^T \cdot n - pn$$

Dirichlet boundary (non-slip) $\phi = 0$.

IPC weak form:

1. prediction

$$\begin{aligned} & \frac{\rho}{\Delta t} (v^* - v^{n-1}, \phi) + (v^{n-1} \cdot \nabla v^{n-1}, \phi) \\ & + (\sigma(v^{n-1/2}, p^{n-3/2}) : \epsilon(\phi)) + ((p^{n-3/2} - \nu(\nabla v^{n-1/2})^T) \cdot n, \phi)_{\partial\Omega} \\ & = (f^{n-1/2}, \phi) \end{aligned}$$

2. incremental pressure update

$$\Delta t (\nabla p^{n-1/2}, \nabla \psi) = \Delta t (\nabla p^{n-3/2}, \nabla \psi) - \rho (\nabla \cdot v^*, \psi)$$

3. velocity correction

$$\rho (v^n, \phi) = \rho (v^*, \phi) - \Delta t (\nabla (p^{n-1/2} - p^{n-3/2}), \phi)$$

demo: NS_dolfin_IPC.py

incompressible flow with heat transfer

$$\begin{aligned}v_t + v \cdot \nabla v + \nabla p &= \frac{1}{Re} \Delta v + f \\ \nabla \cdot v &= 0 \\ T_t + v \cdot \nabla T &= \frac{1}{Pe} \Delta T + s\end{aligned}$$

Chorin's projection:

1. tentative velocity v^*

$$\frac{v^* - v^{n-1}}{\Delta t} + (V \cdot \nabla)v^{n-1} = \frac{1}{Re}\Delta v^* + f^{n-1}$$

2. pressure $\Delta t \Delta p^n = \nabla \cdot v^*$

3. velocity $v^n = v^* - \Delta t \nabla p^n$

4. temperature update

$$\frac{c^n - c^{n-1}}{\Delta t} + (V \cdot \nabla)c^{n-1} = \frac{1}{Pe}\Delta c^n + s^{n-1}$$

$$V = v^{n-1} \text{ or, alternatively } V = \frac{1}{2}(v^n + v^{n-1}).$$

Chorin in weak form:

1. prediction

$$\left(\frac{\mathbf{v}^* - \mathbf{v}^{n-1}}{\Delta t}, \phi \right) + (\mathbf{V} \cdot \nabla \mathbf{v}^{n-1}, \phi) + \left(\frac{1}{Re} \nabla \mathbf{v}^*, \nabla \phi \right) = 0$$

2. pressure $\Delta t (\nabla p^n, \nabla \psi) = -(\nabla \cdot \mathbf{v}^*, \psi)$

3. correction $(\mathbf{v}^n, \phi) = (\mathbf{v}^*, \phi) - \Delta t (\nabla p^n, \phi)$

4. temperature

$$\left(\frac{c^n - c^{n-1}}{\Delta t}, \psi \right) + (\mathbf{V} \cdot \nabla c^{n-1}, \psi) + \left(\frac{1}{Pe} \nabla c^n, \nabla \psi \right) = 0$$

demo: NST_dolfin.py

Navier–Stokes

IPC:

1. tentative velocity v^*

$$\frac{v^* - v^{n-1}}{\Delta t} + v^{n-1} \cdot \nabla v^{n-1} = \nabla \cdot \sigma(v^{n-1/2}, p^{n-3/2}) + f^{n-1/2}$$

2. pressure update

$$-\Delta t \Delta p^{n-1/2} = -\Delta t \Delta p^{n-3/2} - \rho \nabla \cdot v^*$$

3. velocity correction

$$v^n = v^* - \Delta t \nabla (p^{n-1/2} - p^{n-3/2})$$

4. temperature update

$$\frac{c^n - c^{n-1}}{\Delta t} + (v \cdot \nabla) c^{n-1} = \frac{1}{Pe} \Delta c^n + s^{n-1}$$

IPC weak form:

1. prediction

$$\begin{aligned} & \frac{1}{\Delta t} (v^* - v^{n-1}, \phi) + (v^{n-1} \cdot \nabla v^{n-1}, \phi) \\ & + (\sigma(v^{n-1/2}, p^{n-3/2}) : \epsilon(\phi)) + ((p^{n-3/2}) - \frac{1}{Re} (\nabla v^{n-1/2})^T) \cdot n, \phi)_{\partial\Omega} \\ & = (f^{n-1/2}, \phi) \end{aligned}$$

2. pressure update

$$\Delta t (\nabla p^{n-1/2}, \nabla \psi) = \Delta t (\nabla p^{n-3/2}, \nabla \psi) - \rho (\nabla \cdot v^*, \psi)$$

3. velocity correction

$$(v^n, \phi) = (v^*, \phi) - \Delta t (\nabla (p^{n-1/2} - p^{n-3/2}), \phi)$$

4. temperature update

$$\left(\frac{c^n - c^{n-1}}{\Delta t}, \psi \right) + (V \cdot \nabla c^{n-1}, \psi) + \left(\frac{1}{Pe} \nabla c^n, \nabla \psi \right) = 0$$

Chorin's pressure correction at higher order?

recall trapezoidal rule:

$$\frac{\Delta t}{2} (f(0) + f(h)) = \int_0^{\Delta t} f(t) \, dt + \mathcal{O}(\Delta t^3)$$

analogous

$$\frac{\Delta t}{2} (f(0)g(\Delta t) + f(\Delta t)g(0)) = \int_0^{\Delta t} (fg)(t) \, dt + \mathcal{O}(\Delta t^3)$$

Navier–Stokes

$$\frac{v^n - v^{n-1}}{\Delta t} + \int_{t_{n-1}}^{t_n} (v \cdot \nabla) v \, dt + \int_{t_{n-1}}^{t_n} \nabla p \, dt = \nu \int_{t_{n-1}}^{t_n} \Delta v \, dt$$

2. order projection method:

1. find a tentative velocity v^*

$$\frac{v^* - v^{n-1}}{\Delta t} + \frac{1}{2} (v^* \cdot \nabla v^{n-1} + v^{n-1} \cdot \nabla v^*) = \frac{\nu}{2} \Delta (v^* + v^{n-1})$$

thus, $v^n - v^* + \Delta t \nabla p^{n-1/2} = \mathcal{O}(\Delta t^3)$

2. compute the pressure

$$\Delta t \Delta p^{n-1/2} = \nabla \cdot v^*$$

3. update the velocity

$$v^n = v^* - \Delta t \nabla p^{n-1/2}$$

$$\frac{v^n - v^{n-1}}{\Delta t} + \int_{t_{n-1}}^{t_n} (v \cdot \nabla) v \, dt = \int_{t_{n-1}}^{t_n} \nabla \cdot \sigma(v, p) \, dt$$

2. order IPC method:

1. find a tentative velocity v^*

$$\frac{v^* - v^{n-1}}{\Delta t} + \frac{1}{2} (v^* \cdot \nabla v^{n-1} + v^{n-1} \cdot \nabla v^*) = \nabla \cdot \sigma(v^{n-1/2}, p^{n-3/2})$$

$$\text{thus, } v^n - v^* + \Delta t \nabla (p^{n-1/2} - p^{n-3/2}) = \mathcal{O}(\Delta t^3)$$

2. compute the pressure update

$$\Delta t \Delta \delta p = \nabla \cdot v^* \quad , \quad p^{n-1/2} = p^{n-3/2} + \delta p$$

3. update the velocity

$$v^n = v^* - \Delta t \nabla \delta p$$

nonlinear heat transfer $T_t + v \cdot \nabla T = \frac{1}{Pe} \nabla \cdot ((1 + T^2) \nabla T)$

$$\left(\frac{c^n - c^{n-1}}{\Delta t}, \psi \right) + (V \cdot \nabla c^{n-1}, \psi) + \frac{1}{Pe} ((1 + (c^n)^2) \nabla c^n, \nabla \psi) = 0$$

```
F = inner(c-c0, psi)*dx/dt \
    + inner(dot(V, grad(c0)), psi)*dx \
    + (1+c**2)*inner(grad(c), grad(psi))*dx/Pe
```

```
solve(F == 0, c, bc)
```

demo: NST_nonlin_dolfin.py

Lax & Riesz theorems

$$a(u, v) = l(v) \quad , \quad \forall v \in V$$

V Hilbert space, a and l are (bi-)linear, V -elliptic and continuous

Lax–Milgram theorem (1954):

There **is** a unique, bounded solution in V .

Riesz representation theorem (1907):

There is a unique solution $u \in V$ to

$$(u, v)_V = l(v) \quad , \quad \forall v \in V$$

and $\|u\| = \|l\|$.

Lax & Riesz theorems

Remark: $l : V \rightarrow \mathbb{R}$, linear i.e. $l \in V^*$. Riesz: $V \leftrightarrow V^*$

Lemma: The variational problem

$$(u, v)_V = l(v) \quad , \quad \forall v \in V$$

is equivalent to

$$u = \arg \min_{v \in V} J(v) \quad , \quad J(v) = \frac{1}{2}(v, v) - l(v)$$

calibration of diffusive PDE models

Find α in $u_t = \nabla \cdot (\alpha \nabla u)$ s.t. $J(u(\alpha)) = \int_D (\tilde{u} - u(\alpha))^2 \, dx = \min$

$$PDE(u, \alpha) = 0 \quad , \quad J(u(\alpha)) = \min \iff J_\alpha(u(\alpha)) = 0$$

Landweber algorithm (1951):

1. solve $PDE(u^k, \alpha^k)$ for u^k
2. evaluate $d^k = J_\alpha(u^k(\alpha^k))$
3. update $\alpha^{k+1} = \alpha^k - \Delta \alpha^k d^k$ (steepest descent)

gradients via duals

$$u_t = \nabla \cdot (\alpha \nabla u) \quad , \quad \int_{\Omega} u_t \phi + \alpha \nabla u \cdot \nabla \phi \, dx = 0$$

$$\mathcal{A}(u_\alpha, \phi) := \int_{\Omega} u_{\alpha,t} \phi + \alpha \nabla u_\alpha \cdot \nabla \phi \, dx = - \int_{\Omega} \nabla u \cdot \nabla \phi \, dx$$

$$J = \int_D (\tilde{u} - u)^2 \, dx \quad , \quad J_\alpha = -2 \int_D (\tilde{u} - u) u_\alpha \, dx$$

gradients via duals

$$u_t = \nabla \cdot (\alpha \nabla u) \quad , \quad \int_{\Omega} u_t \phi + \alpha \nabla u \cdot \nabla \phi \, dx = 0$$

$$\mathcal{A}(u_\alpha, \phi) := \int_{\Omega} u_{\alpha,t} \phi + \alpha \nabla u_\alpha \cdot \nabla \phi \, dx = - \int_{\Omega} \nabla u \cdot \nabla \phi \, dx$$

$$J = \int_D (\tilde{u} - u)^2 \, dx \quad , \quad J_\alpha = -2 \int_D (\tilde{u} - u) u_\alpha \, dx$$

$$\mathcal{A}^*(\omega, \phi) = \mathcal{A}(\phi, \omega) = -2 \int_D (\tilde{u} - u) \phi \, dx \quad , \quad \forall \phi$$

gradients via duals

$$u_t = \nabla \cdot (\alpha \nabla u) \quad , \quad \int_{\Omega} u_t \phi + \alpha \nabla u \cdot \nabla \phi \, dx = 0$$

$$\mathcal{A}(u_\alpha, \phi) := \int_{\Omega} u_{\alpha,t} \phi + \alpha \nabla u_\alpha \cdot \nabla \phi \, dx = - \int_{\Omega} \nabla u \cdot \nabla \phi \, dx$$

$$J = \int_D (\tilde{u} - u)^2 \, dx \quad , \quad J_\alpha = -2 \int_D (\tilde{u} - u) u_\alpha \, dx$$

$$\mathcal{A}^*(\omega, \phi) = \mathcal{A}(\phi, \omega) = -2 \int_D (\tilde{u} - u) \phi \, dx \quad , \quad \forall \phi$$

$$\mathcal{A}(u_\alpha, \omega) = -2 \int_D (\tilde{u} - u) u_\alpha \, dx = J_\alpha$$

gradients via duals

$$u_t = \nabla \cdot (\alpha \nabla u) \quad , \quad \int_{\Omega} u_t \phi + \alpha \nabla u \cdot \nabla \phi \, dx = 0$$

$$\mathcal{A}(u_\alpha, \omega) := \int_{\Omega} u_{\alpha,t} \omega + \alpha \nabla u_\alpha \cdot \nabla \omega \, dx = - \int_{\Omega} \nabla u \cdot \nabla \omega \, dx$$

$$J = \int_D (\tilde{u} - u)^2 \, dx \quad , \quad J_\alpha = -2 \int_D (\tilde{u} - u) u_\alpha \, dx$$

$$\mathcal{A}^*(\omega, \phi) = \mathcal{A}(\phi, \omega) = -2 \int_D (\tilde{u} - u) \phi \, dx \quad , \quad \forall \phi$$

$$\mathcal{A}(u_\alpha, \omega) = -2 \int_D (\tilde{u} - u) u_\alpha \, dx = J_\alpha = - \int_{\Omega} \nabla u \cdot \nabla \omega \, dx$$

$$\mathcal{A}^*(\omega, \phi) = -2 \int_D (\tilde{u} - u) \phi \, dx \, , \quad \forall \phi$$

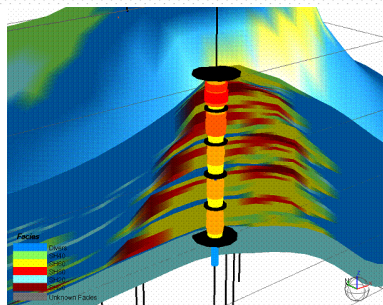
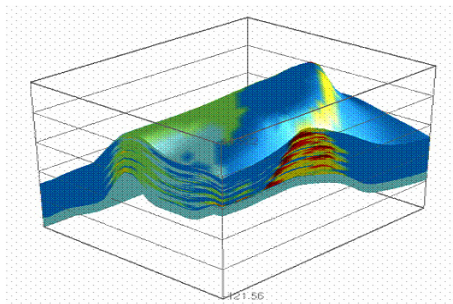
dual problem

$$\mathcal{A}^*(\omega, \phi) = -2 \int_D (\tilde{u} - u) \phi \, dx, \quad \forall \phi$$

$$\int_{\Omega} \phi_t \omega + \alpha \nabla \phi \cdot \nabla \omega \, dx = -2 \int_D (\tilde{u} - u) \phi \, dx, \quad \forall \phi$$

$$-\omega_t = \nabla \cdot (\alpha \nabla \omega) - 2(\tilde{u} - u)|_D, \quad t : T \rightarrow 0$$

depositional modelling



$$\begin{pmatrix} A & s \\ -A & 1-s \end{pmatrix} \begin{pmatrix} s \\ h \end{pmatrix}_t = \nabla \cdot \begin{pmatrix} \alpha s \nabla h \\ \beta (1-s) \nabla h \end{pmatrix}$$

well-output: $J(p) = \frac{1}{|W|} \int_0^T \int_W (\tilde{u}(t, x) - u(p, t, x))^2 dx dt$

forward problem:

- ▶ mapping: parameter \rightarrow observation
- ▶ transport coefficients \rightarrow well-output

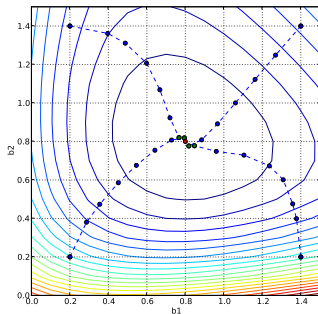
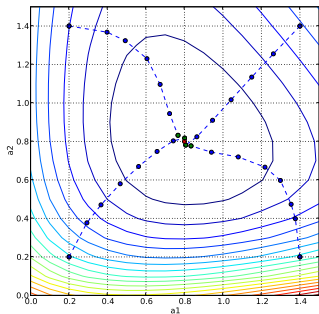
$$\begin{pmatrix} A & s \\ -A & 1-s \end{pmatrix} \begin{pmatrix} s \\ h \end{pmatrix}_t = \nabla \cdot \begin{pmatrix} \alpha s \nabla h \\ \beta (1-s) \nabla h \end{pmatrix}$$

well-output: $J(p) = \frac{1}{|W|} \int_0^T \int_W (\tilde{u}(t, x) - u(p, t, x))^2 dx dt$

inverse problem:

- ▶ inv. mapping: parameter \leftarrow observation
- ▶ transport coefficients \leftarrow well-output

the output functional



the output functional

$$u_t = d\Delta u \quad , \quad d > 0$$

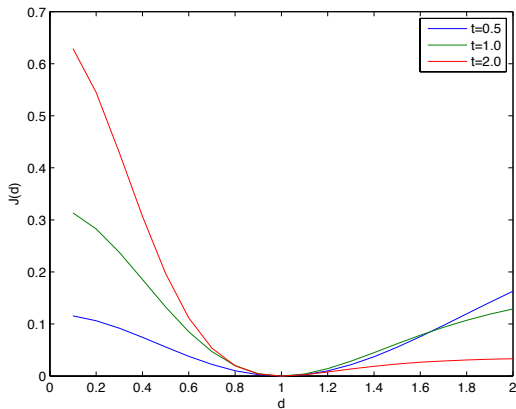
$$K(td, x) = \frac{1}{(4\pi td)^{n/2}} e^{-|x|^2/4td} \quad , \quad x \in \mathbb{R}^n$$

$$u(td, x) = \int K(td, x - y) f(y) \, dy \quad , \quad u(0, x) = f(x)$$

$$J(d) = \int (u(td, x) - u(t, x))^2 \, dx$$

$$J(1) = 0 \quad , \quad J'(1) = 0 \quad , \quad J''(1) = 2t^2 \|\Delta u(t, x)\|^2 > 0$$

the output functional



$$u(t, x, d) = e^{-d^2 t} \sin(\sqrt{d/2} x_1) \cos(\sqrt{d/2} x_2)$$

outlook – automated, adaptive calibration

$$u_t = \nabla \cdot (\alpha \nabla u) + f$$

$$\int_{\Omega} u_t \phi + \alpha \nabla u \cdot \nabla \phi \, dx = \int_{\Omega} f \phi \, dx, \quad \forall \phi$$

$$a(\alpha, u, \phi) = l(\phi), \quad \forall \phi$$

misfit functional:

$$J(\alpha, u) = \|\tilde{u} - u\|^2 + \eta \|\alpha\|^2 \rightarrow \min$$

Lagrange functional:

$$L(\alpha, u, \omega) = \underbrace{J(\alpha, u)}_{\rightarrow \min} - \underbrace{a(\alpha, u, \omega) + l(\omega)}_{\rightarrow 0}$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

primal problem ∂_ω :

$$-a_\omega(\alpha, u, \omega)(\phi) + l_\omega(\omega)(\phi) = 0 \quad , \quad \forall \phi$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

primal problem ∂_ω :

$$-a_\omega(\alpha, u, \omega)(\phi) + l_\omega(\omega)(\phi) = 0 \quad , \quad \forall \phi$$

$$a(\alpha, u, \phi) = l(\phi) \quad , \quad \forall \phi$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

primal problem ∂_ω :

$$-a_\omega(\alpha, u, \omega)(\phi) + l_\omega(\omega)(\phi) = 0 \quad , \quad \forall \phi$$

$$a(\alpha, u, \phi) = l(\phi) \quad , \quad \forall \phi$$

$$\int_{\Omega} u_t \phi + \alpha \nabla u \cdot \nabla \phi \, dx = \int_{\Omega} f \phi \, dx \quad , \quad \forall \phi$$

$$u_t = \nabla \cdot (\alpha \nabla u) + f$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

dual problem ∂_u :

$$J_u(\alpha, u)(\psi) - a_u(\alpha, u, \omega)(\psi) = 0 \quad , \quad \forall \psi$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

dual problem ∂_u :

$$J_u(\alpha, u)(\psi) - a_u(\alpha, u, \omega)(\psi) = 0 \quad , \quad \forall \psi$$

$$a(\alpha, \psi, \omega) = -2 \int_{\Omega} (\tilde{u} - u) \psi \, dx \quad , \quad \forall \psi$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + I(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

dual problem ∂_u :

$$J_u(\alpha, u)(\psi) - a_u(\alpha, u, \omega)(\psi) = 0 \quad , \quad \forall \psi$$

$$a(\alpha, \psi, \omega) = -2 \int_{\Omega} (\tilde{u} - u) \psi \, dx \quad , \quad \forall \psi$$

$$\int_{\Omega} \psi_t \omega + \alpha \nabla \psi \cdot \nabla \omega \, dx = -2 \int_{\Omega} (\tilde{u} - u) \psi \, dx \quad , \quad \forall \psi$$

$$-\omega_t = \nabla \cdot (\alpha \nabla \omega) - 2(\tilde{u} - u) \quad , \quad t : T \rightarrow 0$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

control problem ∂_α :

$$J_\alpha(\alpha, u)(\chi) - a_\alpha(\alpha, u, \omega)(\chi) = 0 \quad , \quad \forall \chi$$

outlook – automated, adaptive calibration

$$L(\alpha, u, \omega) = J(\alpha, u) - a(\alpha, u, \omega) + l(\omega)$$

optimality system, $L'(\alpha, u, \omega) = 0$:

control problem ∂_α :

$$J_\alpha(\alpha, u)(\chi) - a_\alpha(\alpha, u, \omega)(\chi) = 0 \quad , \quad \forall \chi$$

$$2\eta \int_{\Omega} \alpha \chi \, dx = \int_{\Omega} \nabla u \cdot \nabla \omega \chi \, dx \quad , \quad \forall \chi$$

$$\chi = 1 : \quad 2\eta \int_{\Omega} \alpha \, dx = \int_{\Omega} \nabla u \cdot \nabla \omega \, dx$$

conclusion: there is a free lunch!

solve

1. primal
 2. dual (linear problem)
 3. control
-
- ▶ *self calibrating, adaptive models*
 - ▶ *adaptivity for free!*