

Lecture 1: A Review of Goal-Oriented Error Estimation and Adaptive Methods

Serge Prudhomme

Département de mathématiques et de génie industriel
Ecole Polytechnique de Montréal

SRI Center for Uncertainty Quantification
King Abdullah University of Science and Technology

eVITA Winter School 2015
Uncertainty Quantification for Physical Phenomena
Geilo, Norway, January 18-23, 2015



Plan of Lectures



Plan of Lectures

Lecture 1:

A Review of Goal-Oriented Error Estimation and Adaptive Methods

Lecture 2:

Error Estimation and Control for Problems with Uncertain Coefficients

Lecture 3:

Adaptive Construction of Response Surface Approximations for Bayesian Inference

Outline

1) Introduction

2) Estimation and control of discretization errors in quantities of interest

- Linear problems (adjoint and error representation).
- Extension to nonlinear problems.
- Extension to time-dependent problems.
- Multiphysics coupled problems: Micro-fluidics application.

3) Conclusions

Introduction

Error estimation is useful for two purposes:

1. To provide a measure of the accuracy in approximations.
2. To control errors in those approximations (for example via mesh adaptation in the case of finite element solutions).

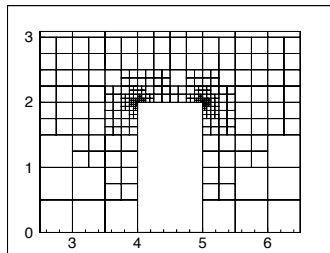
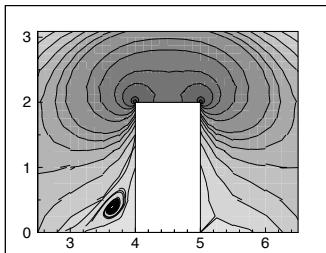
Error estimates have been constructed with respect to:

- a) Norms associated with the solution function spaces.
- b) Quantities of interest \Rightarrow goal-oriented error estimation.

Flow around obstacle (Stokes flow)

Top plots:

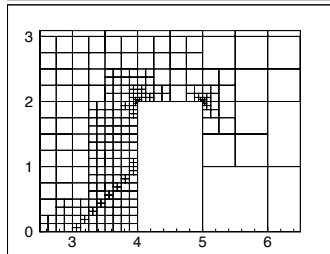
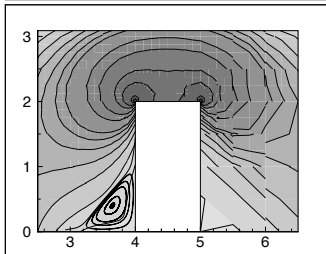
Residual-based
Error Estimation
and adaptivity



Bottom plots:

Goal-Oriented
Error Estimation
and adaptivity

QoI = averaged
vorticity in lower
left corner.



Catenary (linearized) model:

$$\begin{aligned} -Tu'' &= -\rho g, & \text{in } \Omega = (0, 1) \\ u &= u_0, & \text{at } x = 0 \\ u &= u_1, & \text{at } x = 1 \end{aligned}$$



Exact solution:

$$u(x) = -\frac{\rho g}{2T}x(1-x) + u_0(1-x) + u_1x$$

Quantity of interest:

Deflection at center point:	$Q(u) = u(0.5)$	$= \left[\frac{u_0 + u_1}{2} \right] - \frac{\rho g}{8T}$
Average deflection:	$Q(u) = \int_0^1 u(x)dx$	$= \left[\frac{u_0 + u_1}{2} \right] - \frac{\rho g}{12T}$
Slope at origin:	$Q(u) = u'(0)$	$= [u_1 - u_0] - \frac{\rho g}{2T}$

Green's function

Suppose that $u_0 = u_1 = 0$, i.e.

$$\begin{aligned} -Tu'' &= -\rho g, & \text{in } \Omega = (0, 1) \\ u &= 0, & \text{at } x = 0 \\ u &= 0, & \text{at } x = 1 \end{aligned}$$

Weak formulation:

$$\begin{aligned} &\text{Given } \rho g, \\ &\text{Find } u \in V = H_0^1(0, 1) \text{ s.t.} \\ &B(u, v) = F(v) \quad \forall v \in V \end{aligned}$$

where

$$\begin{cases} B(u, v) = \int_0^1 Tu'v' dx \\ F(v) = \int_0^1 -\rho g v dx \end{cases}$$

Green's function

Let $Q(u) = u(x_0)$.

The **Green function** is the function $G_0 = G_0(x) \in V$ that satisfies:

$$Q(u) = u(x_0) = \int_0^1 -\rho g G_0 dx = F(G_0)$$

that is:

$$Q(u) = F(G_0) = B(u, G_0)$$

Note that:

$$Q(u) = u(x_0) = \int_0^1 u(x) \delta(x - x_0) dx$$

Green's function

$$Q(u) = B(u, G_0) = F(G_0)$$

Primal problem:

$$\begin{aligned} -Tu'' &= -\rho g & \text{in } (0, 1) \\ u &= 0 & \text{at } x = 0, 1 \end{aligned}$$

Weak form:

$$\begin{aligned} &\text{Given } \rho g, \text{ find } u \in V \text{ s.t.} \\ &B(u, v) = F(v) \quad \forall v \in V \end{aligned}$$

Provide Qol:

$$Q(u) = \int_0^1 u(x)\delta(x - x_0)dx$$

Adjoint problem:

Find $G_0 \in V$ s.t.

$$B(v, G_0) = Q(v) \quad \forall v \in V$$

Strong form:

$$\int_0^1 T v' G_0' dx = \int_0^1 v \delta dx$$

so that

$$\begin{aligned} -TG_0'' &= \delta & \text{in } (0, 1) \\ G_0 &= 0 & \text{at } x = 0, 1 \end{aligned}$$

Generalized Green Function

Abstract linear BVP:

$$\text{Find } u \in V, \quad B(u, v) = F(v), \quad \forall v \in V$$

Quantity of interest:

$$Q(u) = \int_{\Omega} u(x)k(x)dx$$

Adjoint problem:

$$\text{Find } z \in V, \quad B(v, z) = Q(v), \quad \forall v \in V$$

FE approximation: Let $V^h \subset V$

$$\text{Find } u_h \in V^h, \quad B(u_h, v_h) = F(v_h), \quad \forall v_h \in V^h$$

Error equation:

$$\text{Find } e \in V, \quad B(e, v_h) = \mathcal{R}(u_h; v) \equiv F(v) - B(u_h, v), \quad \forall v \in V$$

Error Representation

Goal is to estimate $\mathcal{E} = Q(u) - Q(u_h)$

$$\mathcal{E} = B(u, z) - B(u_h, z) \quad (\text{From adjoint problem})$$

$$= F(z) - B(u_h, z) \quad (\text{From primal problem})$$

$$= \mathcal{R}(u_h; z) \quad (\text{From definition of residual})$$

$$= \mathcal{R}(u_h; z - z_h) \quad (\text{From orthogonality property})$$

$$\mathcal{E} = Q(u - u_h) = Q(e) \quad (\text{From linearity of } Q)$$

$$= B(e, z) \quad (\text{From adjoint problem})$$

$$= B(e, z - z_h) \quad (\text{From orthogonality property})$$

$$= \mathcal{R}(u_h; z - z_h) \quad (\text{From error equation})$$

Adaptive strategies

Let \tilde{z} be a higher-order approximation of the adjoint solution on same mesh as u_h , i.e.

$$\text{Find } \tilde{z} \in V^{h,p+1}, \quad B(v, \tilde{z}) = Q(v), \quad \forall v \in V^{h,p+1}$$

Approach 1: Using only \tilde{z} and $\tilde{z}_h = \Pi^{h,p} \tilde{z}$

$$\mathcal{E} \approx \eta = \mathcal{R}(u_h; \tilde{z} - \tilde{z}_h) = \sum_K \mathcal{R}_K(u_h; \tilde{z} - \tilde{z}_h)$$

Approach 2: Using \tilde{z} , but also $\tilde{u} \in V^{h,p+1}$,

$$\mathcal{E} \approx \eta = B(\tilde{u} - u_h, \tilde{z} - \tilde{z}_h) = \sum_K B_K(\tilde{u} - u_h, \tilde{z} - \tilde{z}_h)$$

Strategy favored by [Demkowicz et al.](#) in hp -adaptation scheme.

Adaptive Strategies

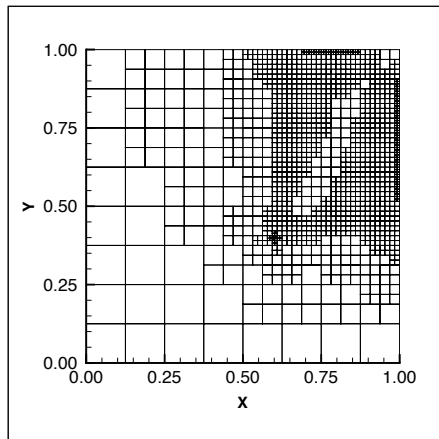
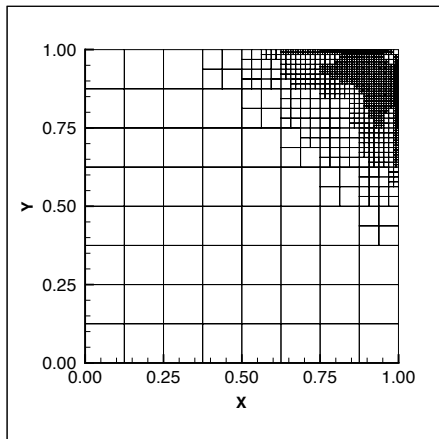
Adaptive scheme (catenary exemple):

$$\begin{aligned}
 \eta &\approx \mathcal{R}(u_h; \tilde{z} - \tilde{z}_h) = F(\tilde{z} - \tilde{z}_h) - B(u_h, \tilde{z} - \tilde{z}_h) \\
 &= \int_0^1 -\rho g (\tilde{z} - \tilde{z}_h) dx - \int_0^1 T u_h' (\tilde{z} - \tilde{z}_h)' dx \\
 &= \sum_{K=1}^{N_e} \left[\int_K -\rho g (\tilde{z} - \tilde{z}_h) + T u_h'' (\tilde{z} - \tilde{z}_h) dx - \int_{\partial K} T u_h' (\tilde{z} - \tilde{z}_h) ds \right] \\
 &= \underbrace{\sum_{K=1}^{N_e} \left[\int_K (-\rho g + T u_h'') (\tilde{z} - \tilde{z}_h) dx - \sum_{i=1,2} \frac{1}{2} [[T u_h']] (\tilde{z} - \tilde{z}_h)|_{x_i^K} \right]}_{\eta_K}
 \end{aligned}$$

Refinement criterion:

$$\text{If } \frac{|\eta_K|}{\max_K |\eta_K|} \geq \gamma_{\text{tol}}, \text{ then refine element } K$$

Example: Elliptic problem



Optimal?

Nonlinear Problems

Abstract nonlinear problem:

$$\text{Find } u \in V, \quad B(u; v) = F(v), \quad \forall v \in V$$

where $V =$ Banach space and $B(\cdot; \cdot) =$ differentiable semilinear form.
Let $Q(u)$ denote a possibly nonlinear differentiable functional on V .

Linearization (Taylor with exact remainder):

$$\begin{aligned} B(u + w; v) &= B(u; v) + B'(u; w, v) + \Delta_B(u, w, v) \\ Q(u + w) &= Q(u) + Q'(u; w) + \Delta_Q(u, w) \end{aligned}$$

$$\begin{aligned} \Delta_B(u, w, v) &= \int_0^1 B''(u + sw; w, w, v)(1 - s) ds \\ \Delta_Q(u, w) &= \int_0^1 Q''(u + sw; w, w)(1 - s) ds \end{aligned}$$

Secant Form (exact) Representation

$$B(u; v) = B(u_h; v) + \underbrace{\int_0^1 B'(su + (1-s)u_h; e, v) ds}_{\equiv B^s(u, u_h; e, v) = \text{secant form of B}}, \quad \forall v \in V$$

Then

$$\begin{aligned} Q(u) - Q(u_h) &= Q(e) = B^s(u, u_h; e, z) \\ &= B(u; z) - B(u_h; z) \\ &= F(z) - B(u_h; z) \\ &= \mathcal{R}(u_h; z) \end{aligned}$$

where z is the solution of the dual problem:

Find $z \in V$ such that $B^s(u, u_h; v, z) = Q(v), \quad \forall v \in V$

Optimal Approach

Constrained minimization problem*

Find $u \in V$ such that

$$Q(u) = \inf_{v \in M} Q(v)$$

where

$$M = \{v \in V; B(v; q) = F(q), \forall q \in V\}$$

Lagrangian:

$$L(u, z) = Q(u) + F(z) - B(u; z)$$

Here z = influence function (or Lagrange multiplier or adjoint solution) corresponding to the choice Q of the quantity of interest.

*Becker & Rannacher (2001).

Optimal Approach

The critical points (u, p) of $L(\cdot, \cdot)$ satisfy:

$$L'((u, z); (v, q)) = 0 \quad \forall (v, q) \in V \times V$$

$$L'((u, z); (v, q)) = \underbrace{Q'(u; v) - B'(u; v, z)}_{\Rightarrow \text{adjoint}} + \underbrace{F(q) - B(u; q)}_{\Rightarrow \text{primal}}$$

where

$$B'(u; v, z) = \lim_{\theta \rightarrow 0} \theta^{-1} [B(u + \theta v; z) - B(u; z)]$$

$$Q'(u; v) = \lim_{\theta \rightarrow 0} \theta^{-1} [Q(u + \theta v) - Q(u)]$$

Primal and adjoint problems:

$$\begin{aligned} B(u; v) &= F(v), & \forall q \in V \\ B'(u; v, z) &= Q'(u; v), & \forall v \in V \end{aligned}$$

Optimal Approach

Approximation of primal problem:

$$\text{Find } u_h \in V^h \text{ such that } B(u_h; v_h) = F(v_h) \quad \forall v_h \in V^h$$

Approximation of adjoint problem:

Define FE space \tilde{V}^h such that $V^h \subset \tilde{V}^h$. Then

$$\text{Find } \tilde{z}_h \in \tilde{V}^h \text{ such that } B'(u_h; v_h, \tilde{z}_h) = Q'(u_h; v_h), \quad \forall v_h \in \tilde{V}^h$$

Error Estimation:

Define residual as before, i.e. $\mathcal{R}(u_h; v) = F(v) - B(u_h; v)$. Then:

$$Q(u) - Q(u_h) = \mathcal{R}(u_h; \tilde{z}_h) + \Delta \approx \mathcal{R}(u_h; \tilde{z}_h)$$

where Δ represents higher-order terms in $u - u_h$ and $z - \tilde{z}_h$.

Optimal Approach

Expression for the remainder Δ :

$$\begin{aligned}\Delta &= \frac{1}{2} \int_0^1 B''(u_h + se; e, e, z_h + s\varepsilon) ds \\ &\quad - \frac{1}{2} \int_0^1 Q''(u_h + se; e, e) ds \\ &\quad - \frac{3}{2} \int_0^1 B''(u_h + se; e, e, \varepsilon)(s-1) s ds \\ &\quad - \frac{1}{2} \int_0^1 B'''(u_h + se; e, e, z_h + s\varepsilon)(s-1) s ds \\ &\quad + \frac{1}{2} \int_0^1 Q'''(u_h + se; e, e, e)(s-1) s ds\end{aligned}$$

Linearization approach: Error Equation, Adjoint problem

Error Equation: Let $e = u - u_h$

$$B(u; v) = B(u_h + e; v) = B(u_h; v) + B'(u_h; e, v) + \Delta_B(u_h, e, v) = F(v)$$

So

$$\text{Find } e \in V, \quad B'(u_h; e, v) + \Delta_B(u_h, e, v) = \mathcal{R}(u_h; v), \quad \forall v \in V$$

Dropping higher-order terms, error can be approximated by:

$$\text{Find } \hat{e} \in V, \quad B'(u_h; \hat{e}, v) = \mathcal{R}(u_h; v), \quad \forall v \in V$$

Adjoint problem:

$$\text{Find } z \in V \text{ such that } B'(u_h; v, z) = Q'(u_h; v), \quad \forall v \in V$$

Linearization approach

Representation of the error:

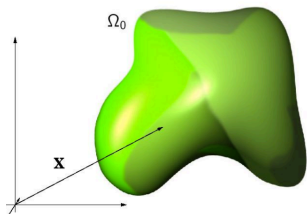
$$\begin{aligned}
 \mathcal{E} &= Q(u) - Q(u_h) = Q'(u_h; e) + \Delta_Q(u_h, e) \\
 &= B'(u_h; e, z) + \Delta_Q(u_h, e) \\
 &= \mathcal{R}(u_h; z) + \Delta_Q(u_h, e) - \Delta_B(u_h, e, z) \\
 &= \underbrace{\mathcal{R}(u_h; z - z_h)}_{\text{Discretization error}} + \underbrace{\Delta_Q(u_h, e) - \Delta_B(u_h, e, z)}_{\text{Linearization error}}
 \end{aligned}$$

with

$$\begin{aligned}
 \Delta_B(u_h, e, z) &= \int_0^1 B''(u_h + se; e, e, z)(1 - s)ds \\
 \Delta_Q(u_h, e) &= \int_0^1 Q''(u_h + se; e, e)(1 - s)ds
 \end{aligned}$$

Tumor Growth Model (Cahn-Hilliard)

Mixture theory: diffuse-interface phase-field model

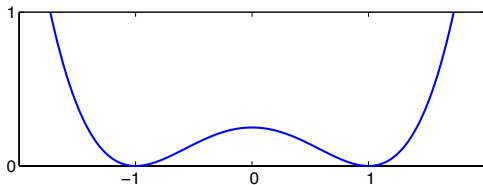


The tumor cell concentration u satisfies

$$\begin{aligned} u_t &= \Delta \mu(u) + g(\sigma, u) && \text{in } \Omega \\ \mu(u) &= f'(u) - \epsilon^2 \Delta u && \text{in } \Omega \\ \partial_n u &= \partial_n \mu = 0 && \text{on } \partial \Omega \\ u(0) &= u_0 && \text{in } \Omega \end{aligned}$$

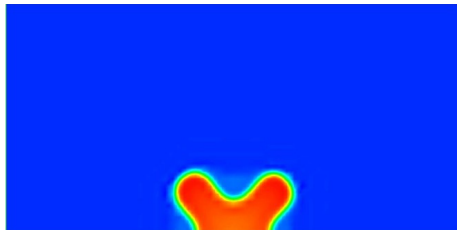
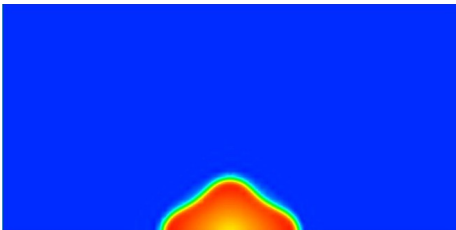
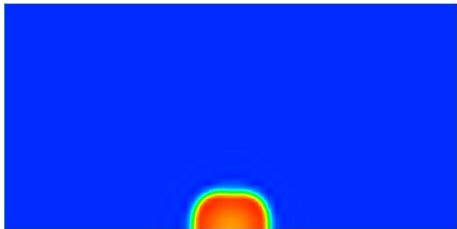
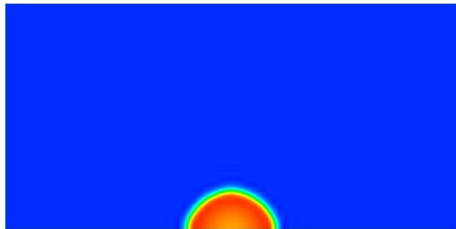
Nonlinear free energy density
 $f(u)$ drives phase separation

$$f(u) = \frac{1}{4}(u^2 - 1)^2$$

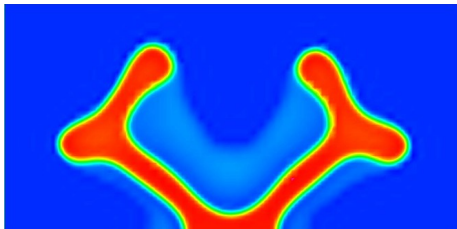
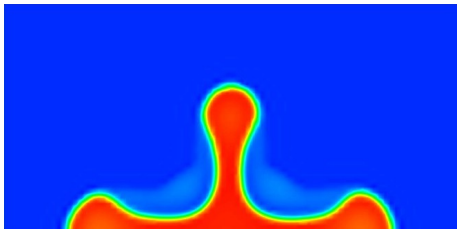
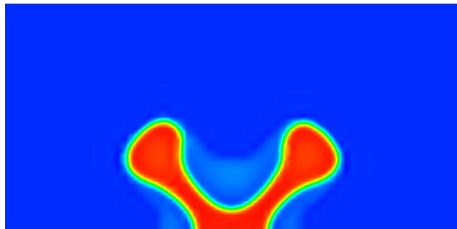
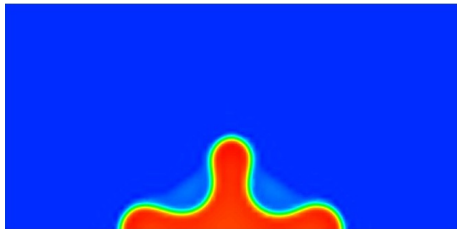


Lowengrub et al., NL (2010), Cristini et al., JMB (2009), Oden et al, M3AS (2010)

Model is sensitive!



Model is sensitive!



Theorem (Extension of Céa's Lemma to Nonlinear Problems)

Assume:

- B is differentiable.
- u is a nonsingular solution.
- Residual is Lipschitz continuous with constant c_L
- Linearized problem is well posed with discrete inf-sup constant $\gamma_{B'_h}(u)$

$$\|u - u_h\|_V \leq \left(1 + \frac{c_{B'}(u)}{\gamma_{B'_h}(u)}\right) \inf_{\phi \in V^h} \|u - \phi\|_V + \frac{c_L}{2\gamma_{B'_h}(u)} \|u - u_h\|_V^2$$

where $c_{B'}(u)$ is the continuity constant for $B'(u; \cdot, \cdot)$

Corollary: if $\|u - u_h\|_V \leq \gamma_{B'_h}(u)/c_L$

$$\|u - u_h\|_V \leq 2 \left(1 + \frac{c_{B'}(u)}{\gamma_{B'_h}(u)}\right) \inf_{\phi \in V^h} \|u - \phi\|_V$$

Burgers' problem

Let $\Omega = (-1, 1)$. Function $u(x) = -\tanh(x/2\mu)$ is exact solution of:

$$\mu u_{xx} + uu_x = 0, \quad x \in \Omega$$

$$u = u_- \quad \text{at } x = -1$$

$$u = u_+ \quad \text{at } x = 1$$

Find $u \in V$ such that $B(u; v) = F(v), \quad \forall v \in V$

Find $z \in V$ such that $B'(u; v, z) = Q'(u; v), \quad \forall v \in V$

where

$$B(u; v) = \int_{\Omega} \mu u_x v_x - \frac{1}{2} u^2 v_x \, dx \quad Q(u) = \frac{1}{2} \int_{\Omega} u^2 \, dx$$

$$B'(u; w, v) = - \int_{\Omega} u w v_x \, dx \quad Q'(u; w) = \int_{\Omega} u w \, dx$$

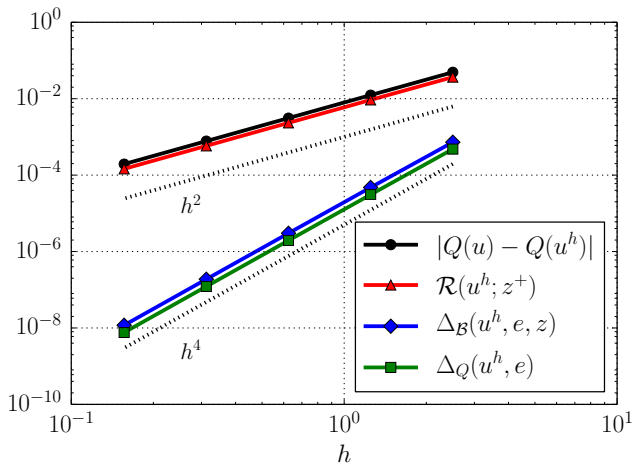
Burgers' problem – Convergence estimates

If z_x is continuous,

$$\begin{aligned}
 |\Delta_B(u_h, e, z)| &= \left| \int_0^1 \left(\int_{\Omega} e^2 z_x dx \right) (1-s) ds \right| \\
 &\leq \frac{1}{2} \int_{\Omega} |e^2| |z_x| dx \\
 &\leq \|e\|_{L^2(\Omega)}^2 \|z_x\|_{L^\infty} \quad \sim \mathcal{O}(h^4)
 \end{aligned}$$

$$\begin{aligned}
 |\Delta_Q(u_h, e)| &= \left| \int_0^1 \left(2 \int_{\Omega} e^2 dx \right) (1-s) ds \right| \\
 &\leq \int_{\Omega} |e|^2 dx \\
 &\leq \|e\|_{L^2(\Omega)}^2 \quad \sim \mathcal{O}(h^4)
 \end{aligned}$$

Burgers' problem – Convergence of error terms



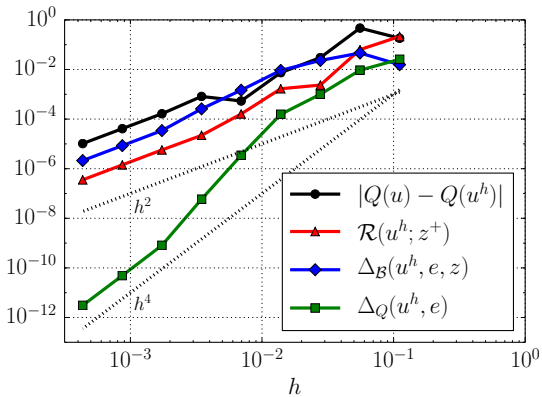
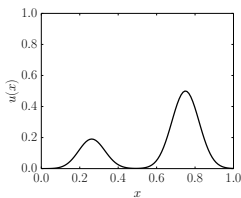
Nonlinear diffusion problem

Let $\Omega = (0, 1)$

$$-\nabla \cdot (A(u_x)u_x) = f$$

$$u(0) = u(1) = 0$$

$$A(u_x) = 1 + \frac{1}{10}u_x^2$$



Nonlinear diffusion - Convergence estimates

Observe that for small h , $\|(u_h + se)_x\|_{L^2} \sim \|(u_h)_x\|_{L^2}$,

$$\begin{aligned}
 |\Delta_B(u_h, e, z)| &= \left| \frac{3}{5} \int_0^1 \left(\int_{\Omega} (u_h + se)_x e_x^2 z_x dx \right) (1-s) ds \right| \\
 &\leq \frac{3}{5} \int_0^1 \|(u_h + se)_x\|_{L^2} \|e_x\|_{L^2}^2 \|z_x\|_{L^\infty} (1-s) ds \quad \sim \mathcal{O}(h^2)
 \end{aligned}$$

$$\begin{aligned}
 |\Delta_Q(u_h, e)| &= \left| \int_0^1 \left(2 \int_{\Omega} e^2 dx \right) (1-s) ds \right| \\
 &\leq \int_{\Omega} |e|^2 dx \\
 &\leq \|e\|_{L^2}^2 \quad \sim \mathcal{O}(h^4)
 \end{aligned}$$

GOEE: Nonlinear problems – Error estimators

$$Q(u) - Q(u_h) = \mathcal{R}(u_h; z) - \Delta_B(u_h; e, z) + \Delta_Q(u_h; e)$$

Classical estimator: Let $\tilde{z} \in \tilde{V} \supset V^h$

$$Q(u) - Q(u_h) \approx \mathcal{R}(u_h; \tilde{z} - \pi^h \tilde{z}) := \eta^{\mathcal{R}}$$

Nonlinear estimator Let $\tilde{e} = \tilde{u} - u_h$,

$$Q(u) - Q(u_h) \approx -\Delta_B(u_h; \tilde{e}, \tilde{z}) + \Delta_Q(u_h; \tilde{e}) := \eta^{\Delta}$$

Composite estimator

$$Q(u) - Q(u_h) \approx B'(u_h; \tilde{e}, \tilde{z}) + \Delta_Q(u_h; \tilde{e}) := \eta^{B'}$$

We also define corresponding element error indicators $\eta_{K}^{\mathcal{R}}$, η_{K}^{Δ} , and $\eta_{K}^{B'}$.

Incompressible flow past an obstacle (Re = 40)

Navier-Stokes equations:

$$-\nu \Delta \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} + \nabla p = 0$$

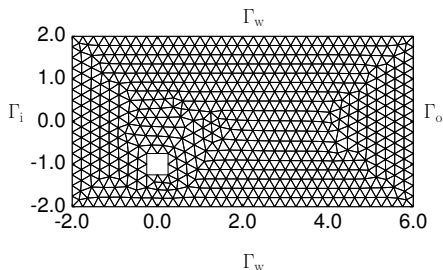
$$\nabla \cdot \mathbf{u} = 0$$

Boundary conditions:

$$\mathbf{u} = \mathbf{u}_{\text{in}}, \quad \mathbf{x} \in \Gamma_{\text{in}}$$

$$\mathbf{u} = 0, \quad \mathbf{x} \in \Gamma_{\text{w}} \cup \Gamma_{\text{sq}}$$

$$(\nu \nabla \mathbf{u} - p \mathbf{I}) \cdot \mathbf{n} = 0, \quad \mathbf{x} \in \Gamma_{\text{o}}$$



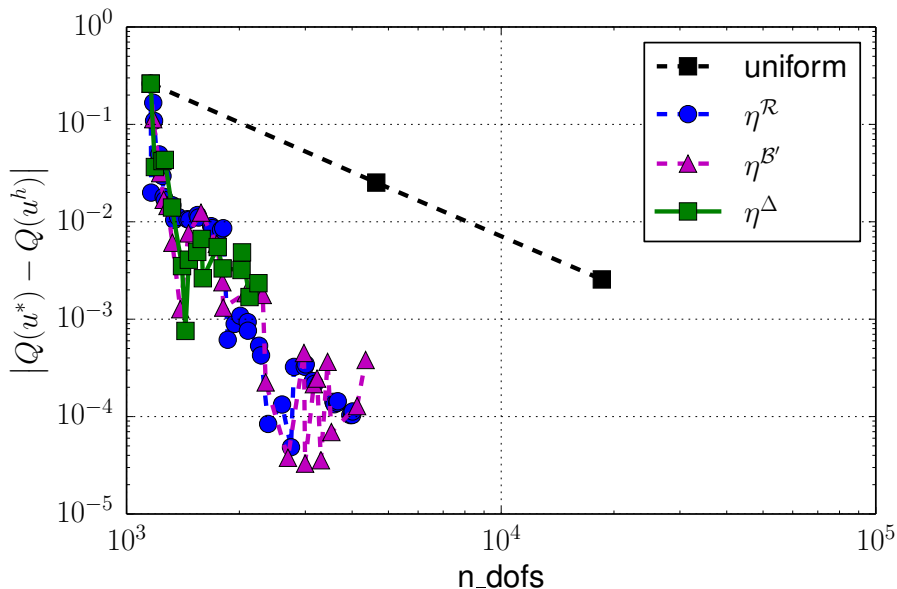
Quantity of interest:

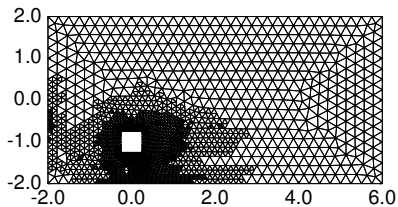
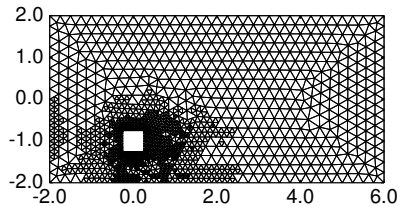
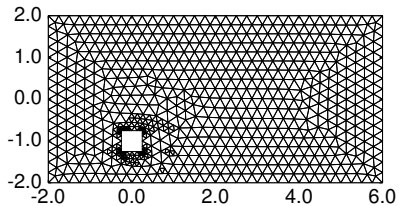
$$Q(\mathbf{u}) = u_x(\mathbf{x}_0), \quad \mathbf{x}_0 = (1, -1)^T$$

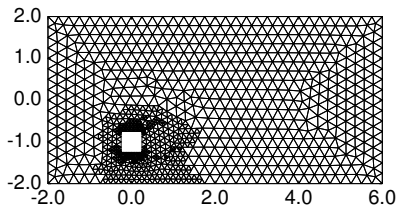
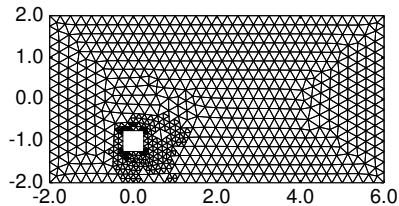
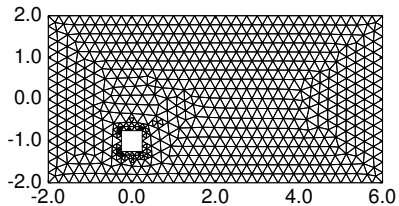
	$ \eta $	$\frac{ \eta }{ \mathcal{E}^* }$
$\eta^{\mathcal{R}}$	0.342	1.308
η^{Δ}	0.0496	0.189
$\eta^{B'}$	0.297	1.135

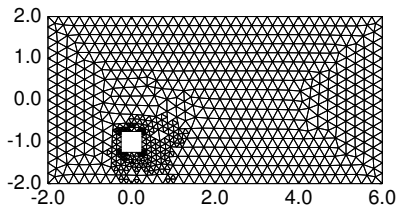
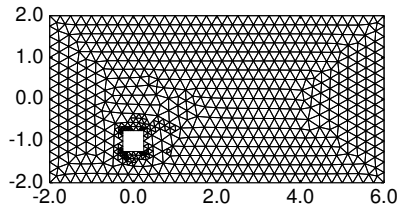
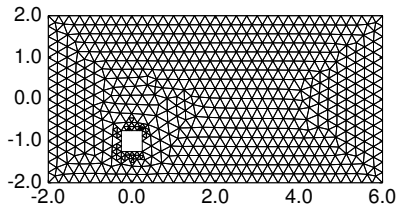
$$\mathcal{E}^* = Q(u^*) - Q(u_h)$$

Incompressible flow - convergence of error terms



Adaptivity based on contribution $\eta_K^{\mathcal{R}}$ 

Adaptivity based on contribution $\eta_K^{B'}$ 

Adaptivity based on contribution η_K^Δ 

Approaches for mesh adaptivity

- Ignore $\Delta = \Delta_Q - \Delta_B$. Compute \tilde{z} and $\tilde{z}_h = \Pi^{h,p}\tilde{z}$ such that:

$$\text{Find } \tilde{z} \in V^{h,p+1}, \quad B'(u_h; v, \tilde{z}) = Q'(u_h; v), \quad \forall v \in V^{h,p+1}$$

$$\Rightarrow \eta = \mathcal{R}(u_h; \tilde{z} - \tilde{z}_h)$$

- Estimate Δ :

- ▶ From linear case: $Q(u_h) = F(z_h)$. In nonlinear case, is $Q(u_h) - F(z_h)$ related to Δ ?
- ▶ Approximate error e , i.e. \hat{e} , using the linearized error equation:

$$\text{Find } \tilde{e} \in V^{h,p+1}, \quad B'(u_h; \tilde{e}, v) = \mathcal{R}(u_h; v), \quad \forall v \in V^{h,p+1}$$

Estimate $\Delta \approx \Delta_Q(u_h, \tilde{e}) - \Delta_B(u_h, \tilde{e}, \tilde{z})$. Relatively cheap but ignores linearization error in error equation.

Approaches for mesh adaptivity

- Approximate iteratively error \tilde{e}_i from \tilde{e}_{i-1}

$$\tilde{e}_i \in V^{h,p+1}, \quad B'(u_h + \tilde{e}_{i-1}; v) = \mathcal{R}(u_h + \tilde{e}_{i-1}; v), \quad v \in V^{h,p+1}$$

- Compute:

$$\Delta_i \approx \Delta_Q(u_h, \tilde{e}_i) - \Delta_B(u_h, \tilde{e}_i, \tilde{z})$$

or simply:

$$\Delta_i \approx Q(u_h + \tilde{e}_i) - Q(u_h) - \mathcal{R}(u_h; \tilde{z} - \tilde{z}_h)$$

- Stop iterative process if

$$|\mathcal{R}(u_h; \tilde{z} - \tilde{z}_h)| \gg |\Delta_i - \Delta_{i-1}|$$

Approaches for mesh adaptivity

- If $|\Delta| \ll |\mathcal{R}(u_h; \tilde{z} - \tilde{z}_h)|$, adapt mesh as usual based on the element-wise contributions to

$$\eta = \mathcal{R}(u_h; \tilde{z} - \tilde{z}_h) = \sum_K \mathcal{R}_K(u_h; \tilde{z} - \tilde{z}_h)$$

- Otherwise, consider

$$\eta = \mathcal{R}(u_h; \tilde{z} - \tilde{z}_h) + \Delta_Q(u_h, \tilde{e}) - \Delta_B(u_h, \tilde{e}, \tilde{z})$$

Note that Δ_Q and Δ_B are integrals defined over the whole domain that can be decomposed into element-wise contributions.

Question: should we refine linearization errors based on Δ_Q and Δ_B or simply on $\|\tilde{e}\|^2$?

Extension to Time-dependent Problems: Cahn-Hilliard

$$\left. \begin{aligned} u_t &= \Delta \mu \\ \mu &= f'(u) - \epsilon^2 \Delta u \end{aligned} \right\} \text{ in } \Omega + \text{BCs}$$

Weak formulation:

$$\text{Find } (u, \mu) \in W_0 \times V \text{ s.t. } \mathcal{B}((u, \mu); (v, \eta)) = 0 \quad \forall (v, \eta) \in V \times V$$

$$\mathcal{B}(U, V) = \int_0^T \left(\langle u_t, v \rangle + (\nabla \mu, \nabla v) \right) + \left((\mu, \eta) - (f'(u), \eta) - \epsilon^2 (\nabla u, \nabla \eta) \right) dt$$

$$W_0 = \{u \in V; u_t \in V^*, u(0) = u_0\}$$

$$V = L^2(0, T; H^1(\Omega))$$

Adjoint of the Cahn-Hilliard problem

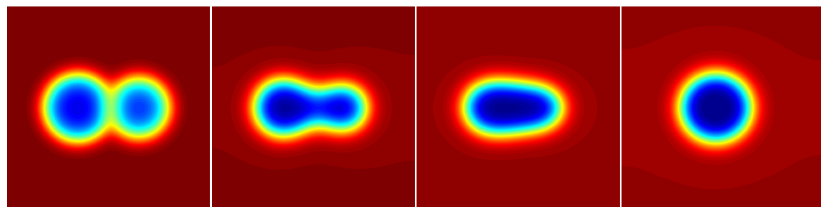
Goal:

$$Q(u, \mu) = (k_T, u(T)) + \int_0^T \left((k_u, u) + (k_\mu, \mu) \right) dt$$

Adjoint: (Backward-in-time linearized-adjoint problem)

$$\begin{aligned} -\partial_t p_u + \epsilon^2 \Delta p_\mu - f''(u_h) p_\mu &= k_u && \text{in } \Omega \\ p_\mu - \Delta p_u &= k_\mu && \text{in } \Omega \\ p_u(T) &= k_T && \text{(Final condition) + [Natural BCs]} \end{aligned}$$

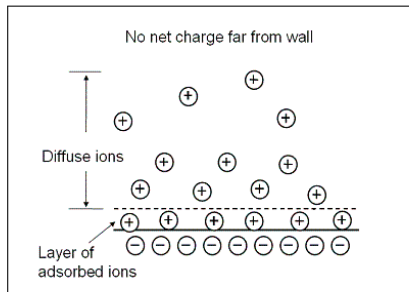
Example: merging bubbles



Time steps		$\mathcal{Q}(u, \mu) - \mathcal{Q}(\hat{u}, \hat{\mu})$			$\text{Est}(\hat{u}, \hat{\mu}; \hat{p}, \hat{\chi})$			Effectivity		
		64	256	1,024	64	256	1,024	64	256	1,024
Elements	64	-0.01183	-0.02205	-0.02306	-0.26392	-1.78933	-7.76063	22.311	81.162	336.469
	256	0.00946	0.01751	0.01982	0.02108	0.12233	0.38726	2.229	6.985	19.537
	1,024	0.00199	0.00104	0.00073	0.00322	0.00362	0.00315	1.620	3.484	4.308
	4,096	0.00165	0.00049	0.00012	0.00173	0.00077	0.00021	1.045	1.572	1.774
	16,384	0.00162	0.00048	0.00011	0.00144	0.00055	0.00013	0.886	1.146	1.159

van der Zee, Oden, Prudhomme, Hawkins, NMPDE 2011

Multiphysics Coupled Problems: Micro-fluidics



Structure of the Electric Double Layer (EDL) between the fluid and the wall.

Under the effect of an electric field tangent to the wall, charged particles are subjected to an electric force and thus move in the direction of the electric field.

Garg, Prudhomme, van der Zee, Carey, 2014.

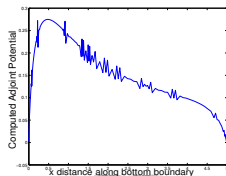
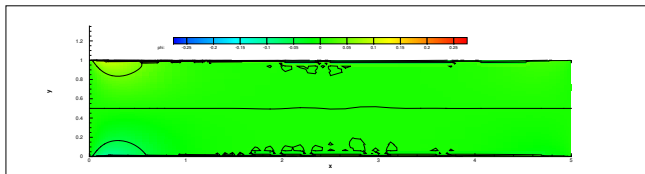
Coupled model (simplified):

$$\begin{aligned} -\nabla \cdot (\sigma_c \nabla \phi) &= 0 & \text{in } \Omega \\ \mu \Delta u + \nabla p &= 0 & \text{in } \Omega \\ \nabla \cdot u &= 0 & \text{in } \Omega \end{aligned}$$

Boundary conditions:

$$\begin{aligned} n \cdot (\sigma_c \nabla \phi) &= 0 & \text{on } \Gamma_w \\ \phi &= g & \text{on } \Gamma_{io} \\ u + \lambda \nabla \phi &= 0 & \text{on } \Gamma_w \\ u \cdot t &= 0 & \text{on } \Gamma_{io} \\ n \cdot (\sigma \cdot n) &= 0 & \text{on } \Gamma_{io} \end{aligned}$$

Micro-Fluidic Flows



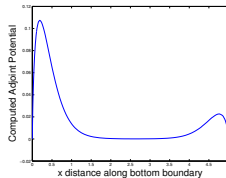
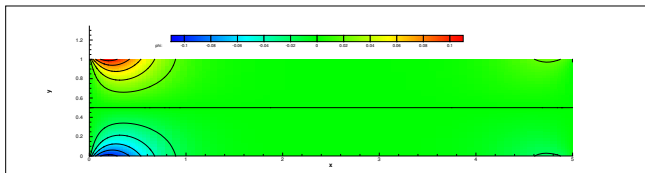
$$u + \lambda \nabla \phi = 0$$

 \Rightarrow

$$\begin{aligned} u \cdot t + \lambda \partial_t \phi &= 0 \\ u \cdot n + \lambda \partial_n \phi &= 0 \end{aligned}$$

 \Rightarrow

$$\begin{aligned} u \cdot t + \lambda \partial_t \phi &= 0 \\ u \cdot n &= 0 \end{aligned}$$



Micro-Fluidic Flows

Strong form of adjoint problem:

$$\begin{array}{ll}
 -\nabla \cdot (\sigma_c \nabla \varphi^*) = 0 & \text{in } \Omega \\
 -\Delta w^* + \nabla p^* = k\alpha & \text{in } \Omega \\
 -\nabla \cdot w^* = 0 & \text{in } \Omega
 \end{array}$$

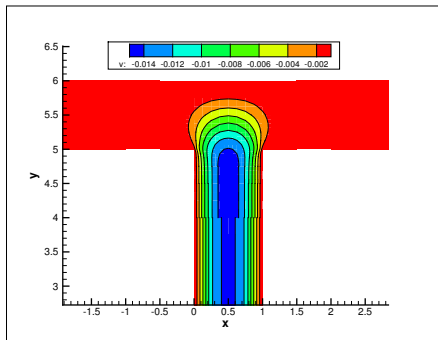
with three boundary conditions on Γ_{io} :

$$\begin{array}{ll}
 \varphi^* = 0 & \text{on } \Gamma_{io} \\
 w^* \cdot t = 0 & \text{on } \Gamma_{io} \\
 n \cdot (\sigma^* \cdot n) = k & \text{on } \Gamma_{io}
 \end{array}$$

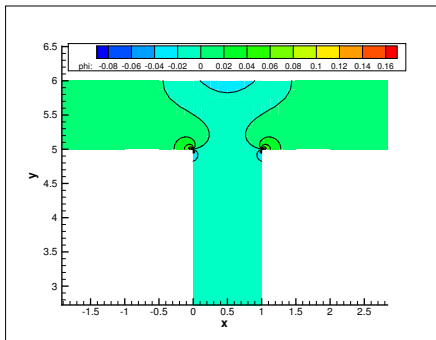
and three BCs on Γ_w :

$$\begin{array}{ll}
 n \cdot (\sigma_c \nabla \varphi^*) + \nabla_{\Gamma_w} \cdot ((\lambda t \cdot (\sigma^* \cdot n))t) = 0 & \text{on } \Gamma_w \\
 w^* = 0 & \text{on } \Gamma_w
 \end{array}$$

Micro-Fluidic Flows

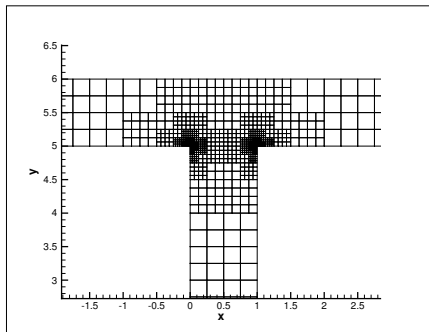


Adjoint velocity u^* (in y direction). It is mainly different from zero inside the vertical channel indicating that the primal solution needs to be accurate in that region.

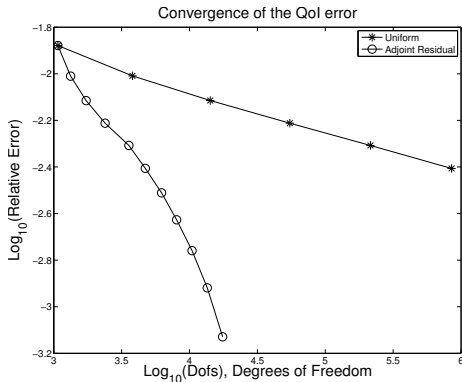


Adjoint potential ϕ^* . Note that ϕ^* almost vanishes everywhere except at the corners and along the middle section of the top wall.

Micro-Fluidic Flows



Adaptive mesh obtained using adjoint-based error estimates. Note that the elements get refined almost exclusively near the corners due to the singularities in the primal velocity and adjoint potential.



Convergence plots for the relative error in QoI using uniform and adjoint based refinements.

Conclusions

- Goal-oriented error estimation is based on the notion of the adjoint/dual problem.
- The adjoint problem is “straightforwardly” derived from the weak formulation of the primal problem.
- The error in the quantity of interest is represented as the product of the residual by the adjoint solution.
- Error estimates and refinement indicators are obtained by solving for approximations to the adjoint problem.
- Approach is generic!
- Goal-oriented error estimates can be applied to other discretization methods (FD, FV, DGM, spectral methods, meshless methods, etc.).