

A contour plot showing a flow field around a curved surface, likely an airfoil. The plot uses a color scale from blue (low) to red (high) to represent a scalar quantity. Concentric contour lines are visible, indicating a source or sink. A small inset in the top right corner shows a coordinate system with x and z axes.

# MANE 6960:

## Adjoint for Scientists and Engineers

Lecture 13

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JEC 2036

# Motivation

Thus far, I have focused on using the adjoint for sensitivity analysis, that is, computing total derivatives like  $DJ_h/D\alpha$

Today, we begin to investigate a different application of the adjoint: output-based error estimation.

## Motivation (cont.)

We often solve (numerically) PDEs in order to predict quantities of interest that depend on the state, that is, functionals.

- output, quantity of interest, and functional are synonymous in this context

Examples:

- lift, drag, and moment on an aerodynamic body
- maximum stress in a structure
- average power produced by an I.C. engine

## Motivation (cont.)

Given the central role of outputs/functionals, it is important to be able to estimate the numerical error

$$\delta J_h \equiv J_h(u_h) - J(u),$$

since such an error estimate can tell us if the prediction can be trusted, or if the mesh needs to be refined.

How can we estimate  $\delta J_h$  without knowing the exact solution?

# Adjoint-Weighted-Residual Method for Linear Problems

# Coarse and Fine Spaces

We cannot, in general, use the exact solution in order to estimate the functional error. However, we can approximate the exact solution using a computational mesh that is more resolved than the baseline mesh.

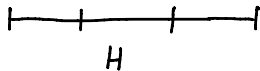
Therefore, suppose we have two meshes: a coarse mesh and a fine mesh.

- More generally, we just need a coarse- and fine-solution spaces, which might exist on the same mesh.
- We will use  $H$  to denote the coarse mesh/space and  $h$  to denote the fine mesh/space.

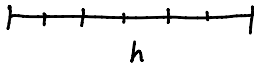
# Coarse and Fine Spaces (cont.)

Examples of coarse and fine spaces:

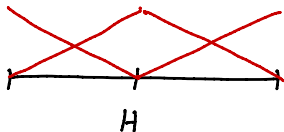
FV/FD  
or FE



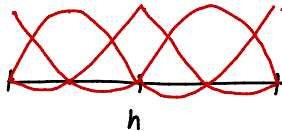
$h$  enrichment



FE

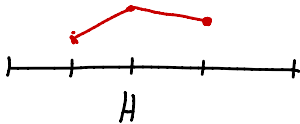


$p$  enrichment

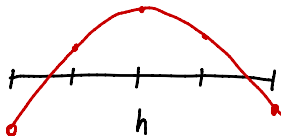


FD

or  
spline  
FE



High order  
interpolation



# Adjoint-Weighted-Residual Method

Now, suppose we have solved the discretized BVP of interest on mesh/space  $H$ :

$$L_H u_H = f_H.$$

We want to estimate the error in the (linear) functional

$$J_H(u_H) = (g_H, u_H)_H.$$

That is, we want to estimate the quantity

$$\delta J_H = J_H(u_H) - J(u).$$



# Adjoint-Weighted-Residual Method (cont.)

We begin by replacing the exact functional with an approximation based on the fine mesh/space:

$$\begin{aligned}\delta J_H &\approx J_H(u_H) - J_h(u_h) \\ &= (g_H, u_H)_H - (g_h, u_h)_h\end{aligned}$$



To proceed, I would like to group these products as

$$(g?, u_H - u_h)?$$

but  $g$  and  $(,)$  are different, in general

# Adjoint-Weighted-Residual Method (cont.)

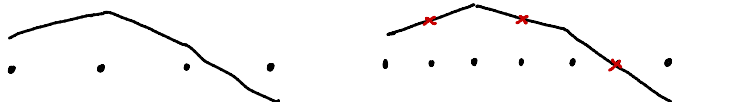
In order to relate the two (discrete) functional values, we need to work in the same space.

- Moving entirely to the coarse mesh/space will, in general, result in a loss of information.
- We need to represent the coarse solution,  $u_H$ , on the fine mesh/space.

Therefore, we introduce a prolongation operator  $I_h^H$  from the coarse to the fine mesh/space, and we define

$$u_h^H \equiv I_h^H u_H.$$

Example



# Adjoint-Weighted-Residual Method (cont.)

Then we have

$$\begin{aligned}
 (g_H, u_H)_H - (g_h, u_h)_h &= (g_h, u_h^H - u_h)_h \\
 &= (L_h^T \psi_h, u_h^H - u_h)_h, \quad \because L_h^T \psi_h = g_h \\
 &= (\psi_h, L_h(u_h^H - u_h))_h - \underbrace{(\psi_h, f_h - f_h)_h}_{\text{zero}} \\
 &= (\psi_h, L_h u_h^H - f_h)_h - (\psi_h, \cancel{L_h u_h} - f_h)_h \\
 &\qquad \qquad \qquad \searrow \text{0} \\
 &\qquad \qquad \qquad u_h \text{ solves } L_h u_h = f_h
 \end{aligned}$$

# Adjoint-Weighted-Residual Method (cont.)

## Theorem: Adjoint-Weighted Residual (AWR)

Let  $u_H$  denote the solution to the discretized BVP  $L_H u_H = f_H$  on a coarse space. Analogously, let  $u_h$  denote the solution of  $L_h u_h = f_h$ , the fine-space discretization of the same BVP. Then the difference

$$J_H(u_H) - J_h(u_h) = (g_H, u_H)_H - (g_h, u_h)_h = (\psi_h, L_h u_h^H - f_h)_h,$$

where  $\psi_h$  is the solution to the fine-space adjoint equation

$$L_h^T \psi_h = g_h.$$

# Remarks on the AWR

*hence the name AWR*

This result says that the output/functional error can be estimated by weighting the residual  $L_h u_h^H - f_h$  by the fine-space adjoint  $\psi_h$ .

While interesting, this result is not yet practical:

- solving for  $\psi_h$  is as expensive as solving for  $u_h$ , at least in the linear BVP case; and
- if we had  $u_h$ , we could just as easily compute  $J_H(u_H) - J_h(u_h)$  directly.

# Remarks on the AWR (cont.)

In the case of finite-difference and finite-volume discretizations, the AWR can be made practical by approximating the fine-space adjoint using the prolongation operator:

$$\psi_h \approx I_h^H \psi_H = \psi_h^H$$

Thus we have the estimate

$$J_H(u_H) - J_h(u_h) \approx (\psi_h^H, L_h u_h^H - f_h)_h$$

- This approach requires us to solve for the adjoint on the coarse mesh/space only.

## Remarks on the AWR (cont.)

Unfortunately, this technique does not work for Galerkin finite-element methods since

$$\begin{aligned} (\psi_h^H, L_h u_h^H - f_h)_h &= (\Psi_H, L_H u_H - f_H)_H \\ &= 0 \end{aligned}$$

(assumes no "variational crimes")

Follows from Galerkin orthogonality,  
since residual is perpendicular to  
all functions in coarse space,  
including  $\Psi_H$ .

## Example 1: Linear Advection [Hic12]

As a simple example, consider 2D linear advection as the BVP:

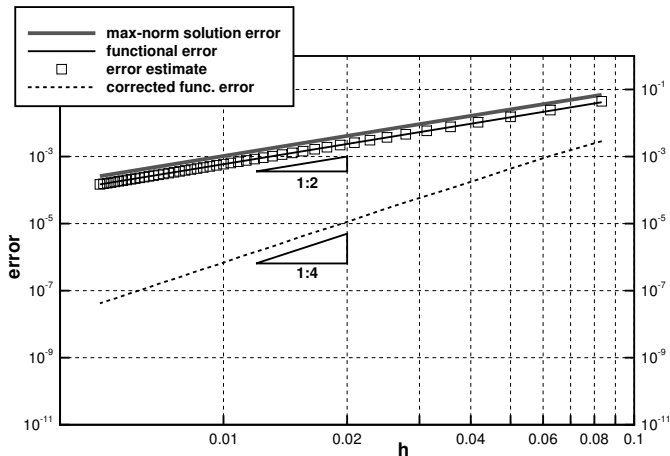
$$\begin{aligned}\nabla \cdot (\lambda u) &= f, & \forall x \in [0, 1]^2 \\ u(x) &= b(x) & \forall x \in \Gamma_-.\end{aligned}$$

- functional is an integral over the outlet
- SBP-SAT finite-difference discretization
- fine “space” is based on using a higher-order FD operator on the same mesh
- thus,  $I_h^H$  is the identity matrix here



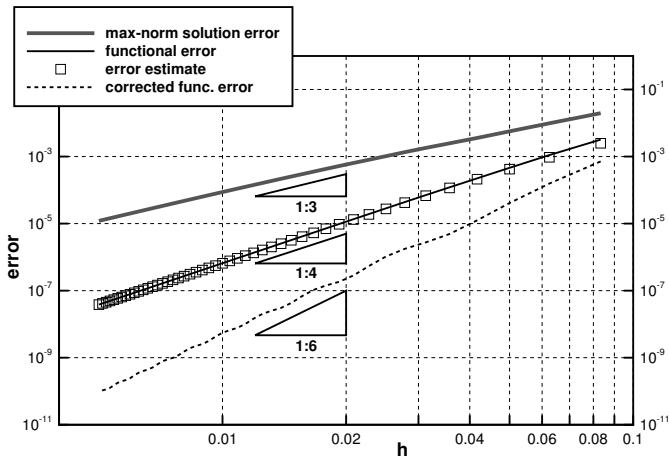
# Example 1: Linear Advection [Hic12] (cont.)

2nd-order coarse space; 3rd-order fine space



# Example 1: Linear Advection [Hic12] (cont.)

3rd-order coarse space; 4th-order fine space



## Example 2: Poisson BVP [Hic12]

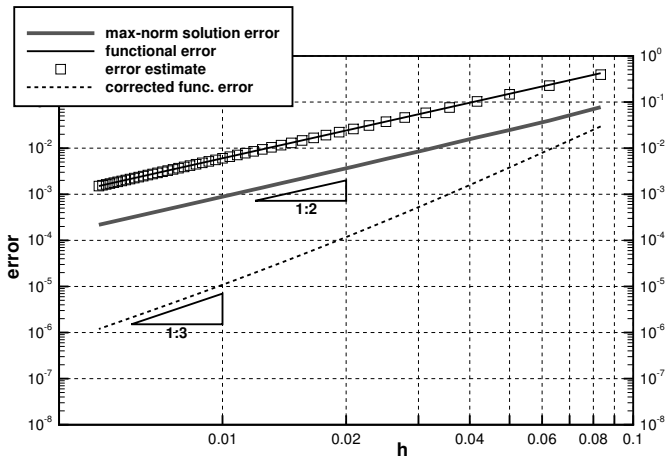
Next, consider the 2D Poisson BVP

$$\begin{aligned} -\nabla \cdot (\gamma \nabla u) &= f, & \forall x \in [0, 1]^2 \\ u(x) &= b(x) & \forall x \in \Gamma. \end{aligned}$$

- functional is an integral over the entire boundary
- SBP-SAT finite-difference discretization
- fine “space” is based on using a higher-order FD operator on the same mesh
- thus,  $I_h^H$  is the identity matrix here

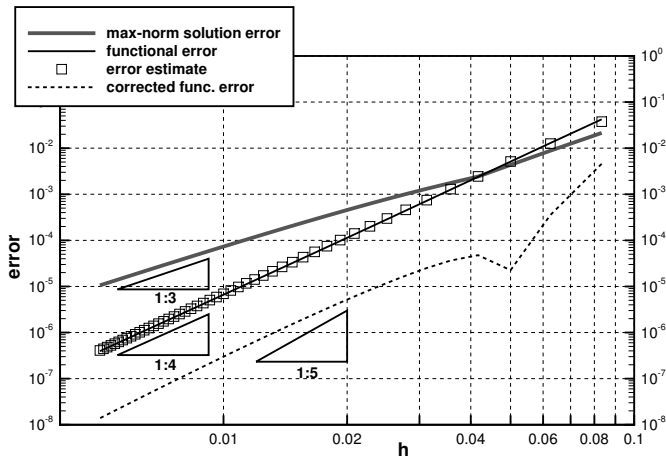
# Example 2: Poisson BVP [Hic12] (cont.)

2nd-order coarse space; 3rd-order fine space



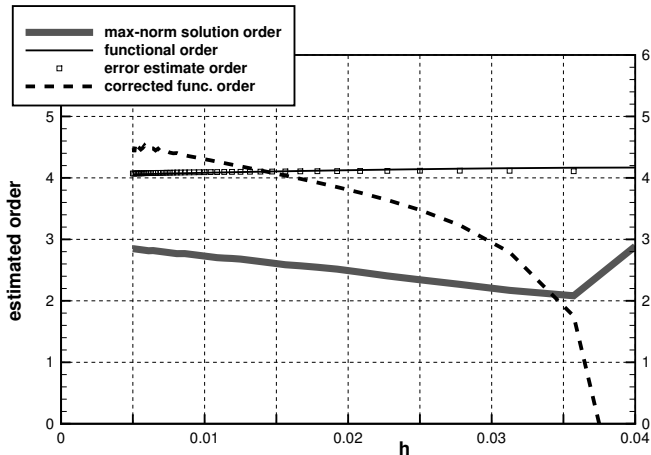
# Example 2: Poisson BVP [Hic12] (cont.)

3rd-order coarse space; 4th-order fine space



## Example 2: Poisson BVP [Hic12] (cont.)

Asymptotically, the error estimate appears to approach 5th order



# References

- [Hic12] Jason E. Hicken, *Output error estimation for summation-by-parts finite-difference schemes*, Journal of Computational Physics **231** (2012), no. 9, 3828–3848.