

Motivation

Last class I introduced the adjoint-weighted residual (AWR) method in the context of linear BVPs and linear functionals.

In this lecture, I will first generalize the AWR method to nonlinear problems. Subsequently, I will describe various approximations that are used to make the method useful in practice.

AWR Method for Nonlinear Problems

Hicken (RPI) Adjoints Spring 2018 3 / 30

Problem Statement

Let $R_H: \mathbb{R}^S \to \mathbb{R}^S$ be the residual corresponding to the discretization of a nonlinear BVP on a mesh of nominal size H. Furthermore, let $u_H \in \mathbb{R}^S$ be the solution to

$$R_H(u_H) = 0.$$

Our goal is to estimate the functional error

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

where J(u) is the exact (nonlinear) functional value based on the exact solution of the nonlinear BVP, and $J_H(u_H)$ is the discretized functional value based on the discrete solution.

 Hicken (RPI)
 Adjoints
 Spring 2018
 4 / 30

Averaged derivatives

Proceeding as in the linear case, we have

$$\delta J_H \approx J_H(u_H) - J_h(u_h),$$

where $u_h \in \mathbb{R}^s$ is the solution to $R_h(u_h) = 0$, the discretized BVP on a fine mesh with nominal element size h.

For the subsequent analysis, we define

$$\delta u_h \equiv u_h^H - u_h,$$

where, as before, $u_h^H \equiv I_h^H u_H$ is the coarse solution represented on the fine mesh/space.

Hicken (RPI) Adjoints Spring 2018 5 / 30

Averaged derivatives (cont.)

Thus, we have

$$J_H(u_H) - J_h(u_h) = J_h(u_h^H) - J_h(u_h)$$

= $J_h(u_h + \delta u_h) - J_h(u_h)$.

Unlike the linear case, we cannot express this functional difference as a functional of the difference. However, notice that

$$J_{h}(u_{h} + \delta u_{h}) - J(u_{h}) = \int_{0}^{1} \frac{d}{ds} J_{h}(u_{h} + s \delta u_{h}) ds$$

$$= \int_{0}^{1} J_{h}' [u_{h} + s \delta u_{h}]^{T} \delta u_{h} ds$$

$$= \left(\int_{0}^{1} J_{h}' [u_{h} + s \delta u_{h}] ds \right)^{T} \delta u_{h}$$

 Hicken (RPI)
 Adjoints
 Spring 2018
 6 / 30

Averaged derivatives (cont.)

Consequently, the functional difference is equal to an averaged derivative times δu_h .

$$(g_h,\delta u_h)_h=g_h^T\delta u_h \qquad o \qquad \left(\int_0^1 J_h'[u_h+s\delta u_h]\,ds
ight)^T\delta u_h.$$
 Linear case

This suggests that we define the adjoint in a slightly different way from the linear case. . .

 Hicken (RPI)
 Adjoints
 Spring 2018
 7 / 30

Definition: Mean-value Adjoint [FD11]

The (discrete) mean-value adjoint is denoted by $\overline{\psi}_h$ and satisfies the linear equation

$$\left(\overline{R}_h[u_h, u_h^H]\right)^T \overline{\psi}_h = \left(\overline{J}_h[u_h, u_h^H]\right)^T,$$

where the averaged Jacobian and gradient are, respectively,

$$\overline{R}_h[u_h, u_h^H] \equiv \int_0^1 R_h'[u_h + s\delta u_h] ds$$

$$\overline{J}_h[u_h, u_h^H] \equiv \int_0^1 J_h'[u_h + s\delta u_h] ds.$$

Hicken (RPI) Adjoints Spring 2018 8 / 30

Mean-value Adjoint

Notice that we have the following identities:

$$\overline{R}_{h}[u_{h}, u_{h}^{H}] \delta u_{h} = \int_{0}^{1} R'_{h}[u_{h} + s \delta u_{h}] \delta u_{h} ds$$

$$= \int_{0}^{1} \frac{d}{ds} R'_{h}(u_{h} + s \delta u_{h}) ds$$

$$= R'_{h}(u_{h} + \delta u_{h}) - R'_{h}(u_{h}) \qquad (1)$$

Similarly

$$\overline{J}_h[u_h, u_h^H] \delta u_h = J_h \left(\mathbf{u_h} + \delta \mathbf{u_h} \right) - J_h \left(\mathbf{u_h} \right) \tag{2}$$

 Hicken (RPI)
 Adjoints
 Spring 2018
 9 / 30

Mean-value Adjoint (cont.)

Using the above identities, as well as the definition of the mean-value adjoint, we find that

$$J_{h}(u_{h} + \delta u_{h}) - J_{h}(u_{h}) = \overline{J_{h}} [u_{h}, u_{h}^{H}] \delta u_{h} , \text{ by (2)}$$

$$= \overline{\Psi_{h}}^{T} \overline{R}_{h} [u_{h}, u_{h}^{H}] \delta u_{h} , \text{ by def}^{n} \overline{\Psi_{h}}$$

$$= \overline{\Psi_{h}}^{T} (R_{h} (u_{h} + \delta u_{h}) - R_{h} (u_{h}))^{O} , \text{ by (1)}$$

$$= \overline{\Psi_{h}}^{T} R_{h} (u_{h}^{H})$$

 Hicken (RPI)
 Adjoints
 Spring 2018
 10 / 30

Mean-value Adjoint (cont.)

Theorem: AWR for nonlinear problems

Let u_H denote the solution to $R_H(u_H)=0$, which corresponds to a discretized BVP on a coarse space. Analogously, let u_h denote the solution of $R_h(u_h)=0$, the fine-space discretization of the same BVP. Then the difference

$$J_H(u_H) - J_h(u_h) = \overline{\psi}_h^T R_h(u_h^H),$$

where $\overline{\psi}_h$ is the solution to the mean-value adjoint equation on the fine-space.

 Hicken (RPI)
 Adjoints
 Spring 2018
 11 / 30

AWR Approximations

Problems with the AWR Method

There is a significant computational cost when applying the AWR method to nonlinear problems: the mean-value adjoint.

 The mean-value adjoint requires the integral of potentially expensive derivatives.

The AWR method, whether it is used on linear or nonlinear problems, has another drawback.

• The adjoint, ψ_h or $\overline{\psi}_h$, is needed on the fine mesh/space.

Let's consider the approximations used to mitigate these costs.

 Hicken (RPI)
 Adjoints
 Spring 2018
 13 / 30

Avoiding the Mean-value Adjoint

We can avoid the mean-value adjoint by replacing the integrated Jacobian and gradient with point derivatives as follows:

$$\begin{split} J_h(u_h + \delta u_h) - J_h(u_h) &= \int_{\bullet}^{I} J_h'[u_h + s \, \delta u_h] \, \delta u_h \, ds \\ &= J_h(u_h) + J_h'[u_h] \, \delta u_h + O(\|\delta u_h\|^2) - J_h(u_h) \\ &= J_h'[u_h] \, \delta u_h + O(\|\delta u_h\|^2) \end{split}$$

 Hicken (RPI)
 Adjoints
 Spring 2018
 14 / 30

Avoiding the Mean-value Adjoint (cont.)

Based on the above approximation, we can use the adjoint corresponding to the linearization about u_h :

$$(R_h'[u_h])^T\psi_h=J_h'[u_h].$$
no averaged Leurvatives

Thus,

$$J_{h}(u_{h} + \delta u_{h}) - J_{h}(u_{h}) = J_{h}'[u_{h}] \delta u_{h} + O(\|\delta u_{h}\|^{2})$$

$$= \Psi_{h}^{T} R_{h}'[u_{h}] \delta u_{h} + O(\|\delta u_{h}\|^{2})$$

$$= \Psi_{h}^{T} (R_{h}(u_{h} + \delta u_{h}) - R_{h}(u_{h}))^{T} + O(\|\delta u_{h}\|^{2})$$

$$= \Psi_{h}^{T} R_{h}(u_{h}^{H}) + O(\|\delta u_{h}\|^{2})$$

 Hicken (RPI)
 Adjoints
 Spring 2018
 15 / 30

Avoiding the Mean-value Adjoint (cont.)

The above approximation for the mean-value adjoint is frequently used in the practice.

- The approximation is justified provided δu_h is sufficiently small.
- If the problem is highly nonlinear (e.g. shocks), then δu_h may become large.

Hicken (RPI) Adjoints Spring 2018 16 / 30

Avoiding Fine-space Solutions

The second problem with the exact AWR is the need for ψ_h , which implies the solution of

$$(R_h'[u_h])^T \psi_h = J_h'[u_h].$$

This is problematic for two reasons:

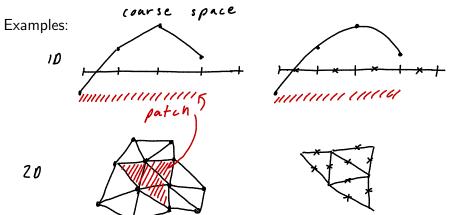
- The linear system for ψ_h may be significantly larger than the baseline problem for u_H .
- The Jacobian and gradient are linearized about u_h , which we do not have.

nonlinear case on ly

Hicken (RPI) Adjoints Spring 2018 17 / 30

Avoiding Fine-space Solutions (cont.)

One common approach to approximate ψ_h (and u_h), is to use a patch-based reconstruction, in which ψ_H is interpolated onto the fine space using a larger stencil.



 Hicken (RPI)
 Adjoints
 Spring 2018
 18 / 30

Avoiding Fine-space Solutions (cont.)

A disadvantage of the reconstruction approach, especially in the case of u_h , is that it does not account for the physics of the problem.

 A high-order interpolation is not justified near a discontinuity, such as a shock.

Hicken (RPI) Adjoints Spring 2018 19 / 30

Avoiding Fine-space Solutions (cont.)

Consequently, a second common approach is to approximately solve for ψ_h using a few iterations of an iterative method.

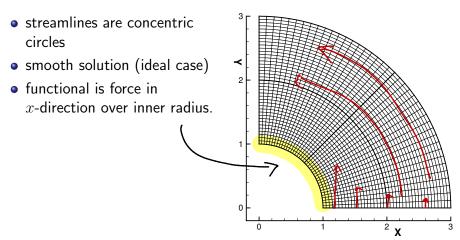
• This is justified, since the difference $\psi_h^H - \psi_h$ is likely to contain mostly high-frequency terms that can be eliminated rapidly using a suitable stationary iterative method (i.e. a smoother).

 Hicken (RPI)
 Adjoints
 Spring 2018
 20 / 30

AWR Examples: Euler Equations

Vortex Flow [Hic12]

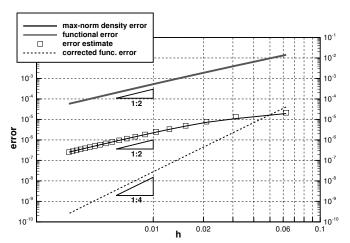
Isentropic Vortex flow:



 Hicken (RPI)
 Adjoints
 Spring 2018
 22 / 30

Vortex Flow [Hic12] (cont.)

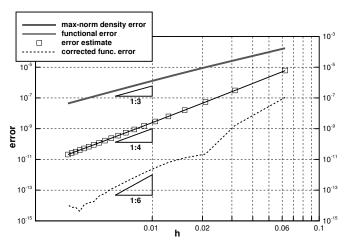
2nd-order coarse space; 3rd-order fine space



Hicken (RPI) Adjoints Spring 2018 23 / 30

Vortex Flow [Hic12] (cont.)

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

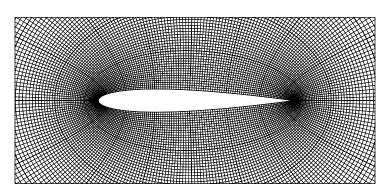


Hicken (RPI) Adjoints Spring 2018 24 / 30

Subsonic Airfoil [Hic12]

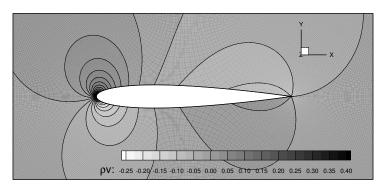
Mach 0.5 flow over NACA 0012

- functional is coefficient of drag
- flow and adjoint fields have singularities at trailing edge



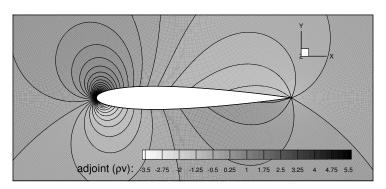
 Hicken (RPI)
 Adjoints
 Spring 2018
 25 / 30

ρv contours



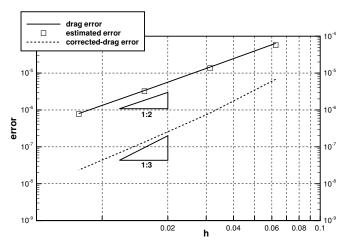
Hicken (RPI) Adjoints Spring 2018 26 / 30

$\psi_{ ho v}$ contours



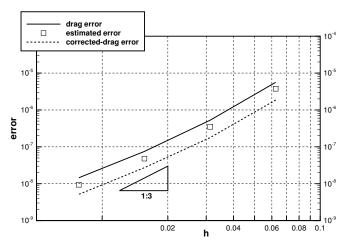
 Hicken (RPI)
 Adjoints
 Spring 2018
 27 / 30

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



Hicken (RPI) Adjoints Spring 2018 28 / 30

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



Hicken (RPI) Adjoints Spring 2018 29 / 30

References

- [FD11] Krzysztof J. Fidkowski and David L. Darmofal, Review of output-based error estimation and mesh adaptation in computational fluid dynamics, AIAA Journal 49 (2011), no. 4, 673–694.
- [Hic12] Jason E. Hicken, Output error estimation for summation-by-parts finite-difference schemes, Journal of Computational Physics **231** (2012), no. 9, 3828–3848.

Hicken (RPI) Adjoints Spring 2018 30 / 30