

Artificial Viscosity Proper Orthogonal Decomposition

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Overview

- Proper Orthogonal Decomposition (POD)
- POD Model
- Artificial Viscosity-POD (AV-POD) Model
- Theoretical results
- Numerical results
- Conclusions

Proper Orthogonal Decomposition

Navier-Stokes equations

$$\partial_t \mathbf{U} + (\mathbf{U} \cdot \nabla) \mathbf{U} = -\nabla P + \text{Re}^{-1} \Delta \mathbf{U}$$

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

- $u^{(1)}, u^{(2)}, \dots, u^{(m)}$ snapshots with rank d
- Find $\{\psi_1, \dots, \psi_r\}$, $r \leq d$, orthonormal basis

$$\min_{\{\psi_k\}_{k=1}^r} \frac{1}{m} \sum_{j=1}^m \left\| u^{(j)}(\cdot, t) - \sum_{k=1}^r (u^{(j)}(\cdot, t), \psi_k(\cdot)) \psi_k(\cdot) \right\|_{L_2}^2$$

- $K \vec{v}_k = \lambda_k \vec{v}_k, \quad K_{ij} = \frac{1}{m} (u^{(j)}, u^{(i)})$
- $\psi_\ell = \frac{1}{\sqrt{\lambda_\ell}} \sum_{j=1}^m (\vec{v}_\ell)_j \vec{u}_j, \ell = 1, \dots, r$ [Kunisch & Volkwein, 2001]

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POD model

- POD truncation

$$\mathbf{U} \approx \mathbf{u}^r = \sum_{i=1}^r \mathbf{a}_i(t) \psi_i(\mathbf{x})$$

- **POD-Galerkin model**

$$\left(\frac{\partial \mathbf{u}^r}{\partial t}, \psi_{\mathbf{k}} \right) + \left((\mathbf{u}^r \cdot \nabla) \mathbf{u}^r, \psi_{\mathbf{k}} \right) + \left(\frac{2}{Re} \mathbb{D}(\mathbf{u}^r), \nabla \psi_{\mathbf{k}} \right) = 0$$

- high Reynolds number $\Rightarrow \{\psi_{r+1}, \dots, \psi_d\}$ important

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Artificial Viscosity POD model

- Energy cascade (Richardson , Kolmogorov)
- Artificial Viscosity \Rightarrow Dissipation
- AV-POD model (Smagorinsky)

$$\left(\frac{\partial \mathbf{u}^r}{\partial t}, \psi_{\mathbf{k}} \right) + ((\mathbf{u}^r \cdot \nabla) \mathbf{u}^r, \psi_{\mathbf{k}}) + \left(\left(\nu_S + \frac{2}{Re} \right) \mathbb{D}(\mathbf{u}^r), \nabla \psi_{\mathbf{k}} \right) = 0$$

$$\nu_S := C_S |\mathbb{D}(\mathbf{u}^r)|$$

ν_S models $\{\psi_{r+1}, \dots, \psi_d\}$

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Theoretical results for Burgers Equation

Burgers equation

$$\begin{cases} u_t - \nu u_{xx} + u u_x = f & \text{in } \Omega \\ u(x, 0) = u_0(x) & \text{in } \Omega \\ u(x, t) = g(x, t) & \text{on } \partial\Omega \end{cases}$$

Error Estimates

$$\frac{1}{m} \sum_{k=1}^m \|U_k - u(t_k)\|^2 \leq C \left(\|u_0 - p^\ell u_0\|^2 + \sum_{k=r+1}^d \lambda_k + \Delta t^2 \right)$$

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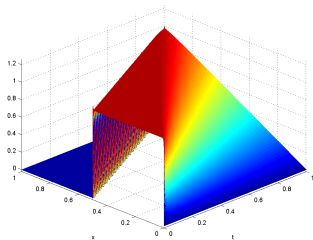
Numerical results for Burgers Equation

Experiment 1 [Kunisch & Volkwein, 1999]

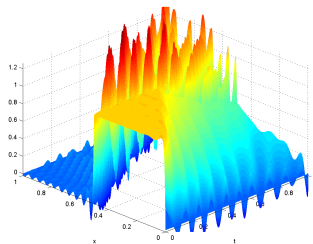
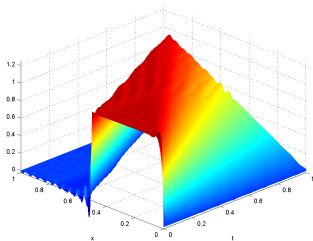
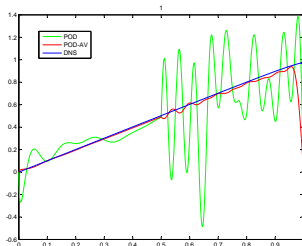
$$u_0(x) = \begin{cases} 1 & \text{if } x \in (0, \frac{1}{2}] \\ 0 & \text{if } x \in (\frac{1}{2}, 1) . \end{cases}$$

- $u(0, t) = u(1, t) = 0$
- $f = 0$
- $[0, T] = [0, 1]$
- $\Delta x = 1/2048;$
- $\Delta t = 10^{-3};$
- $\nu = 10^{-5};$
- $m = 1000$ (snapshots).

DNS



POD model

AV-POD model ($c = 1 \times 10^{-4}$)POD, AV-POD, DNS at $T = 1$ 

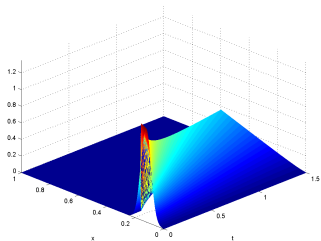
Numerical results for Burgers Equation

Experiment 2 [Mohseni, Zhao & Marsden, 2006]

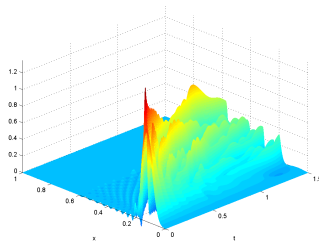
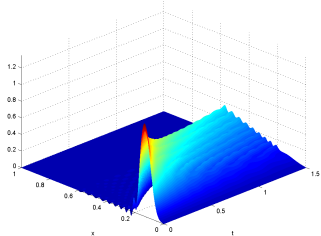
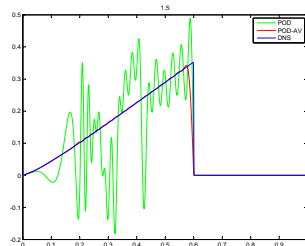
$$u_0(x) = \frac{2}{\sqrt{\pi}} \exp(-400(x - 1/8)^2);$$

- $u(0, t) = u(1, t) = 0$
- $f = 0$
- $[0, T] = [0, 1.5]$
- $\Delta x = 1/2048;$
- $\Delta t = 10^{-3};$
- $\nu = 3.75 \times 10^{-5};$
- $m = 1000$ (snapshots).

DNS



POD model

AV-POD model ($c = 5 \times 10^{-5}$)POD, AV-POD, DNS at $T = 1.5$ 

Conclusions

Accomplishments

- ◇ introduced AV-POD models
- ◇ theoretical error estimates (Burgers Equation)
- ◇ numerical experiments (Burgers Equation)
- ◇ *“Artificial Viscosity Proper Orthogonal Decomposition”*

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Future work

- ◇ 2D, 3D Navier-Stokes Equations
- ◇ turbulent pipe flow
- ◇ optimal model parameters
- ◇ improved models

Thank You!

Reference

- J. Borggaard, A. Duggleby, A. Hay, T. Iliescu and Z. Wang, Reduced-order modeling of turbulent flows, Proceeding of MTNS 2008.
- J. Borggaard, T. Iliescu and Z. Wang, Artificial viscosity proper orthogonal decomposition models, to be submitted.
- M. Couplet, P. Sagaut and C. Basdevant, Intermodal energy transfers in a proper orthogonal decomposition-Galerkin representation of a turbulent separated flow. *J. Fluid Mech.*, 491: 275-284, 2003.
- P. Holmes, J. L. Lumley and G. Berkooz, Turbulence, Coherent Structures, Dynamical Systems and Symmetry. Cambridge, 1996.
- K. Kunisch and S. Volkwein, Control of the Burgers equation by a reduced-order approach using Proper Orthogonal Decomposition, *Journal of optimization theory and application* 1999; **Vol.102, No.2**: 345-371.

Reference

- K. Kunisch and S. Volkwein,
Galerkin proper orthogonal decomposition methods for parabolic problems. *Numer. Math.*, 90(1): 117-148, 2001.
- K. Mohseni, H. Zhao, and Marsden
Shock Regularization for the Burgers equation, *44th AIAA aerospace sciences meeting and exhibit*: 2006-1516
- B. Podvin,
On the adequacy of the ten-dimensional model for the wall layer. *Phys. Fluids*, 13(1): 210-224, 2001.
- B. Podvin and J. Lumley,
A low-dimensional approach for the minimal flow unit. *J. Fluid Mech.*, 362: 121-155, 1998.
- P. Sagaut,
Large eddy simulation for incompressible flows. Scientific Computation. Springer-Verlag, Berlin, third edition, 2006.