

# Reduced Order Models on a Variational Multi-Scale Model of Navier–Stokes: foucs on wall law functions for boundary layer treatment preliminary study



**Davide Torlo, Giovanni Stabile,  
Samuele Rubino, Gianluigi Rozza**

MathLab, Mathematics Area, SISSA International  
School for Advanced Studies, Trieste, Italy  
[davidetorlo.it](http://davidetorlo.it)

CFC - Cannes - 26th April 2023

## Fluid Simulations

- Which **scale** can we approximate?
- Computational **costs** vs **accuracy**
- Large Eddy Simulations (LES)
- Variational Multi-Scale (**VMS**)
- Weak Boundary conditions and **Wall-Law** for boundary layers to improve accuracy
- Turbulence

## Model order reduction

- **Parametric** context or time prediction
- Further reduce costs
- **Reduce** the computational cost for a new parameter/time
- Good approximation
- Challenges:
  - Representability (turbulence, moving discontinuities)
  - Stability

## Fluid Simulations

- Which **scale** can we approximate?
- Computational **costs** vs **accuracy**
- Large Eddy Simulations (LES)
- Variational Multi-Scale (**VMS**)
- Weak Boundary conditions and **Wall-Law** for boundary layers to improve accuracy
- Turbulence

## Model order reduction

- **Parametric** context or time prediction
- Further reduce costs
- **Reduce** the computational cost for a new parameter/time
- Good approximation
- Challenges:
  - Representability (turbulence, moving discontinuities)
  - Stability

## In this talk

- Wall Law model
- **POD-Galerkin** vs **POD-NN**
- Flow past cylinder

# Motivation

## Fluid Simulations

- Which **scale** can we approximate?
- Computational **costs** vs **accuracy**
- Large Eddy Simulations (LES)
- Variational Multi-Scale (**VMS**)
- Weak Boundary conditions and **Wall-Law** for boundary layers to improve accuracy
- Turbulence

## Model order reduction

- **Parametric** context or time prediction
- Further reduce costs
- **Reduce** the computational cost for a new parameter/time
- Good approximation
- Challenges:
  - Representability (turbulence, moving discontinuities)
  - Stability

## In this talk

- Wall Law model
- **POD-Galerkin** vs **POD-NN**
- Flow past cylinder

## Not in this talk

- Not (really) turbulent
- No special techniques for advection dominated structures

# Motivation

## Fluid Simulations

- Which **scale** can we approximate?
- Computational **costs** vs **accuracy**
- Large Eddy Simulations (LES)
- Variational Multi-Scale (**VMS**)
- Weak Boundary conditions and **Wall-Law** for boundary layers to improve accuracy
- Turbulence

## Model order reduction

- **Parametric** context or time prediction
- Further reduce costs
- **Reduce** the computational cost for a new parameter/time
- Good approximation
- Challenges:
  - Representability (turbulence, moving discontinuities)
  - Stability

## In this talk

- Wall Law model
- **POD-Galerkin** vs **POD-NN**
- Flow past cylinder

## Not in this talk

- Not (really) turbulent
- No special techniques for advection dominated structures

## In the future

- VMS-Smagorinskij model
- Hyper-reduction
- NN for wall laws

## Table of contents

---

- ① Introduction to Variational Multi-Scale (VMS) model
- ② Boundary Conditions
- ③ Reduced Order Model: Galerkin Projection
- ④ Conclusions

## Table of contents

---

① Introduction to Variational Multi-Scale (VMS) model

② Boundary Conditions

③ Reduced Order Model: Galerkin Projection

④ Conclusions

## Navier–Stokes equations (strong)

$$\begin{cases} \frac{\partial \underline{u}}{\partial t} + (\underline{u} \cdot \nabla) \underline{u} + \nabla p - 2 \operatorname{div}(\nu \nabla^s \underline{u}) = 0 \\ \nabla \cdot \underline{u} = 0 \\ \text{B.C. and I.C.} \end{cases}$$

---

<sup>1</sup>Y. Bazilevs, T.J.R. Hughes / Computers & Fluids 36 (2007) 12–26

# Navier–Stokes VMS<sup>1</sup>

---

## Navier–Stokes equations (strong)

$$\begin{cases} \frac{\partial \underline{u}}{\partial t} + (\underline{u} \cdot \nabla) \underline{u} + \nabla p - 2 \operatorname{div}(\nu \nabla^s \underline{u}) = 0 \\ \nabla \cdot \underline{u} = 0 \\ \text{B.C. and I.C.} \end{cases}$$

## Weak formulation

$$\begin{cases} \left( \underline{v}, \frac{\partial \underline{u}}{\partial t} \right)_\Omega - (\nabla \underline{v}, \underline{u} \otimes \underline{u})_\Omega + (q, \nabla \cdot \underline{u})_\Omega - \\ (\nabla \cdot \underline{v}, p)_\Omega + (\nabla^s \underline{v}, 2\nu \nabla^s \underline{u})_\Omega = 0 \\ \text{Dirichlet B.C. for } \underline{u} \text{ and } p \text{ and I.C.} \end{cases}$$

---

<sup>1</sup>Y. Bazilevs, T.J.R. Hughes / Computers & Fluids 36 (2007) 12–26

# Navier–Stokes VMS<sup>1</sup>

## Navier–Stokes equations (strong)

$$\begin{cases} \frac{\partial \underline{u}}{\partial t} + (\underline{u} \cdot \nabla) \underline{u} + \nabla p - 2 \operatorname{div}(\nu \nabla^s \underline{u}) = 0 \\ \nabla \cdot \underline{u} = 0 \\ \text{B.C. and I.C.} \end{cases}$$

## Weak formulation

$$\begin{cases} \left( \underline{v}, \frac{\partial \underline{u}}{\partial t} \right)_\Omega - (\nabla \underline{v}, \underline{u} \otimes \underline{u})_\Omega + (q, \nabla \cdot \underline{u})_\Omega - \\ (\nabla \cdot \underline{v}, p)_\Omega + (\nabla^s \underline{v}, 2\nu \nabla^s \underline{u})_\Omega = 0 \\ \text{Dirichlet B.C. for } \underline{u} \text{ and } p \text{ and I.C.} \end{cases}$$

## Variational Multi-Scale Residual based

- $\underline{u} = \underline{u}_h + \underline{u}'$  (all variables)
- Residual based

$$\underline{u}'_h = -\tau_M r_M(\underline{u}_h, p_h)$$

$$p'_h = -\tau_C r_C(\underline{u}_h)$$

$$\begin{aligned} r_M(\underline{u}_h, p_h) &= \frac{\partial \underline{u}_h}{\partial t} + \operatorname{div}(\underline{u}_h \otimes \underline{u}_h) + \nabla p_h \\ &\quad - \operatorname{div}(2\nu \nabla^s \underline{u}_h) \end{aligned}$$

$$\tau_M = \left( \frac{4}{\Delta t^2} + \underline{u}_h \cdot G \underline{u}_h + C_{inv} \nu^2 G : G \right)^{-\frac{1}{2}}$$

$$r_C(\underline{u}_h) = \operatorname{div}(\underline{u}_h), \quad \tau_C = (\tau_M \underline{g} \cdot \underline{g})^{-1}$$

$$G = \left( \frac{d\xi}{dx} \right)^T \frac{d\xi}{dx}, \quad \underline{g}_i = \sum_j \left( \frac{d\xi}{dx} \right)_{ji}$$

<sup>1</sup>Y. Bazilevs, T.J.R. Hughes / Computers & Fluids 36 (2007) 12–26

## Weak VMS formulation

$$\begin{aligned}
 a^{\text{VMS}}(\underline{u}_h, p_h, \underline{v}_h, q_h) := & \\
 & \sum_e (\underline{v}_h, \partial_t \underline{u}_h)_{\Omega_e} - (\nabla \underline{v}_h, \underline{u}_h \otimes \underline{u}_h)_\Omega + (q_h, \nabla \cdot \underline{u}_h)_\Omega \\
 & - \sum_e (\nabla \cdot \underline{v}_h, p_h)_{\Omega_e} + (\nabla^s \underline{v}_h, 2\nu \nabla^s \underline{u}_h)_\Omega \\
 & + 2 \sum_e (\underline{u}_h \cdot \nabla^s \underline{v}_h, \underline{u}'_h)_{\Omega_e} - \sum_e (\nabla \underline{v}_h, \underline{u}'_h \otimes \underline{u}'_h)_{\Omega_e} \\
 & + \sum_e (\nabla \cdot \underline{v}_h, \underline{p}'_h)_{\Omega_e} = 0
 \end{aligned}$$

## Variational Multi-Scale Residual based

- $\underline{u} = \underline{u}_h + \underline{u}'$  (all variables)
- Residual based

$$\underline{u}'_h = -\tau_M r_M(\underline{u}_h, p_h)$$

$$\underline{p}'_h = -\tau_C r_C(\underline{u}_h)$$

$$\begin{aligned}
 r_M(\underline{u}_h, p_h) = & \frac{\partial \underline{u}_h}{\partial t} + \operatorname{div}(\underline{u}_h \otimes \underline{u}_h) + \nabla p_h \\
 & - \operatorname{div}(2\nu \nabla^s \underline{u}_h)
 \end{aligned}$$

$$\tau_M = \left( \frac{4}{\Delta t^2} + \underline{u}_h \cdot G \underline{u}_h + C_{inv} \nu^2 G : G \right)^{-\frac{1}{2}}$$

$$r_C(\underline{u}_h) = \operatorname{div}(\underline{u}_h), \quad \tau_C = (\tau_M \underline{g} \cdot \underline{g})^{-1}$$

$$\begin{aligned}
 G = & \left( \frac{d\xi}{dx} \right)^T \frac{d\xi}{dx}, \quad \underline{g}_i = \sum_j \left( \frac{d\xi}{dx} \right)_{ji}
 \end{aligned}$$

<sup>1</sup>Y. Bazilevs, T.J.R. Hughes / Computers & Fluids 36 (2007) 12–26

## Weak VMS formulation

$$\begin{aligned} a^{\text{VMS}}(\underline{u}_h, p_h, \underline{v}_h, q_h) := & \\ & \sum_e (\underline{v}_h, \partial_t \underline{u}_h)_{\Omega_e} - (\nabla \underline{v}_h, \underline{u}_h \otimes \underline{u}_h)_\Omega + (q_h, \nabla \cdot \underline{u}_h)_\Omega \\ & - \sum_e (\nabla \cdot \underline{v}_h, p_h)_{\Omega_e} + (\nabla^s \underline{v}_h, 2\nu \nabla^s \underline{u}_h)_\Omega \\ & + 2 \sum_e (\underline{u}_h \cdot \nabla^s \underline{v}_h, \underline{u}'_h)_{\Omega_e} - \sum_e (\nabla \underline{v}_h, \underline{u}'_h \otimes \underline{u}'_h)_{\Omega_e} \\ & + \sum_e (\nabla \cdot \underline{v}_h, p'_h)_{\Omega_e} = 0 \end{aligned}$$

## Advantages

- Coarse scale  $\implies$  Discretized
- Fine scale  $\implies$  Modeled
- Extra accuracy by modeling higher order terms without solving them
- Stabilization effect: we can use  $\mathbb{P}^p$  for both velocity and pressure (no need of  $\mathbb{P}^1\mathbb{P}^2$  formulations)
- Duality: LES modeling and stabilization

# Table of contents

---

1 Introduction to Variational Multi-Scale (VMS) model

2 Boundary Conditions

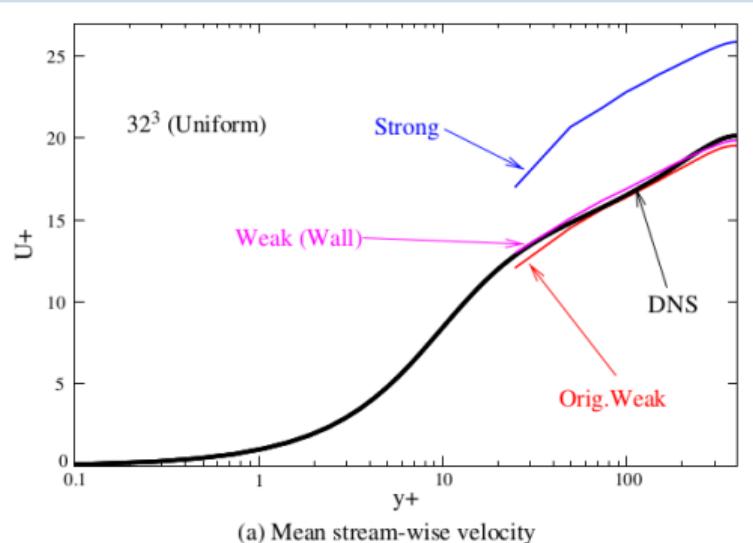
3 Reduced Order Model: Galerkin Projection

4 Conclusions

## Boundary conditions

### No slip Boundary conditions

- Can create boundary layers
- If strongly huge impact on the solution



### Weak Enforcement of no slip BC

$$\begin{aligned} & a_{\text{weakBC}}^{\text{VMS}}(\underline{u}_h, p_h, \underline{v}_h, q_h) := a^{\text{VMS}}(\underline{u}_h, p_h, \underline{v}_h, q_h) \\ & - \sum_b (\underline{v}_h, 2\nu \nabla^s \underline{u}_h \cdot \underline{n})_{\partial\Omega \cap \Gamma_b} \\ & - \sum_b (2\nu \nabla^s \underline{v}_h \cdot \underline{n}, \underline{u}_h - \underline{0})_{\partial\Omega \cap \Gamma_b} \\ & + \sum_b \left( \underline{v}_h \frac{C_b^I \nu}{h_b}, \underline{u}_h - \underline{0} \right)_{\partial\Omega \cap \Gamma_b} = 0 \end{aligned}$$

- Consistency term
- Adjoint consistency term
- Penalization of Dirichlet BC

## Spalding Wall Law<sup>34</sup>

Weak penalty for no slip condition

$$\sum_b \left( \underline{v}_h \frac{C'_b \nu}{h_b}, \underline{u}_h - \underline{0} \right)_{\partial\Omega \cap \Gamma_b}$$

- $C'_b = 4$  user set coefficient
- $h_b = 2 (\underline{n}^T \underline{G} \underline{n})^{-1/2}$  wall-normal element mesh size

### Spalding Wall law

- More **physical intuition**
- Exploiting notion of fully developed turbulence
- Boundary layer typically not interesting for turbulent phenomena
- No-slip Dirichlet BC replaced by **traction Neumann boundary**

$$\sum_b \left( \underline{v}_h, u^{*2} \frac{\underline{u}_h}{\|\underline{u}_h\|} \right)_{\partial\Omega \cap \Gamma_b}$$

- $u^{*2}$  magnitude of the **wall shear stress**
- Consistent with the “law of the wall”

<sup>3</sup>D.B. Spalding, A single formula for the law of the wall, J. Appl. Mech. 28 (1961) 444–458

<sup>4</sup>Y. Bazilevs et al. / Comput. Methods Appl. Mech. Engrg. 196 (2007) 4853–4862

## Spalding Wall Law

### Spalding Law

$$\sum_b \left( \underline{v}_h, u^{*2} \frac{\underline{u}_h}{\|\underline{u}_h\|} \right)_{\partial\Omega \cap \Gamma_b}$$

- Empirical relation between the mean fluid speed and the normal distance to the wall
- Spalding Law

$$y^+ \stackrel{!}{=} f(u^+) = u^+ + e^{-\chi B} \left( e^{\chi u^+} - 1 - \chi u^+ - \frac{(\chi u^+)^2}{2} - \frac{(\chi u^+)^3}{6} \right),$$

$y^+$  :=  $\frac{yu^*}{\nu}$  distance from the wall in nondimensional wall units,

$u^+$  :=  $\frac{\|\underline{u}_h\|}{u^*}$  mean fluid speed in nondimensional wall units,

$$\chi = 0.4, B = 5.5.$$

## Spalding Law

$$\sum_b \left( \underline{v}_h, u^{*2} \frac{\underline{u}_h}{\|\underline{u}_h\|} \right)_{\partial\Omega \cap \Gamma_b} = \sum_b \left( \underline{v}_h \frac{u^{*2}}{\|\underline{u}_h\|}, \underline{u}_h - \underline{0} \right)_{\partial\Omega \cap \Gamma_b} = \sum_b (\underline{v}_h \tau_B, \underline{u}_h - \underline{0})_{\partial\Omega \cap \Gamma_b}, \quad \tau_B := \frac{u^{*2}}{\|\underline{u}_h\|}$$

- $\tau_B$  makes the connection with the weak formulation where  $\tau_B = \frac{C_b^l \nu}{h_b}$
- One extra scalar nonlinear equation to be solved  $y^+ \stackrel{!}{=} f(u^+)$  for each boundary cell (not too expensive)
- Rewrite the equation in terms of  $\tau_B$
- $r(\tau_B) = 0$  with  $r'(\tau) > 0$  and  $r''(\tau) < 0$  for  $\tau > 0$
- Newton's method converges if  $\tau^0$  small enough (worst case bisection not too expensive)
- Initial guess  $\tau^0 = \frac{C_b^l \nu}{h_b}$  (from weak formulation)

## FOM

- $\mathbb{P}^2 - \mathbb{P}^2$  **continuous Galerkin FEM** formulation
- Residual based Variational MultiScale discrete model
- Boundary consistency terms
- Weak penalty for no-slip BC
- **Spalding Wall Law** (very fast 1% cost)

## FOM

- $\mathbb{P}^2 - \mathbb{P}^2$  continuous Galerkin FEM formulation
- Residual based Variational MultiScale discrete model
- Boundary consistency terms
- Weak penalty for no-slip BC
- Spalding Wall Law (very fast 1% cost)

## Test: Flow Past Cylinder

$\mathcal{R} = [0, 2.2] \times [-0.41, 0.41]$ ,  $\mathcal{C} = \mathcal{B}([0.2, 0], 0.05)$ ,  
 $D = \mathcal{R} \setminus \mathcal{C}$ ,  $T_{end} = 3$ ,  $N_h = 3 \cdot 122145$ ,  
 $u_{in} = (\mu_1, 0)$ ,  $\nu = \mu_2$ ,  
Test 1:  $Re \in [50, 2.5 \cdot 10^3]$ ,  
 $\mu_1 \in [0.5, 5.0]$ ,  $\mu_2 \in [2 \cdot 10^{-4}, 10^{-3}]$ ,  
No slip BC on top, bottom and circle,

## FOM

- $\mathbb{P}^2 - \mathbb{P}^2$  continuous Galerkin FEM formulation
- Residual based Variational MultiScale discrete model
- Boundary consistency terms
- Weak penalty for no-slip BC
- Spalding Wall Law (very fast 1% cost)

## Test: Flow Past Cylinder

$\mathcal{R} = [0, 2.2] \times [-0.41, 0.41]$ ,  $\mathcal{C} = \mathcal{B}([0.2, 0], 0.05)$ ,  
 $D = \mathcal{R} \setminus \mathcal{C}$ ,  $T_{end} = 3$ ,  $N_h = 3 \cdot 122145$ ,  
 $u_{in} = (\mu_1, 0)$ ,  $\nu = \mu_2$ ,  
Test 2:  $Re \in [2.5 \cdot 10^3, 2.5 \cdot 10^5]$ ,  
 $\mu_1 \in [0.5, 5.0]$ ,  $\mu_2 \in [2 \cdot 10^{-6}, 2 \cdot 10^{-5}]$ ,  
No slip BC on top, bottom and circle,

## Table of contents

---

① Introduction to Variational Multi-Scale (VMS) model

② Boundary Conditions

③ Reduced Order Model: Galerkin Projection

④ Conclusions

## Solution Manifold Compression

- Proper Orthogonal Decomposition (**POD**)
- Collection of **snapshots**  
 $\{[u_h, p_h, \tau_{B,h}](t^{i_0}, \mu_1^{i_1}, \mu_2^{i_2})\}_{i \in \mathcal{T}} \in V_h$
- Generation of  **$V_{RB}$**  component by component
- No need of supremizer<sup>5</sup>

---

<sup>5</sup>Stabile, Giovanni, et al. "A reduced order variational multiscale approach for turbulent flows." *Advances in Computational Mathematics* 45 (2019): 2349-2368.

# Reduced Order Model<sup>5</sup>

## Solution Manifold Compression

- Proper Orthogonal Decomposition (**POD**)
- Collection of **snapshots**  
 $\{[u_h, p_h, \tau_{B,h}](t^{i_0}, \mu_1^{i_1}, \mu_2^{i_2})\}_{i \in \mathcal{T}} \in V_h$
- Generation of  **$V_{RB}$**  component by component
- No need of supremizer<sup>5</sup>

## Reconstruction

- **POD-Galerkin**
  - $F(u_h) = 0 \implies V_{RB}^T F(V_{RB} u_{RB}) = 0$
  - Less equations
  - **Hyper-reduction** needed to decrease costs (not today)
  - **Physics** based
  - Less nonlinear iterations
  - For the moment no computational advantage
- **POD-NN**
  - NN learn map  $u_{RB}(t, \mu_1, \mu_2)$  from training set
  - Very **fast** reconstruction
  - **No physics** involved
  - Needs **many snapshots** for different parameters (too expensive offline phase)

<sup>5</sup>Stabile, Giovanni, et al. "A reduced order variational multiscale approach for turbulent flows." Advances in Computational Mathematics 45 (2019): 2349-2368.

# Reduced Order Model<sup>5</sup>

## Solution Manifold Compression

- Proper Orthogonal Decomposition (**POD**)
- Collection of **snapshots**  
 $\{[u_h, p_h, \tau_{B,h}](t^{i_0}, \mu_1^{i_1}, \mu_2^{i_2})\}_{i \in \mathcal{T}} \in V_h$
- Generation of  **$V_{RB}$**  component by component
- No need of supremizer<sup>5</sup>

## Spalding coefficient reconstruction

- For  $\tau_{B,h}$  POD-Galerkin and then Newton
- Problem: **Newton does not converge** for reduced  $\tau_{B,RB}$  equation
- Multi-layer perceptron NN  $u_{RB} \rightarrow \tau_{B,RB}$ ?

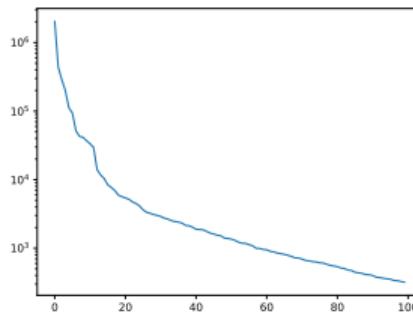
## Reconstruction

- **POD-Galerkin**
  - $F(u_h) = 0 \implies V_{RB}^T F(V_{RB} u_{RB}) = 0$
  - Less equations
  - **Hyper-reduction** needed to decrease costs (not today)
  - **Physics** based
  - Less nonlinear iterations
  - For the moment no computational advantage
- **POD-NN**
  - NN learn map  $u_{RB}(t, \mu_1, \mu_2)$  from training set
  - Very **fast** reconstruction
  - **No physics** involved
  - Needs **many snapshots** for different parameters (too expensive offline phase)

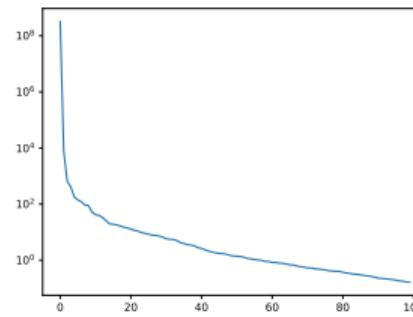
<sup>5</sup>Stabile, Giovanni, et al. "A reduced order variational multiscale approach for turbulent flows." *Advances in Computational Mathematics* 45 (2019): 2349-2368.

## POD results: **Test1**, 20 parameters, 150 timesteps

*u*

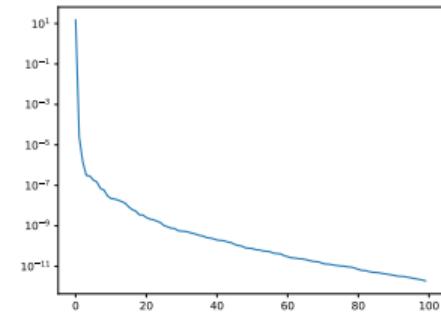
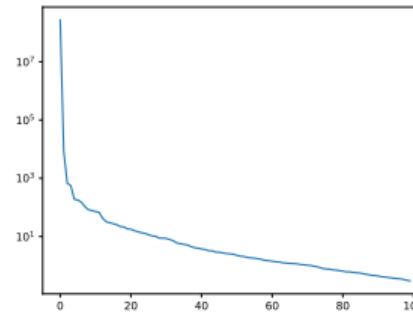
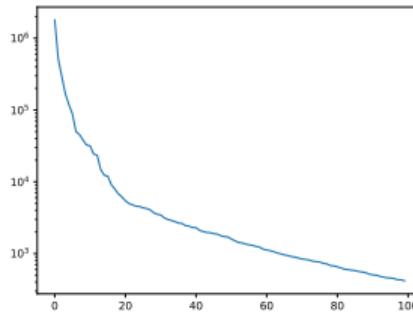


*p*



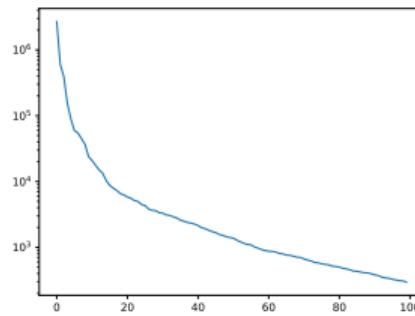
*T*

⇐ Weak BC, Spalding BC ↓

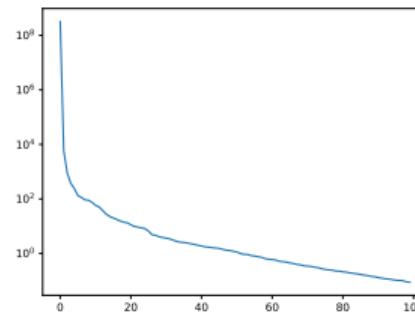


## POD results: **Test2**, 20 parameters, 150 timesteps

*u*

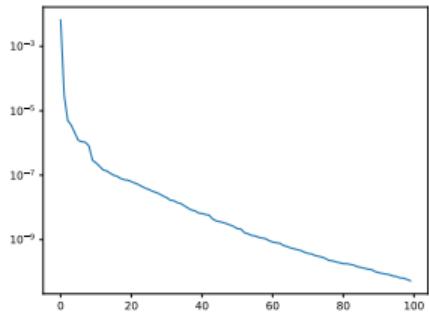
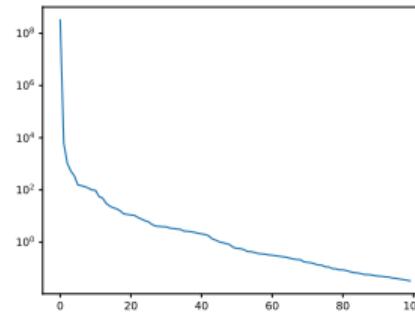
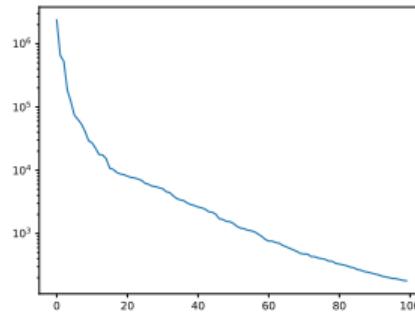


*p*



*T*

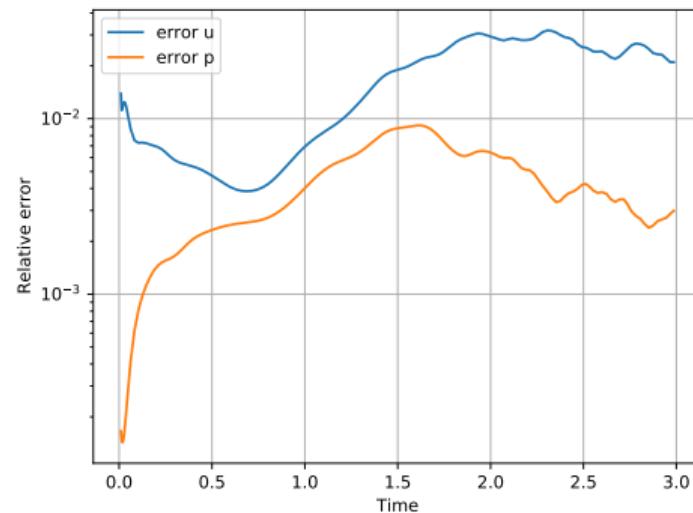
⇐ Weak BC, Spalding BC ↓



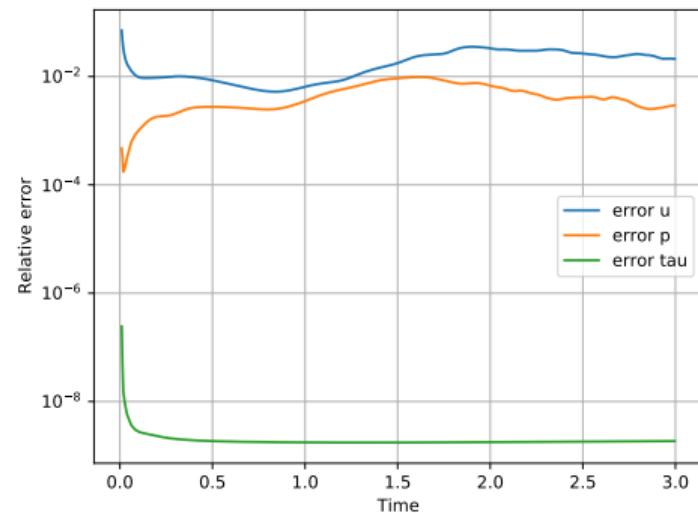
## POD projection error: Test1, $(u_{in}, \nu) = (1, 0.001)$

$$N_{POD}^u = N_{POD}^p = 100, N_{POD}^\tau = 30$$

Weak BC



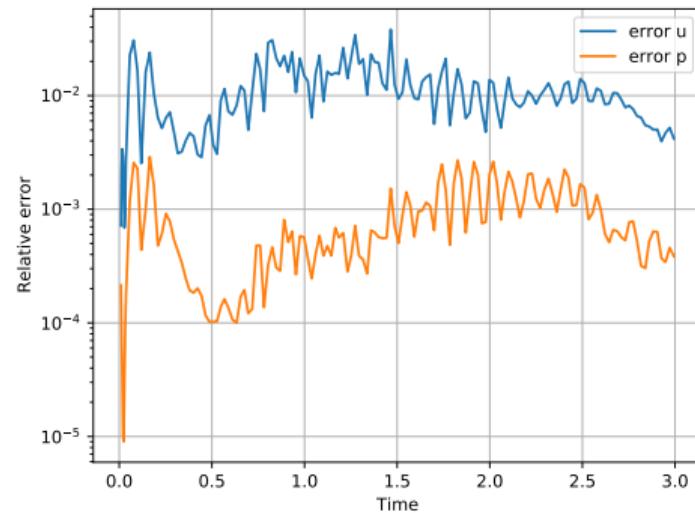
Spalding



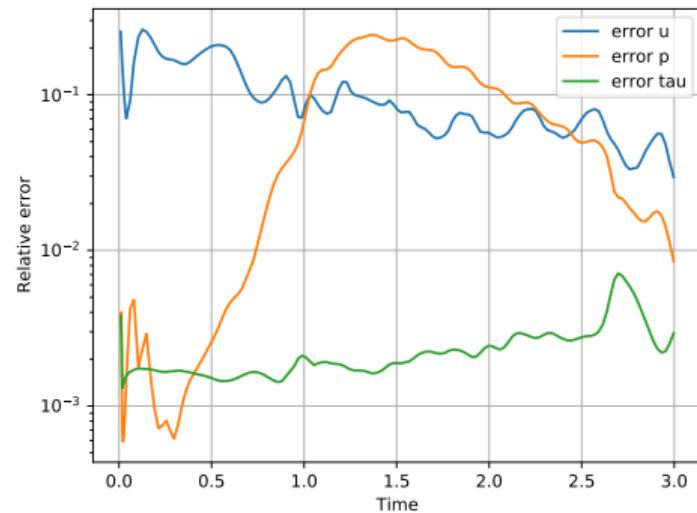
## POD projection error: Test2, $(u_{in}, \nu) = (1, 2 \cdot 10^{-6})$

$$N_{POD}^u = N_{POD}^p = 100, N_{POD}^\tau = 30$$

Weak BC



Spalding

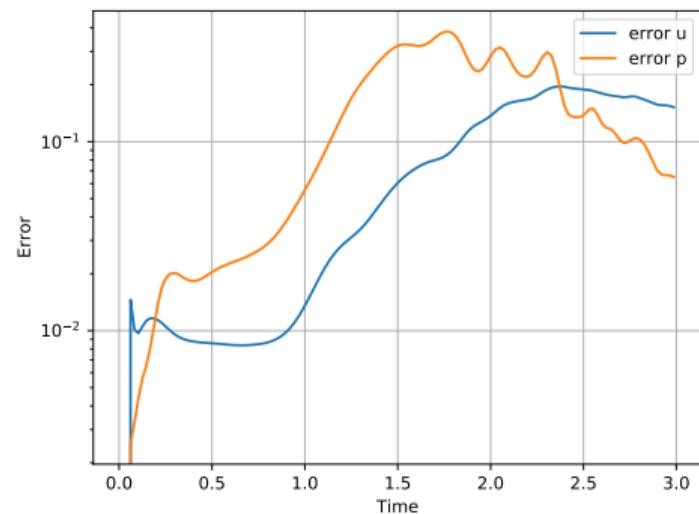


## POD-Galerkin error: **Test1**, $(u_{in}, \nu) = (1, 0.001)$

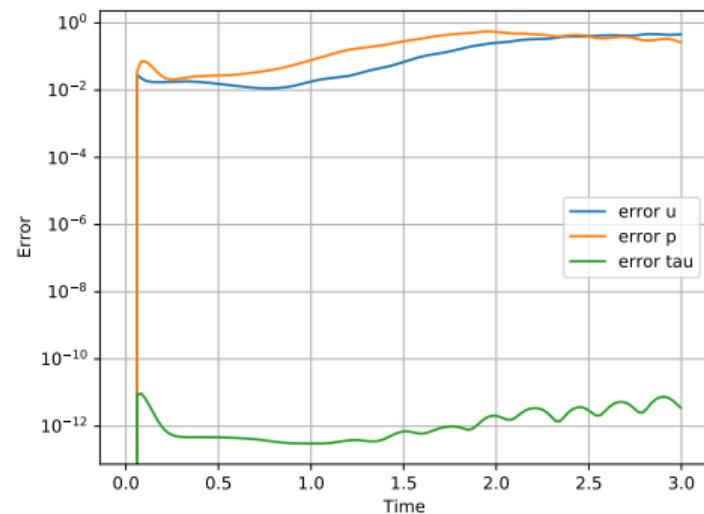
---

$$N_{POD}^u = N_{POD}^p = 100, \tau \text{ exact}$$

Weak BC

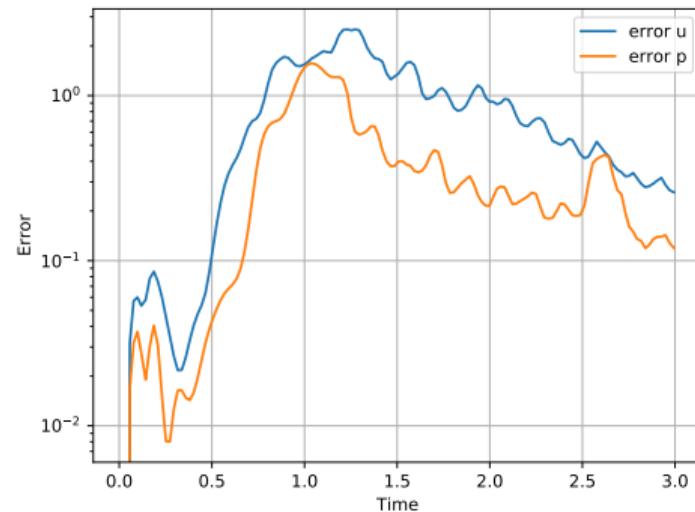


Spalding

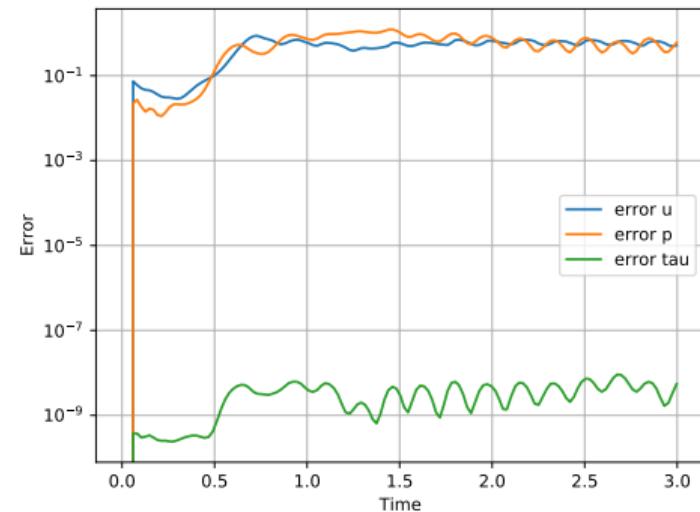


$$N_{POD}^u = N_{POD}^p = 100, \tau \text{ exact}$$

Weak BC



Spalding



## **Test1 Weak BC: POD-Galerkin (top) vs FOM (bottom)**

---

## **Test1 Spalding BC: POD-Galerkin (top) vs FOM (bottom)**

---

## **Test2 Weak BC: POD-Galerkin (top) vs FOM (bottom)**

---

## Test2 Spalding BC: POD-Galerkin (top) vs FOM (bottom)

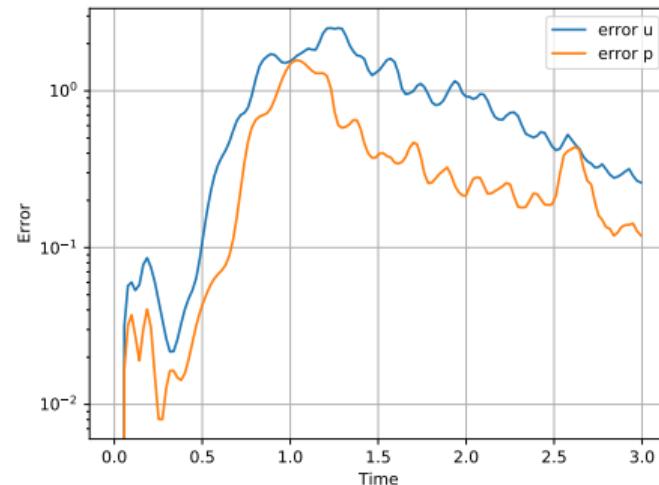
### Vortex shedding start

- The **vortex shedding** in FOM is dictated purely by the (triangular) mesh
- No reasons why also in ROM it should start at the same time and in the same direction
- Starting the ROM simulation after vortex shedding (at time  $t = 1$ )

### Weak BC Test2 ROM from $t = 0$

#### Vortex shedding start

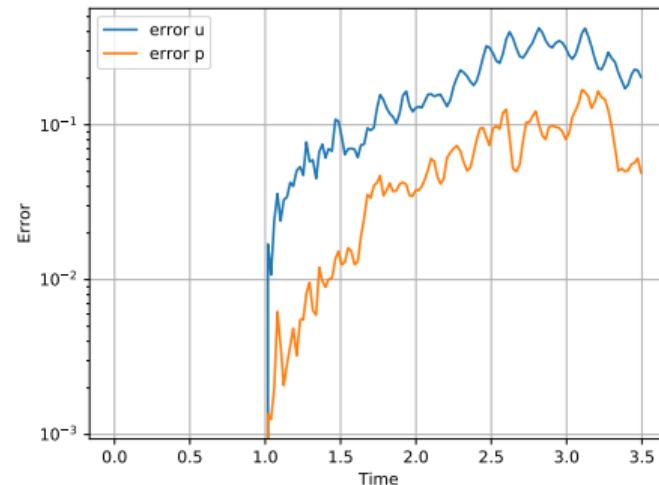
- The **vortex shedding** in FOM is dictated purely by the (triangular) mesh
- No reasons why also in ROM it should start at the same time and in the same direction
- Starting the ROM simulation after vortex shedding (at time  $t = 1$ )



### Weak BC Test2 ROM from $t = 1$

#### Vortex shedding start

- The **vortex shedding** in FOM is dictated purely by the (triangular) mesh
- No reasons why also in ROM it should start at the same time and in the same direction
- Starting the ROM simulation after vortex shedding (at time  $t = 1$ )



Weak BC Test2 ROM from  $t = 1$

### Vortex shedding start

- The **vortex shedding** in FOM is dictated purely by the (triangular) mesh
- No reasons why also in ROM it should start at the same time and in the same direction
- Starting the ROM simulation after vortex shedding (at time  $t = 1$ )

### Weak BC vs Spalding

- Spalding has larger error in representation
- Spalding has little worse behavior in POD-Galerkin
- In past simulations,  $\tau_B$  computed as FOM ( $\approx 250$  DOFs)

## POD-NN

- Training set as POD: 20 params, 150 timesteps (3000 snapshots)
- Goal: learn map  $(t, \mu_0, \mu_1) \rightarrow u_{RB}$
- NN setting
  - multi-layer perceptron
  - 4 hidden nodes
  - 100 neurons each
  - Various activation functions
- For  $u$  and  $p$  the loss struggle at decaying
- For  $\tau$  already better results, but dangerous to be used alone (time consistency)

## Prediction of $\tau$

It might be safer to predict  $\tau_B(u)$

- Learn  $u_{RB} \rightarrow \tau_{B,RB}$
- NN as before
- Errors  $\approx 6\%$  on a test set
- Not really helpful in reducing the computational costs (solving for  $\tau_B$  already cheap (1% of all costs))
- Still not physics based

# Table of contents

---

- 1 Introduction to Variational Multi-Scale (VMS) model
- 2 Boundary Conditions
- 3 Reduced Order Model: Galerkin Projection
- 4 Conclusions

## Summary and perspectives<sup>6</sup>

### Summary

- LES-VMS model for Navier-Stokes in FEM
- Weak Boundary Conditions
- Spalding Law
- POD-Galerkin
- POD-NN (just started)

### Perspectives

- Hyper-reduction (EIM or overcollocation)
- Extend to other models with Local Projection Stabilization (LPS) onto sub-filter scale<sup>6</sup>
- 3D turbulent simulations
- Improve the architecture for POD-NN to have a comparison with POD-Galerkin

<sup>6</sup>N. Ahmed, T. C. Rebollo, V. John and S. Rubino. Analysis of a Full Space–Time Discretization of the Navier–Stokes Equations by a Local Projection Stabilization Method. IMA Journal of Numerical Analysis, Vol. 37, pp. 1437–1467, 2017.

## Literature

- Y. Bazilevs, T.J.R. Hughes. "Weak imposition of Dirichlet boundary conditions in fluid mechanics." *Computers & Fluids* 36 (2007) 12–26.
- Y. Bazilevs, C. Michler, V.M. Calo, T.J.R. Hughes. "Weak Dirichlet boundary conditions for wall-bounded turbulent flows." *Comput. Methods Appl. Mech. Engrg.* 196 (2007) 4853–4862.
- D.B. Spalding, A single formula for the law of the wall, *J. Appl. Mech.* 28 (1961) 444–458.
- Stabile, Giovanni, et al. "A reduced order variational multiscale approach for turbulent flows." *Advances in Computational Mathematics* 45 (2019): 2349–2368.
- N. Ahmed, T. C. Rebollo, V. John and S. Rubino. Analysis of a Full Space–Time Discretization of the Navier–Stokes Equations by a Local Projection Stabilization Method. *IMA Journal of Numerical Analysis*, Vol. 37, pp. 1437–1467, 2017.

## Literature

- Y. Bazilevs, T.J.R. Hughes. "Weak imposition of Dirichlet boundary conditions in fluid mechanics." *Computers & Fluids* 36 (2007) 12–26.
- Y. Bazilevs, C. Michler, V.M. Calo, T.J.R. Hughes. "Weak Dirichlet boundary conditions for wall-bounded turbulent flows." *Comput. Methods Appl. Mech. Engrg.* 196 (2007) 4853–4862.
- D.B. Spalding, A single formula for the law of the wall, *J. Appl. Mech.* 28 (1961) 444–458.
- Stabile, Giovanni, et al. "A reduced order variational multiscale approach for turbulent flows." *Advances in Computational Mathematics* 45 (2019): 2349–2368.
- N. Ahmed, T. C. Rebollo, V. John and S. Rubino. Analysis of a Full Space–Time Discretization of the Navier–Stokes Equations by a Local Projection Stabilization Method. *IMA Journal of Numerical Analysis*, Vol. 37, pp. 1437–1467, 2017.

THANK YOU!