Data-based approach for time-correlated closures of turbulence models

Júlia Domingues Lemos - UFRJ Alexei Mailybaev - IMPA

September 25th, 2023

Turbulence



$$Re = \frac{UL}{\nu}, \quad \eta \sim \left(\frac{\nu^3}{\epsilon}\right)^{1/4}, \quad \epsilon \sim \frac{U^3}{L}$$

For a grid with spacing smaller than the Kolmogorov length scale:

$$N = \left(\frac{L}{\eta}\right)^3 \sim Re^{9/4} \tag{1}$$

• Cup of coffee: $Re = 10^4$, grid must have 10^9 points (8GB per array)

• Moving car: $Re = 10^9$, grid must have 10^{16} points (80000000GB per array)

Road map



4/43 4/43 N.S.

$$\partial_t u_i(\mathbf{n}) = -i \frac{2\pi n_j}{L} \sum_{\mathbf{n}'} \left(\delta_{il} - \frac{n_i n_l'}{n_j n_j} \right) u_j(\mathbf{n}') u_l(\mathbf{n} - \mathbf{n}') -\nu k_j k_j u_i(\mathbf{n}) + f_i(\mathbf{n}).$$
(2)

L'vov et al. (1995) and Gledzer, Ohkitani, Yamada (1973,1988)

$$\frac{du_n}{dt} = i \left(ak_{n+1}u_{n+2}u_{n+1}^* + bk_n u_{n+1}u_{n-1}^* + ck_{n-1}u_{n-1}u_{n-2} \right) - \nu k_n^2 u_n + f_n \tag{3}$$

About Sabra

Parameters

•
$$a = 1, b = -\frac{1}{2}, c = \frac{1}{2}, \lambda = 2, k_0 = 1 \text{ and } k_n = k_0 \lambda^n$$
:

• Inviscid invariants:

$$E = \sum |u_n|^2 \tag{4}$$
$$H = \sum (-1)^n k_n |u_n|^2 \tag{5}$$

Numerics

• n = 30

•
$$u_{-1} = u_0 = u_{n+1} = u_{n+2} = 0$$
 (for now)
• $\nu = 10^{-8}, f_1 = 1 + i, u_n(0) = k^{-1/3} e^{iq_n}$ for $n = 1, 2$

Numerics



7/43 7/43

Intermittency



Data-based approach for time-correlated closur

8/43 8 / 43

Numerics



Road map



023 10 / 43

$$z_n = w_n e^{i\Delta_n},$$

$$w_n = \left| \frac{u_n}{u_{n-1}} \right|,$$

$$\Delta_n = \arg(u_n) - \arg(u_{n-1}) - \arg(u_{n-2}),$$
(6)
(7)
(8)

Data-based approach for time-correlated closur September 25th, 2023 11/43

The need for time conditioning

PHYSICAL REVIEW E 95, 043108 (2017)

Optimal subgrid scheme for shell models of turbulence



12/4312/43

Data-based approach for time-correlated closur

Rescaled Model

Choose a reference shell m [Mailybaev, 2021]

$$T_m(t) = \left(k_0^2 U^2 + \sum_{n < m} k_n^2 |u_n|^2\right)^{-1/2}$$

Define the change of variables

$$\tau = \int_0^t \frac{dt'}{T_m(t')} \tag{10}$$

(9)

13/43

$$\mathcal{U}_N = k_m T_m(t) u_{N+m}(t) \tag{11}$$

- Shrinks long periods of time where nothing happens
- Stretches short periods of time where a lot happens
- Gives a new sense of symmetry

Rescaled Model



Hidden Symmetry



Data-based approach for time-correlated closur

Road map



Rescaled Complete System

$$\frac{d\mathcal{U}_N}{d\tau} = i(k_{N+1}\mathcal{U}_{N+2}\mathcal{U}_{N+1}^* - \frac{1}{2}k_N\mathcal{U}_{N+1}\mathcal{U}_{N-1}^* + \frac{1}{2}k_{N-1}\mathcal{U}_{N-1}\mathcal{U}_{N-2}) + (\xi + \xi_f)\mathcal{U}_N + \nu k_m^2 T_m \Big(-k_N^2 + \sum_{J<0} k_J^4 |\mathcal{U}_J|^2 \Big)\mathcal{U}_N + T_m^2 k_m f_{N+m}$$
(12)

where

$$\xi = \sum_{N<0} k_N^3 \operatorname{Im} \left(2\mathcal{U}_N^* \mathcal{U}_{N+1}^* \mathcal{U}_{N+2} - \frac{1}{2} \right) \qquad \mathcal{U}_{N-1}^* \mathcal{U}_N^* \mathcal{U}_{N+1} - \frac{1}{4} \mathcal{U}_{N-1}^* \mathcal{U}_N \mathcal{U}_{N-2}^* \right)$$
(13)

$$\xi_f = -T_m^2 \sum_{N < 0} k_{N+m} k_N \operatorname{Re}\left(\mathcal{U}_N^* f_{N+m}\right)$$
(14)

$$T_m = \frac{1}{k_0 U} \left(1 - \sum_{N < 0} k_N^2 |\mathcal{U}_N|^2 \right)^{1/2}$$
(15)

17/43

Data-based approach for time-correlated closur September 25th, 2023

Closure for Reduced Models

• Choose a shell s in the inertial range

$$\frac{du_n}{dt} = i \left(ak_{n+1}u_{n+2}u_{n+1}^* + bk_nu_{n+1}u_{n-1}^* + ck_{n-1}u_{n-1}u_{n-2} \right) - \nu k_n^2 u_n + f_n$$
(16)

- for $n = 1, \ldots, s$
 - Choose m = s + 1

$$\frac{d\mathcal{U}_N}{d\tau} = i(k_{N+1}\mathcal{U}_{N+2}\mathcal{U}_{N+1}^* - \frac{1}{2}k_N\mathcal{U}_{N+1}\mathcal{U}_{N-1}^* + \frac{1}{2}k_{N-1}\mathcal{U}_{N-1}\mathcal{U}_{N-2}) + (\xi + \xi_f)\mathcal{U}_N + \nu k_m^2 T_m (-k_N^2 + \sum_{J<0} k_J^4 |\mathcal{U}_J|^2)\mathcal{U}_N + T_m^2 k_m f_{N+m}$$
(17)

For $N = 1 - s, \dots, -1$.

We need expressions for u_{s+1} and u_{s+2} .

- or -

We need expressions for \mathcal{U}_0 and \mathcal{U}_1 .

Kolmogorov's Closure

[Biferale, Mailybaev, Parisi, 2017]

$$\frac{|u_{s+2}|}{|u_{s+1}|} = \frac{|u_{s+1}|}{|u_s|} = \lambda^{-1/3},$$

$$\Delta_{s+1} = \Delta_{s+2} = \frac{\pi}{2}.$$
(18)
(19)

Choosing the shell velocities' phases θ_n as to satisfy $\theta_n = \theta_{n-1} + \theta_{n-2}$,

$$u_{s+1} = |u_s|\lambda^{-1/3}e^{i(\frac{\pi}{2} + \theta_s + \theta_{s-1})},$$
(20)

$$u_{s+2} = |u_{s+1}|\lambda^{-1/3}e^{i(\frac{\pi}{2} + \theta_{s+1} + \theta_s)}.$$
(21)

Kolmogorov's Closure

$$\mathcal{U}_{0} = |\mathcal{U}_{-1}|\lambda^{-1/3}e^{i(\frac{\pi}{2}+\alpha_{-1}+\alpha_{-2})},$$

$$\mathcal{U}_{1} = |\mathcal{U}_{0}|\lambda^{-1/3}e^{i(\frac{\pi}{2}+\alpha_{0}+\alpha_{-1})}$$

$$\alpha_{N} = \arg(\mathcal{U}_{N}) = \theta_{N+m}$$
(22)
(23)



Data-based approach for time-correlated closur

How do we go about making this *probabilistic*?



Gaussian Mixture Models



Say we have N data samples from a distribution we don't know.

$$p(\mathbf{x}) = \sum_{k=1}^{K} \pi_k \mathcal{N}(\mathbf{x} | \boldsymbol{\mu}_k, \boldsymbol{\Sigma}_k)$$
(24)
$$\mathcal{N}(\mathbf{x} | \boldsymbol{\mu}, \boldsymbol{\Sigma}) = \frac{1}{(2\pi)^{D/2} |\boldsymbol{\Sigma}|^{1/2}} \exp\left(-\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu})\right)$$
(25)
$$\sum_{n=1}^{k} \pi_k = 1.$$
(26)

$$\gamma(z_{nk}) = \frac{\pi_k \mathcal{N}(\mathbf{x}_n | \boldsymbol{\mu}_k, \boldsymbol{\Sigma}_k)}{\sum_{j=1}^k \pi_j \mathcal{N}(\mathbf{x}_n | \boldsymbol{\mu}_j, \boldsymbol{\Sigma}_j)}$$
(27)

$$\mathcal{Q}(\boldsymbol{\theta}^*, \boldsymbol{\theta}) = \sum_{n=1}^{N} \sum_{k=1}^{K} \gamma(z_{nk}) (\ln \pi_k^* + \ln \mathcal{N}(\mathbf{x}_n | \boldsymbol{\mu}_k^*, \boldsymbol{\Sigma}_k^*)) + \lambda \left(\sum_{k=1}^{K} \pi_k^* - 1 \right)$$
(28)

$$\widehat{\boldsymbol{\theta}} = \underset{\boldsymbol{\theta}^*}{\arg\max} \, \mathcal{Q}(\boldsymbol{\theta}^*, \boldsymbol{\theta}) \tag{29}$$



Road map



Data-based approach for time-correlated closur September 25th, 2023



Only modules:

$$\mathcal{U}_0 = 2^{z_0} e^{i(\frac{\pi}{2} + \alpha_{-1} + \alpha_{-2})},\tag{30}$$

$$\mathcal{U}_1 = 2^{z_1} e^{i(\frac{\pi}{2} + \alpha_0 + \alpha_{-1})},\tag{31}$$

$$\mathbf{z} = (z_0, z_1) \sim g(\mathbf{z}). \tag{32}$$

Modules and phases:

$$\mathcal{U}_0 = 2^{z_0} e^{i(z_1 + \alpha_{-1} + \alpha_{-2})},\tag{33}$$

$$\mathcal{U}_1 = 2^{z_2} e^{i(z_3 + \alpha_0 + \alpha_{-1})},\tag{34}$$

$$\mathbf{z} = (z_0, z_1, z_2, z_3) \sim g(\mathbf{z}).$$
 (35)



Data-based approach for time-correlated closur September 25th, 2023





Time conditioning

Conditioning to modules of three closest shells:

$$\mathcal{U}_0 = 2^{z_0} e^{i(\frac{\pi}{2} + \alpha_{-1} + \alpha_{-2})},\tag{36}$$

$$\mathcal{U}_1 = 2^{z_1} e^{i(\frac{\pi}{2} + \alpha_0 + \alpha_{-1})},\tag{37}$$

$$\mathbf{z} = (z_0, z_1) \sim g(\mathbf{z}|\log_2 |\mathcal{U}_{-3}(\tau - \Delta \tau)|, \log_2 |\mathcal{U}_{-2}(\tau - \Delta \tau)|, \log_2 |\mathcal{U}_{-1}(\tau - \Delta \tau)|), \qquad (38)$$

Conditioning to modules and phases of themselves:

$$\mathcal{U}_0 = 2^{z_0} e^{i(z_1 + \alpha_{-1} + \alpha_{-2})},\tag{39}$$

$$\mathcal{U}_1 = 2^{z_2} e^{i(z_3 + \alpha_0 + \alpha_{-1})},\tag{40}$$

$$\mathbf{z} = (z_0, z_1, z_2, z_3) \sim g(\mathbf{z} | \log_2 |\mathcal{U}_0(\tau - \Delta \tau)|, \Delta_0, \log_2 |\mathcal{U}_1(\tau - \Delta \tau)|, \Delta_1).$$
(41)

Time conditioning, $\Delta \tau = 2.4$



Time conditioning





Time conditioning, $\Delta \tau = 2.4$



Data-based approach for time-correlated closur September 25th, 2023

Time conditioning





	$ \mathcal{U}_0 $	$ \mathcal{U}_1 $	Δ_0	Δ_1	Conditioning
Half closure	2^{z_0}	$ \mathcal{U}_0 \lambda^{-1/3}$	$\pi/2$	$\pi/2$	×
Joint	2^{z_0}	2^{z_1}	$\pi/2$	$\pi/2$	×
Simple cond	2^{z_0}	$ \mathcal{U}_0 \lambda^{-1/3}$	$\pi/2$	$\pi/2$	$ \mathcal{U}_{-1} $ at $\tau - \Delta \tau$
Joint cond	2^{z_0}	2^{z_1}	$\pi/2$	$\pi/2$	$ \mathcal{U}_{-1} $ at $ au - \Delta au$
3-Clos	2^{z_0}	2^{z_1}	$\pi/2$	$\pi/2$	$ \mathcal{U}_{-3} , \mathcal{U}_{-2} , \mathcal{U}_{-1} $ at $\tau - \Delta \tau$
3-Clos 9	2^{z_0}	2^{z_1}	$\pi/2$	$\pi/2$	$ \mathcal{U}_{-3} , \mathcal{U}_{-2} , \mathcal{U}_{-1} $ at $\tau - \Delta \tau$
Long	2^{z_0}	2^{z_1}	$\pi/2$	$\pi/2$	$ \mathcal{U}_{-s} ,\ldots, \mathcal{U}_{-1} $ at $\tau - \Delta \tau$
Joint phases	2^{z_0}	2^{z_2}	z_1	z_3	×
Joint cond phases	2^{z_0}	2^{z_2}	z_1	z_3	$ \mathcal{U}_{-1} , \Delta -1 \text{ at } \tau - \Delta \tau$
Self	2^{z_0}	2^{z_2}	z_1	z_3	$ \mathcal{U}_0 , \mathcal{U}_1 , \Delta_0, \Delta_1 ext{ at } au - \Delta au$
Self 9	2^{z_0}	2^{z_2}	z_1	z_3	$ \mathcal{U}_0 , \mathcal{U}_1 , \Delta_0, \Delta_1 ext{ at } au - \Delta au$
Clobal	2^{z_0}	$9^{\mathbf{z}_2}$	~.	~~	$ \mathcal{U}_{-2} , \mathcal{U}_{-1} , \Delta_{-2}, \Delta_{-1} \text{ at } \tau$
Giobai	2 0	2 -	~1	~3	$ \mathcal{U}_{-2} , \mathcal{U}_{-1} , \mathcal{U}_{0} , \mathcal{U}_{1} , \Delta_{-2}, \Delta_{-1}, \Delta_{0}, \Delta_{1} \text{ at } \tau - \Delta \tau$
					$ \mathcal{U}_{-2} , \mathcal{U}_{-1} , \Delta_{-2}, \Delta_{-1} ext{ at } au$
2 Times	2^{z_0}	2^{z_2}	z_1	z_3	$ \mathcal{U}_{-2} , \mathcal{U}_{-1} , \mathcal{U}_{0} , \mathcal{U}_{1} , \Delta_{-2}, \Delta_{-1}, \Delta_{0}, \Delta_{1} \text{ at } \tau - \Delta \tau$
					$ \mathcal{U}_{-2} , \mathcal{U}_{-1} , \mathcal{U}_0 , \mathcal{U}_1 , \Delta_{-2}, \Delta_{-1}, \Delta_0, \Delta_1 \text{ at } \tau - 2\Delta \tau$

Road map



41/43 41/43

- We have systematically written data-based closures (**probabilistic** and **time-correlated**) for shell models
- Time-correlated closures are working with the present approach
- They may work much better with a better approximation of the densities (other ML tools)
- High-dimensional problems are a significant step in this ladder
- This framework reduced black-box aspects

Thank you!



Data-based approach for time-correlated closur September 25th, 2023

43/43 43/43