



Quantification d'incertitude en mécaniques des fluides : l'apport de la dynamique sous incertitude de position et de l'advection stochastic par transport de Lie

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Baylor Fox-Kemper, Darryl D Holm, Wei Pan, Valentin Resseguier. Quantification d'incertitude en mécaniques des fluides : l'apport de la dynamique sous incertitude de position et de l'advection stochastic par transport de Lie. Colloque National d'Assimilation de Données, Sep 2018, Rennes, France. hal-01891190

HAL Id: hal-01891190

<https://hal.science/hal-01891190>

Submitted on 9 Oct 2018

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Quantification d'incertitude en mécaniques des fluides : l'apport de la dynamique sous incertitude de position et de l'advection stochastic par transport de Lie

Baylor Fox-Kemper

Darryl D. Holm

Wei Pan

Valentin Resseguier

Motivations

- Rigorously identified subgrid dynamics effects
- Injecting likely small-scale dynamics
- Studying bifurcations and attractors
 - Climate projections
- Quantification of modeling errors
 - Ensemble forecasts and data assimilation

Contents

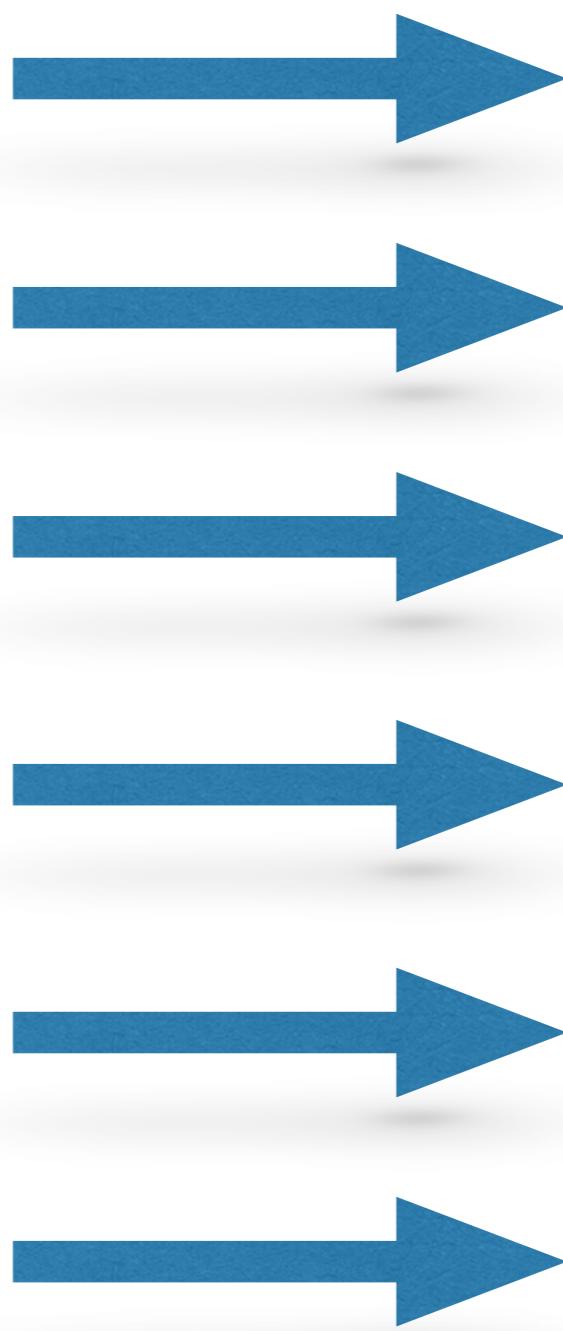
- Stochastic transport
- Stochastic Navier-Stokes : SALT vs LU
- Unresolved velocity parametrisation
- Unresolved velocity non-stationary heterogeneity

Part I

Stochastic transport

Usual random CFD

- Random parameters,
boundary conditions, forcing
- Random initial conditions
- Arbitrary Gaussian forcing
- Averaging, homogenization
- Mori-Zwanzig, MSM
- SPPT, SKEBS + inflation



Other (complementary) issues

Underdispersive
+ need large ensemble

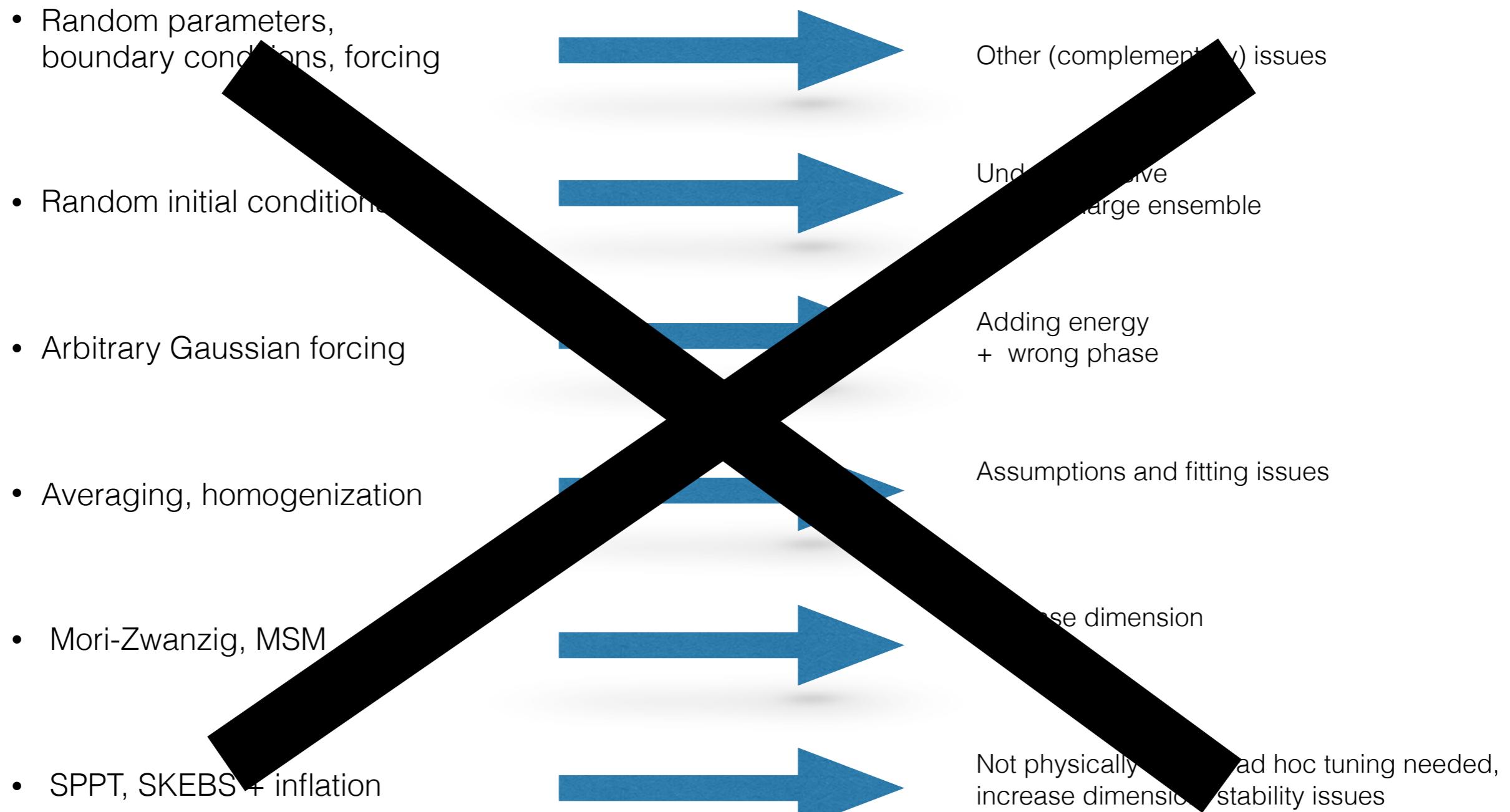
Adding energy
+ wrong phase

Assumptions and fitting issues

Increase dimension

Not physically-based, ad hoc tuning needed,
increase dimension, stability issues

Usual random CFD



SALT & LU :

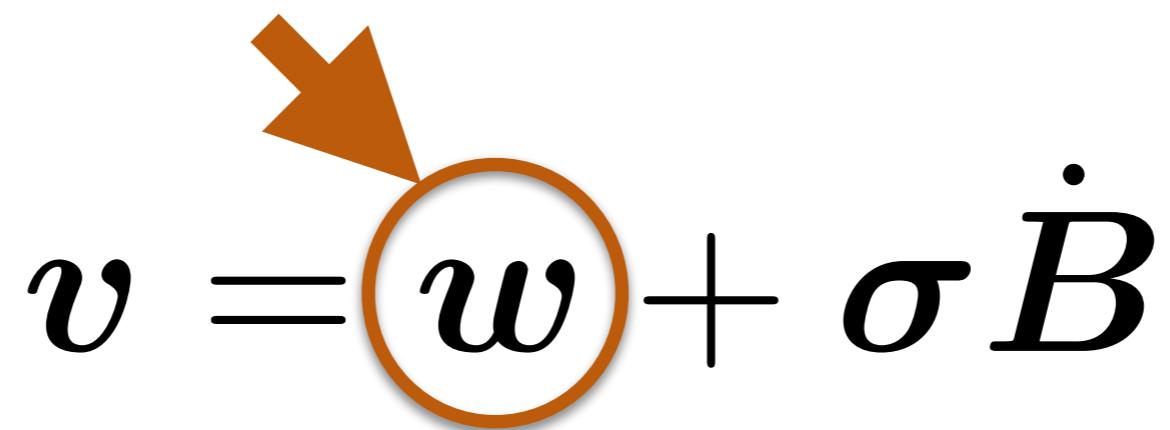
Adding random velocity

$$v = w + \sigma \dot{B}$$

SALT & LU :

Adding random velocity

Resolved
large scales

$$v = w + \sigma \dot{B}$$


SALT & LU:

Adding random velocity

Resolved
large scales

$$v = w + \sigma \dot{B}$$

The diagram illustrates the decomposition of velocity v into two components. On the left, the symbol w is enclosed in an orange circle, with an orange arrow pointing to it from the text "Resolved large scales". On the right, the term $\sigma \dot{B}$ is enclosed in a purple circle, with a purple arrow pointing to it from the text "White-in-time small scales".

Large scales:

$$w$$

Small scales:

$$\sigma \dot{B}$$

Variance tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

SALT & LU :

Adding random velocity

Resolved
large scales

White-in-time
small scales

$$v = w + \sigma \dot{B}$$

The diagram illustrates the decomposition of velocity v into two components. On the left, the symbol w is enclosed in an orange circle with an orange arrow pointing to it from above, labeled "Resolved large scales". On the right, the symbol $\sigma \dot{B}$ is enclosed in a purple circle with a purple arrow pointing to it from above, labeled "White-in-time small scales". A plus sign is placed between the two circles, indicating their sum.

Large scales:

$$\boldsymbol{w}$$

Small scales:

$$\sigma \dot{\boldsymbol{B}}$$

Variance
tensor:

$$\boldsymbol{a} = \boldsymbol{a}(\boldsymbol{x}, \boldsymbol{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\boldsymbol{B} (\boldsymbol{\sigma} d\boldsymbol{B})^T\}}{dt}$$

SALT & LU: Adding random velocity

$$d\boldsymbol{X}_t = \boldsymbol{w}^*(\boldsymbol{X}_t, t)dt + \boldsymbol{\sigma}(\boldsymbol{X}_t, t) \circ dB_t$$

$$= \boldsymbol{w}(\boldsymbol{X}_t, t)dt + \boldsymbol{\sigma}(\boldsymbol{X}_t, t)dB_t$$

Large scales:

$$\boldsymbol{w}$$

Small scales:

$$\sigma \dot{\boldsymbol{B}}$$

Variance
tensor:

$$\boldsymbol{a} = \boldsymbol{a}(\boldsymbol{x}, \boldsymbol{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\boldsymbol{B} (\boldsymbol{\sigma} d\boldsymbol{B})^T\}}{dt}$$

SALT & LU: Adding random velocity

Stratonovich

$$d\boldsymbol{X}_t = \boldsymbol{w}^*(\boldsymbol{X}_t, t)dt + \boldsymbol{\sigma}(\boldsymbol{X}_t, t) \circ d\boldsymbol{B}_t$$

$$= \boldsymbol{w}(\boldsymbol{X}_t, t)dt + \boldsymbol{\sigma}(\boldsymbol{X}_t, t)d\boldsymbol{B}_t$$

Large scales:

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Small scales:

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SALT & LU: Adding random velocity

Stratonovich

$$d\boldsymbol{X}_t = \boldsymbol{w}^*(\boldsymbol{X}_t, t)dt + \boldsymbol{\sigma}(\boldsymbol{X}_t, t) \circ d\boldsymbol{B}_t$$

Ito

$$= \boldsymbol{w}(\boldsymbol{X}_t, t)dt + \boldsymbol{\sigma}(\boldsymbol{X}_t, t)d\boldsymbol{B}_t$$

Large scales:

w

Small scales:

$\sigma \dot{B}$

Variance
tensor:

$$\mathbf{a} = \mathbf{a}(\mathbf{x}, \mathbf{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

SALT & LU: Adding random velocity

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$$= \mathbf{w}(\mathbf{X}_t, t)dt + \boldsymbol{\sigma}(\mathbf{X}_t, t)d\mathbf{B}_t$$

Ito

References : Mikulevicius and Rozovskii, 2004 Flandoli, 2011

Memin, 2014
Resseguier et al. 2017 a, b, c
Chapron et al. 2017
Cai et al. 2017

Holm, 2015
Holm and Tyranowski, 2016
Arnaudon et al., 2017

Cotter and al 2017
Crisan et al., 2017
Gay-Balmaz & Holm 2017
Cotter and al 2018 a, b

Large scales:

w

Small scales:

$\sigma \dot{B}$

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Flandoli, 2011

LU

- Memin**, 2014
Resseguier et al. 2017 a, b, c
Chapron et al. 2017
Cai et al. 2017

SALT

- Holm**, 2015
Holm and Tyranowski, 2016
Arnaudon et al., 2017
- Cotter and al 2017
Crisan et al., 2017
Gay-Balmaz & Holm 2017
Cotter and al 2018 a, b

Large scales:

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Small scales:

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Advection of tracer Θ

$$\frac{D\Theta}{Dt} = 0$$

Large scales:

$$\boldsymbol{w}$$

Small scales:

$$\sigma \dot{\boldsymbol{B}}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(\mathbf{x}, \mathbf{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

Advection of tracer Θ

Ito-Wentzell
formula
(Kunita 1990)

$$\frac{D\Theta}{Dt} = 0$$

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$$\boldsymbol{w}$$

Small scales:

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Stratonovich notations:

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Small scales:

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Advection of tracer Θ

Ito-Wentzell formula
(Kunita 1990)

Stratonovich notations:

$$\partial_t \Theta + (w^* + \sigma \circ \dot{\boldsymbol{B}}) \cdot \nabla \Theta = 0$$

Large scales:

$$\boldsymbol{w}$$

Small scales:

$$\sigma \dot{\boldsymbol{B}}$$

Variance tensor:

$$\boldsymbol{a} = \boldsymbol{a}(\boldsymbol{x}, \boldsymbol{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\boldsymbol{B} (\boldsymbol{\sigma} d\boldsymbol{B})^T\}}{dt}$$

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Ito notations:

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Small scales:

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Advection of tracer Θ

Ito-Wentzell formula
(Kunita 1990)

Ito notations:

$$\partial_t \Theta + w^* \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta = \nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla \Theta \right)$$

Large scales:

w

Small scales:

$\sigma \dot{B}$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} dB (\boldsymbol{\sigma} dB)^T\}}{dt}$$

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Ito-Wentzell formula
(Kunita 1990)

Ito notations:

Advection

$$\partial_t \Theta + \mathbf{w}^* \cdot \nabla \Theta + \sigma \dot{\mathbf{B}} \cdot \nabla \Theta =$$

Diffusion

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w

Small scales:

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Variance tensor:

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$$\nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla \Theta \right)$$

Stratonovich drift : « Drift correction »
in Ito notations

Large scales:

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Small scales:

$$\sigma \dot{\mathbf{B}}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\sigma d\mathbf{B} (\sigma d\mathbf{B})^T\}}{dt}$$

Advection of tracer Θ

Ito-Wentzell formula
(Kunita 1990)

Multiplicative random forcing

Ito notations:

Advection

$$\partial_t \Theta + \mathbf{w}^\star \cdot \nabla \Theta + \sigma \dot{\mathbf{B}} \cdot \nabla \Theta =$$

Diffusion

$$\nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla \Theta \right)$$

Stratonovich drift : « Drift correction »
in Ito notations

Large scales:

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Small scales:

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Variance tensor:

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Ito-Wentzell formula
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Diffusion

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Ito-Wentzell formula
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Diffusion

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in Ito notations

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 w

Small scales:
 $\sigma \dot{B}$

Variance
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

Advection of tracer Θ

Ito-Wentzell
formula
(Kunita 1990)

Multiplicative
random
forcing

Balanced
energy
exchanges

Ito notations:

$$\partial_t \Theta + w^* \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta =$$

Advection

Diffusion

$$\nabla \cdot \left(\frac{1}{2} a \nabla \Theta \right)$$

Stratonovich drift : « Drift correction »
in Ito notations

Large scales:

$$\mathbf{w}$$

Small scales:

$$\sigma \dot{\mathbf{B}}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

Moments of a passive tracer Θ

$$\Theta(\bullet, t') \perp \mathbf{w}(\bullet, t), \forall t' \leq t$$

Advection

$$\partial_t \mathbb{E}(\Theta) + \mathbf{w}^* \cdot \nabla \mathbb{E}(\Theta) =$$

Diffusion

$$\nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla \mathbb{E}(\Theta) \right)$$

$$\partial_t Var(\Theta) + \mathbf{w}^* \cdot \nabla Var(\Theta) = \nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla Var(\Theta) \right) + (\nabla \mathbb{E}(\Theta))^T \mathbf{a} \nabla \mathbb{E}(\Theta)$$

Large scales:

$$\mathbf{w}$$

Small scales:

$$\sigma \dot{\mathbf{B}}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

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Diffusion

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Conversion from deterministic to random energy

Source term

Large scales:

$$\mathbf{w}$$

Small scales:

$$\sigma \dot{\mathbf{B}}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

Moments of a passive tracer Θ

$$\Theta(\bullet, t') \perp \mathbf{w}(\bullet, t), \forall t' \leq t$$

Conservation
of the energy

$$\frac{d}{dt} \int_{\Omega} \Theta^2 = 0$$

Advection

$$\partial_t \mathbb{E}(\Theta) + \mathbf{w}^* \cdot \nabla \mathbb{E}(\Theta) =$$

Diffusion

$$\nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla \mathbb{E}(\Theta) \right)$$
$$\partial_t Var(\Theta) + \mathbf{w}^* \cdot \nabla Var(\Theta) = \nabla \cdot \left(\frac{1}{2} \mathbf{a} \nabla Var(\Theta) \right) + (\nabla \mathbb{E}(\Theta))^T \mathbf{a} \nabla \mathbb{E}(\Theta)$$

Conversion from
deterministic to
random energy

Source term

Part II

Stochastic Navier-Stokes

SALT vs LU

Comparison

	LU	SALT
Scalar (e.g. SQG)	Identical	
Navier-Stokes	2 differences	
Vorticity	2 differences	
Kinetic energy conservation	✓	✗
Helicity (& 2D enstrophy) conservation	✗	✓

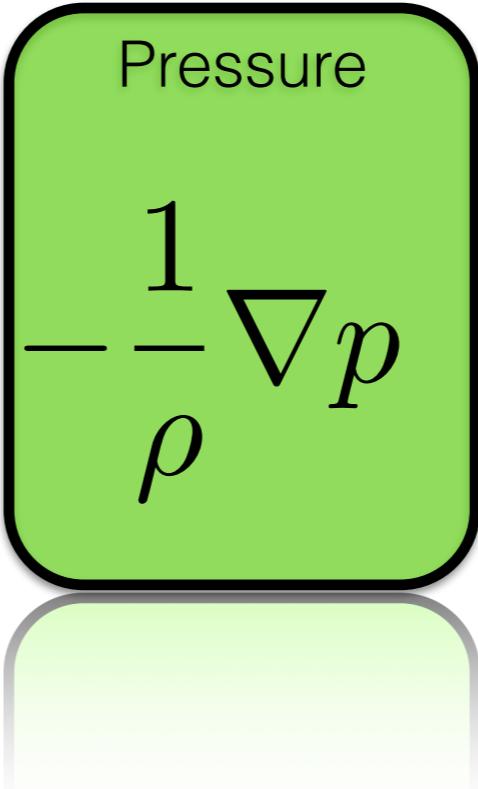
Navier-Stokes LU

$$\frac{Dw}{Dt} = -\frac{1}{\rho} \nabla p$$

Navier-Stokes LU

$$\frac{Dw}{Dt} = -\frac{1}{\rho} \nabla p$$

Pressure



Navier-Stokes LU

Stochastic
transport
of Ito drift

$$\frac{Dw}{Dt} = -\frac{1}{\rho} \nabla p$$

Pressure

Navier-Stokes LU

Stochastic
transport
of Ito drift

$$\frac{Dw}{Dt} = -\frac{1}{\rho} \nabla p$$

Pressure



Conserve (Ito) kinetic energy

Navier-Stokes SALT

$$\frac{Dw^*}{Dt} + \nabla(\sigma \circ \dot{B})^T w^* = -\frac{1}{\rho} \nabla p$$

Navier-Stokes SALT

$$\frac{Dw^*}{Dt} + \nabla(\sigma \circ \dot{B})^T w^* = -\frac{1}{\rho} \nabla p$$

Pressure

Navier-Stokes SALT

Stochastic
transport of
Stratonovich drift

$$\frac{Dw^*}{Dt}$$

$$+ \nabla(\sigma \circ \dot{B})^T w^* =$$

Pressure

$$-\frac{1}{\rho} \nabla p$$

Navier-Stokes SALT

Stochastic
transport of
Stratonovich drift

$$\frac{Dw^*}{Dt}$$

Additional term

$$+ \nabla(\sigma \circ B)^T w^*$$

Pressure

$$-\frac{1}{\rho} \nabla p$$



Conserve (Stratonovich) helicity

A word about reduced order models LU

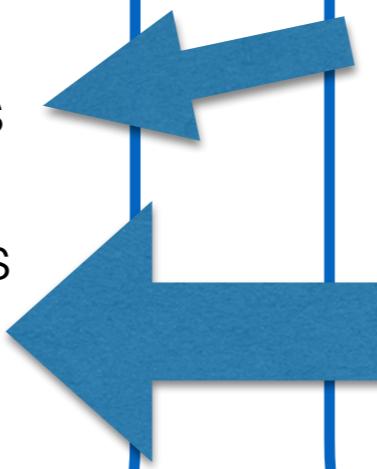
- Very fast simulation of very complex system (e.g. for industrial application)
- Physical model (PDE) simplified using observations

Forecast of reduced models LU

- Rigorous and low-cost estimators
- Stabilization of the unstable modes
- Maintain variability of stable modes
- Uncertainty quantification

State of art

- Possible parametrization with eddy viscosity
- Impossible to parametrize with additional dissipation
Need ad hoc closure like MQG
(Sapsis and Majda, 2013a,b,c)



Part III

Unresolved velocity parametrisation

Code available online

Large scales:

w

Small scales:

$\sigma \dot{B}$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(\mathbf{x}, \mathbf{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

Parameter-free models for $\boldsymbol{\sigma}$

Cotter et al.
2018b

Resseguier et al.
2017b

Applicable to :

Homogeneous

LU & SALT

LU & SALT

No

Yes

Stationary

Yes

Not anymore

Self-similar assumption

No

Yes

Data-driven

Yes

No

Large scales:

$$\boldsymbol{w}$$

Small scales:

$$\sigma \dot{\boldsymbol{B}}$$

Variance
tensor:

$$\boldsymbol{a} = \boldsymbol{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\boldsymbol{B} (\boldsymbol{\sigma} d\boldsymbol{B})^T\}}{dt}$$

Cotter et al. 2018b: Karhunen–Loève decomposition (EOF)

$$\sigma dB_t = \sum_{i=1}^n \xi_i(x) dW_i(t)$$

Large scales:

w

Small scales:

$\sigma \dot{B}$

Variance
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (dB)^T\}}{dt}$$

Cotter et al. 2018b: Karhunen–Loève decomposition (EOF)

$$\sigma dB_t = \sum_{i=1}^n \xi_i(x) dW_i(t)$$

iid
Brownian
motion

Large scales:

w

Small scales:

$\sigma \dot{B}$

Variance tensor:

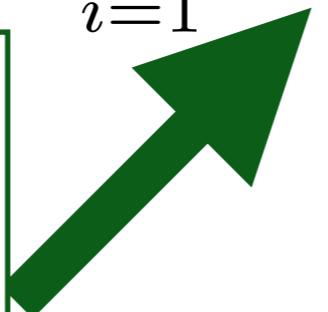
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Cotter et al. 2018b: Karhunen–Loève decomposition (EOF)

$$\sigma dB_t = \sum_{i=1}^n \xi_i(x) dW_i(t)$$

iid
Brownian motion

weighted
EOFs learned
on data



Large scales:

w

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Cotter et al. 2018b: Karhunen–Loève decomposition (EOF)

$$\sigma dB_t = \sum_{i=1}^n \xi_i(x) dW_i(t)$$

iid
Brownian motion

weighted
EOFs learned
on data

PCA on : $\Delta X_k(x) = X_{t_k, t_k + \Delta t}^{HR}(x) - X_{t_k, t_k + \Delta t}^{LR}(x)$

Large scales:
 w

Small scales:
 $\sigma \dot{B}$

Variance
tensor:
 $a = a(x, x) =$

$$\frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

Cotter et al. 2018b:

Karhunen–Loève decomposition (EOF)

$$\sigma dB_t = \sum_{i=1}^n \xi_i(x) dW_i(t)$$

iid
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PCA on : $\Delta X_k(x) = X_{t_k, t_k + \Delta t}^{HR}(x) - X_{t_k, t_k + \Delta t}^{LR}(x)$

Flow from **high-**
resolution velocity

Flow from **low-**
resolution velocity

Large scales:

$$w$$

Small scales:

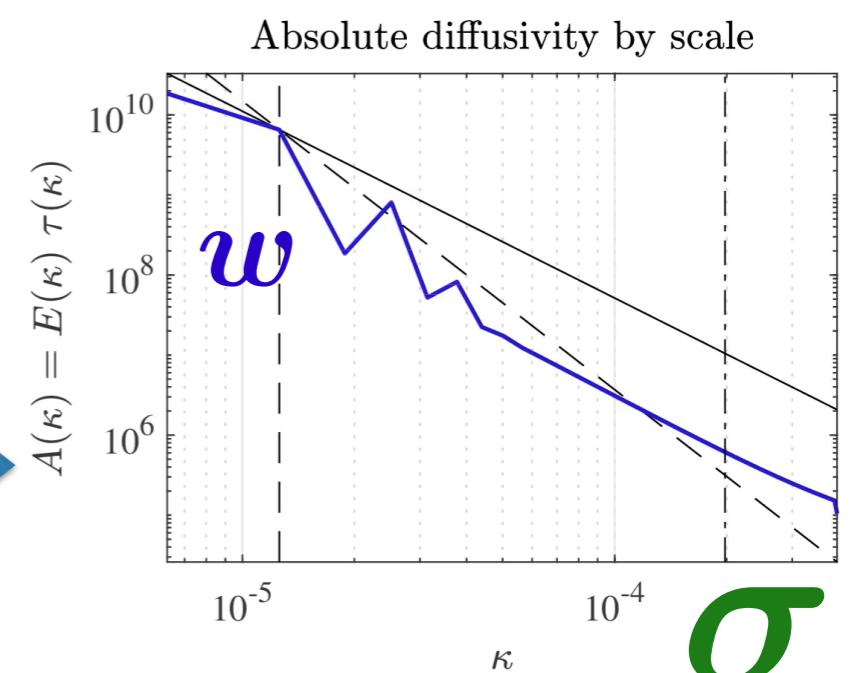
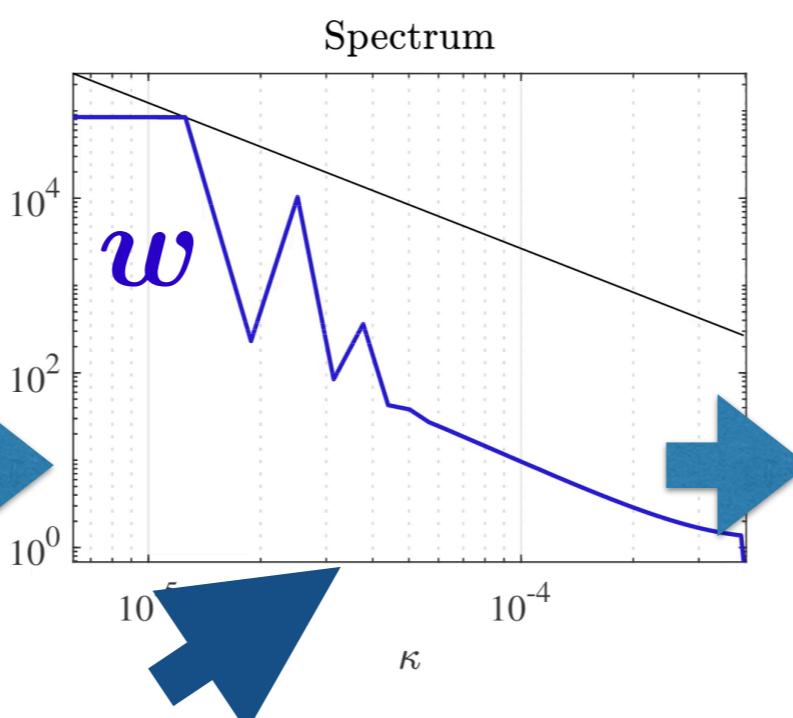
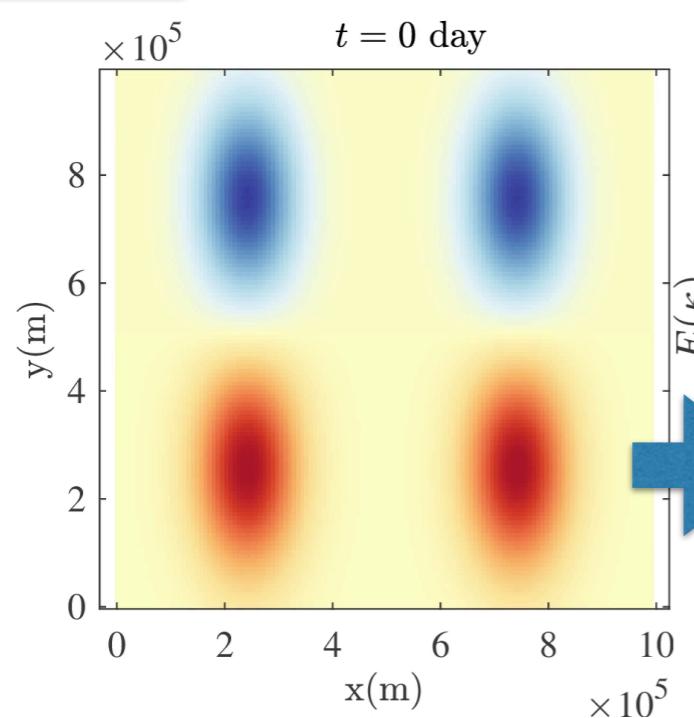
$$\sigma \dot{B}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} dB (\boldsymbol{\sigma} dB)^T\}}{dt}$$

(improvement of)

Resseguier et al. 2017b: Self-similar model



Kinetic energy
Spectrum

$$E(\kappa) = \frac{1}{\mu(\Omega)} \mathbb{E} \oint_{\|k\|=\kappa} d\theta_k \kappa \|\hat{v}(k)\|^2$$

Large scales:

$$w$$

Small scales:

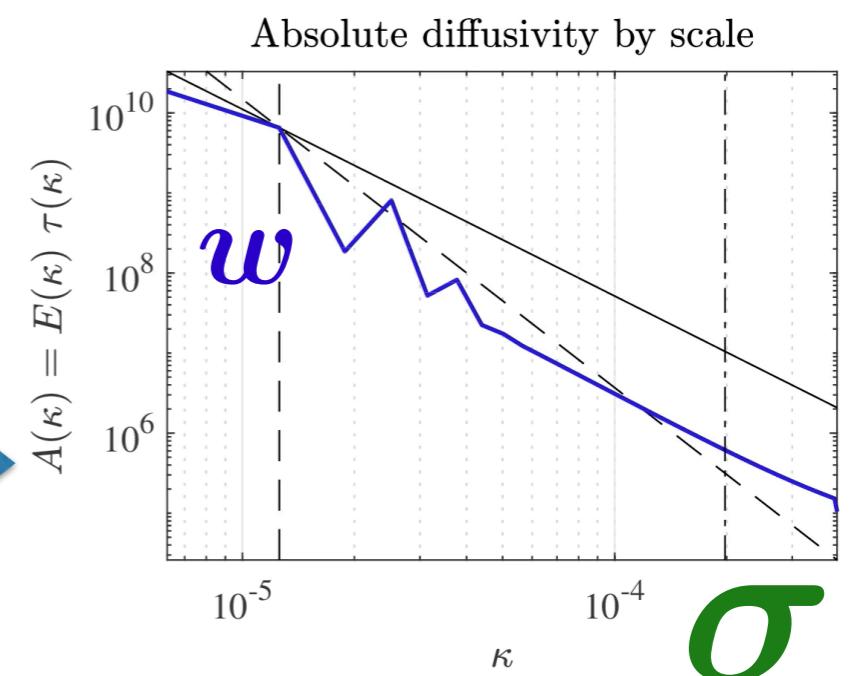
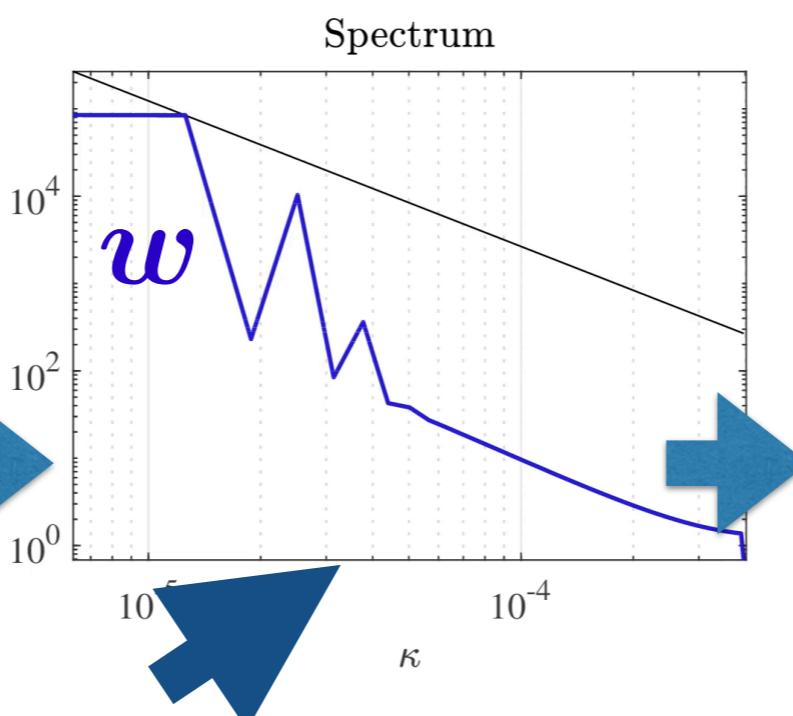
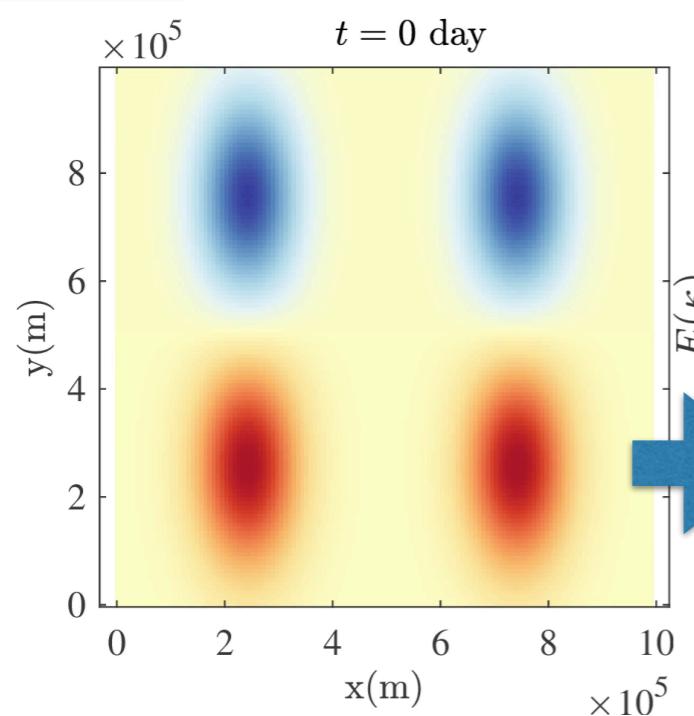
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Large scales:

w

Small scales:

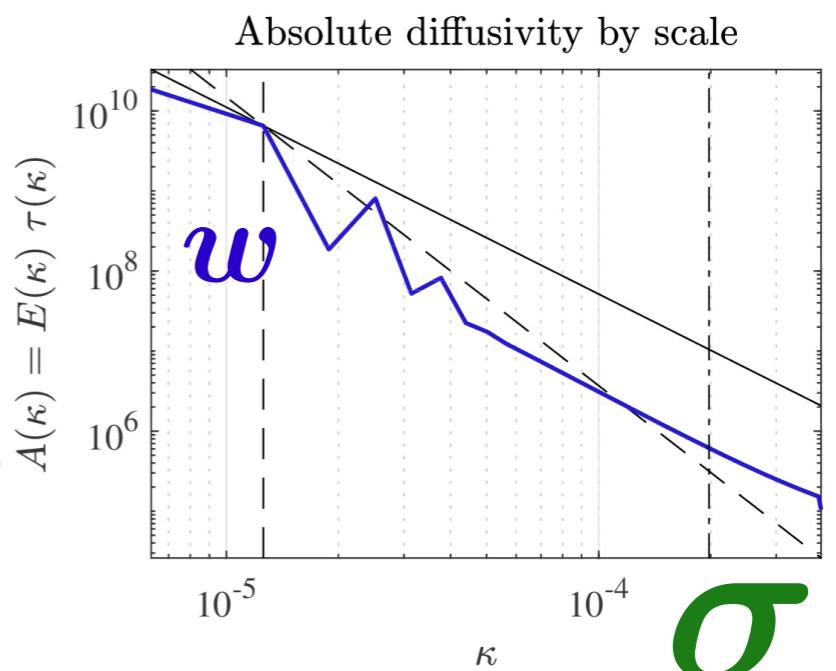
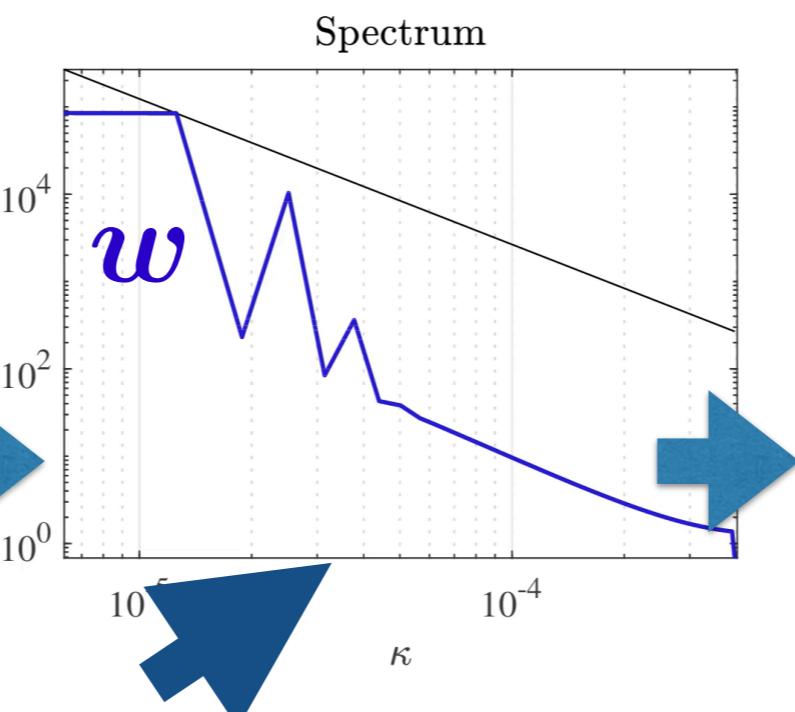
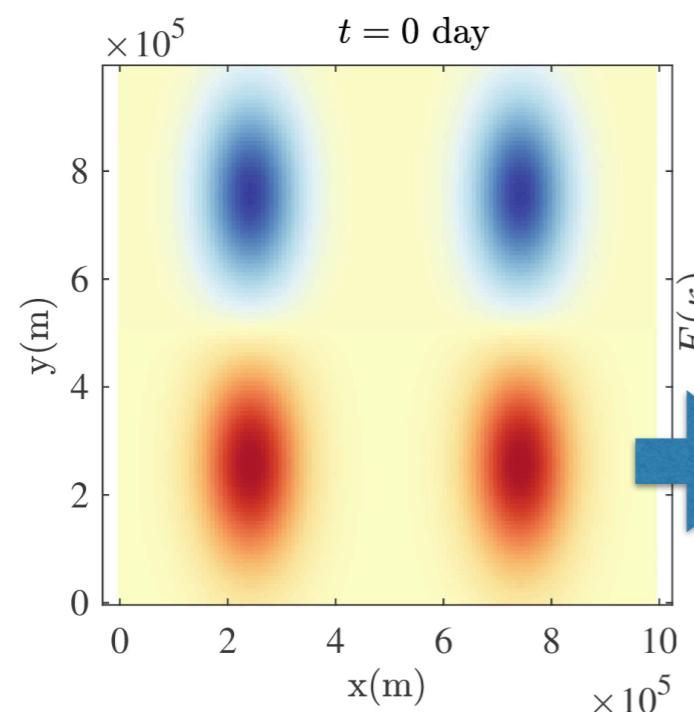
$\sigma \dot{B}$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(\mathbf{x}, \mathbf{x}) = \frac{\mathbb{E}\{\boldsymbol{\sigma} d\mathbf{B} (\boldsymbol{\sigma} d\mathbf{B})^T\}}{dt}$$

Absolute Diffusivity Spectral Density

$$A(\kappa) = E(\kappa)\tau(\kappa) = \kappa^{-3/2} E^{1/2}(\kappa)$$



Kinetic energy Spectrum

$$E(\kappa) = \frac{1}{\mu(\Omega)} \mathbb{E} \oint_{\|k\|=\kappa} d\theta_k \kappa \|\hat{v}(k)\|^2$$

SQG

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

$$- \frac{1}{\tau} (b - F) \quad \text{Forcing}$$

$$u = \left(\text{cst.} \nabla^\perp \Delta^{-1/2} \right) b$$

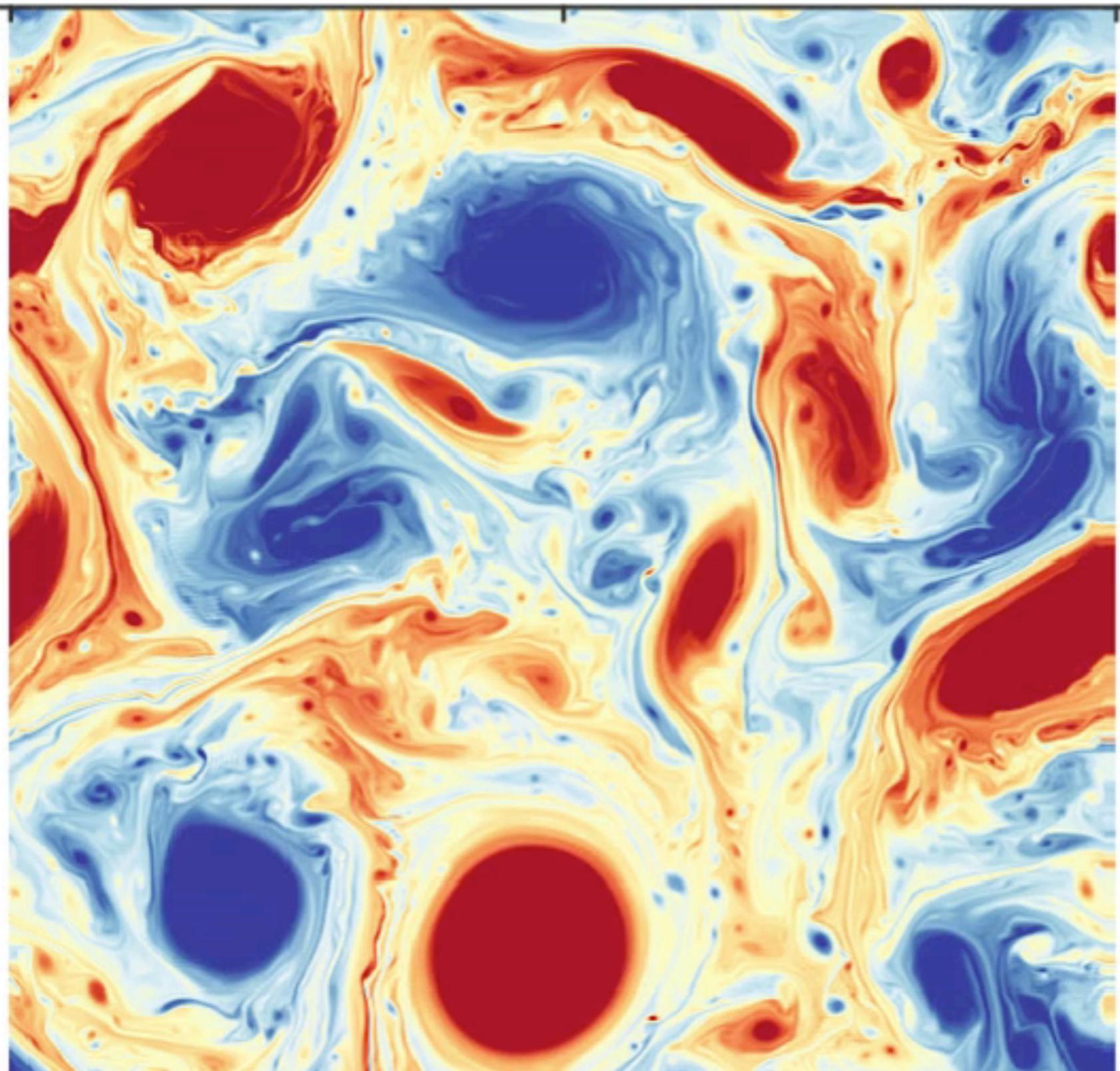
Reference flow:

deterministic

SQG

512 x 512

$t = 100$ day



SQG

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

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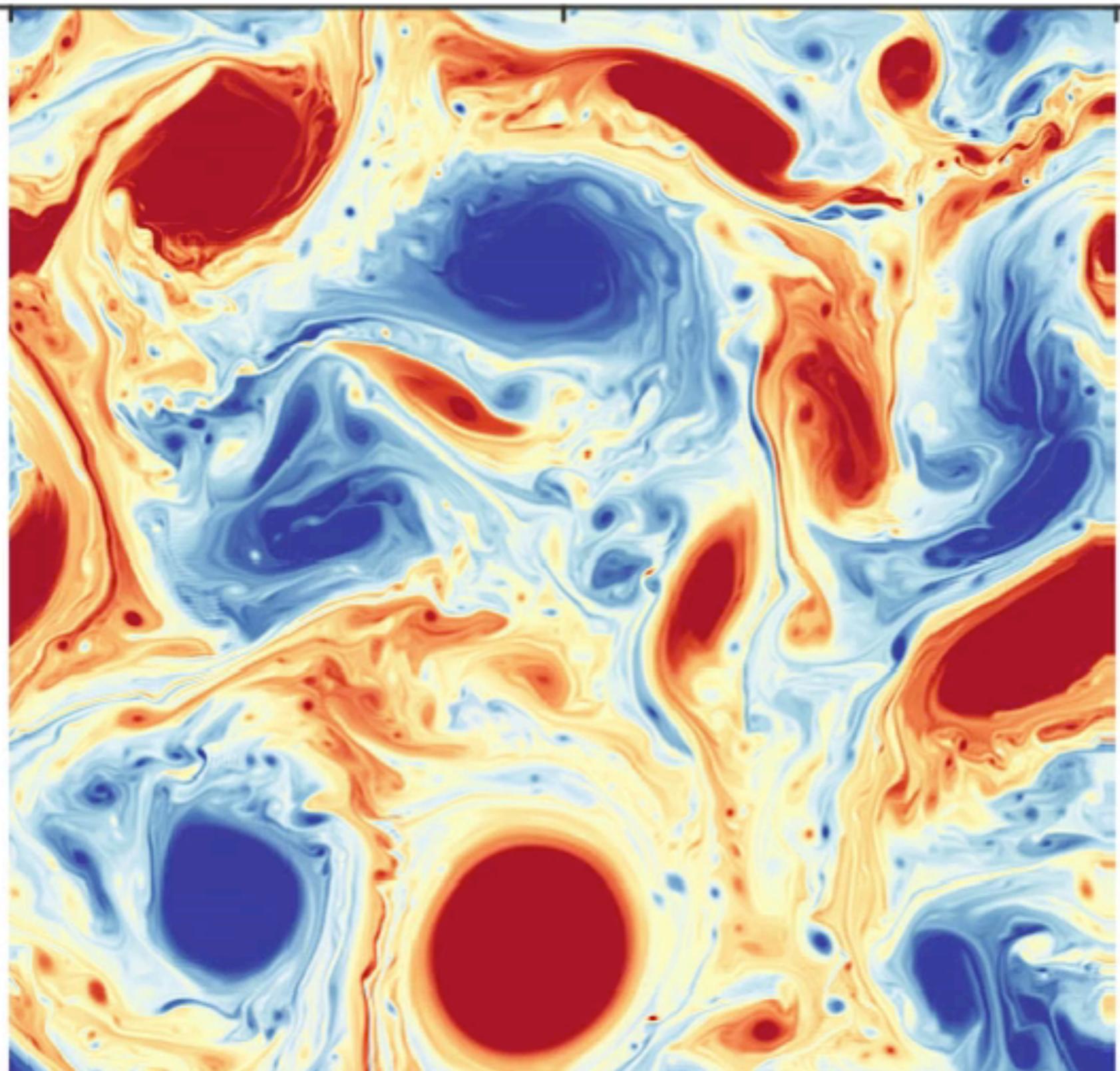
Reference flow:

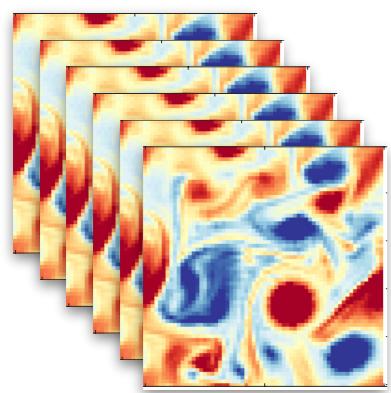
deterministic

SQG

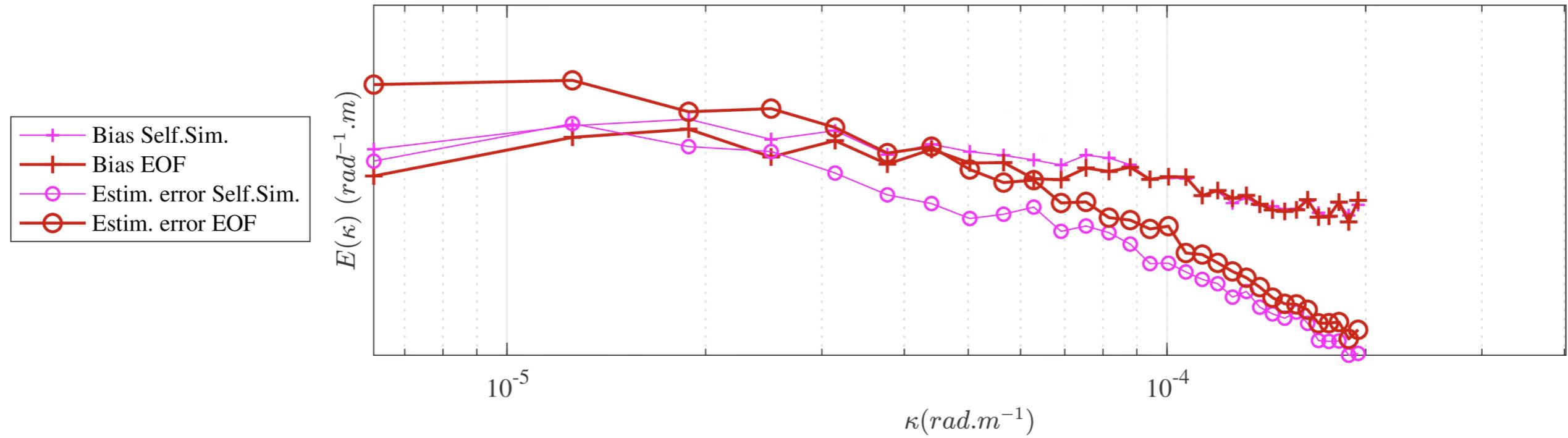
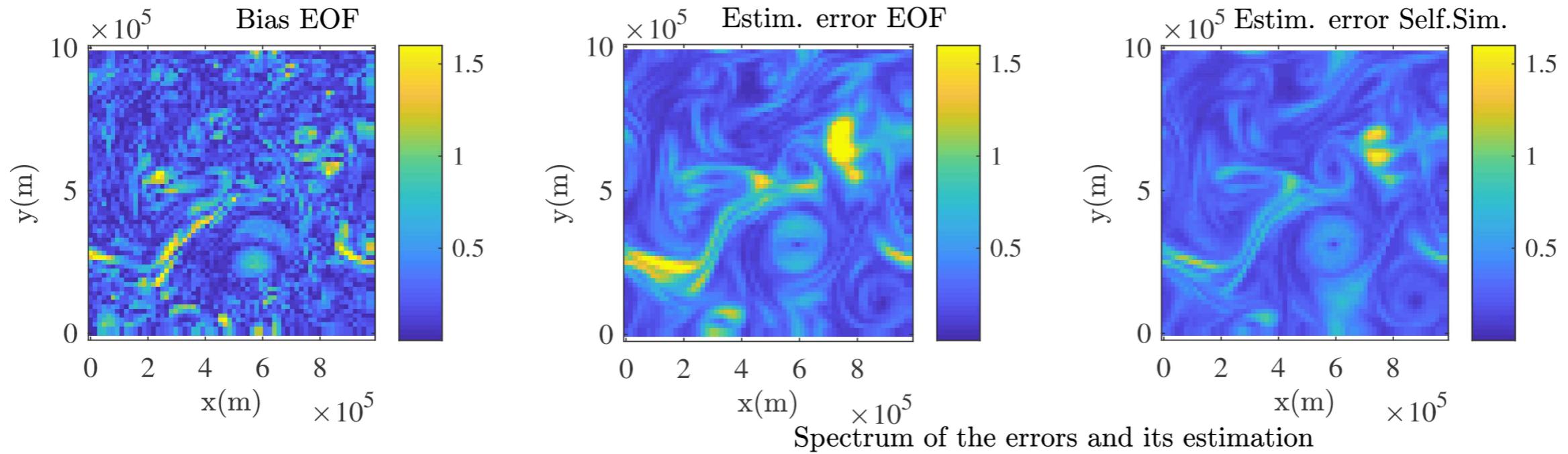
512 x 512

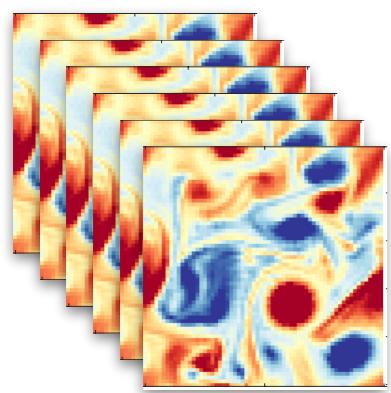
$t = 100$ day



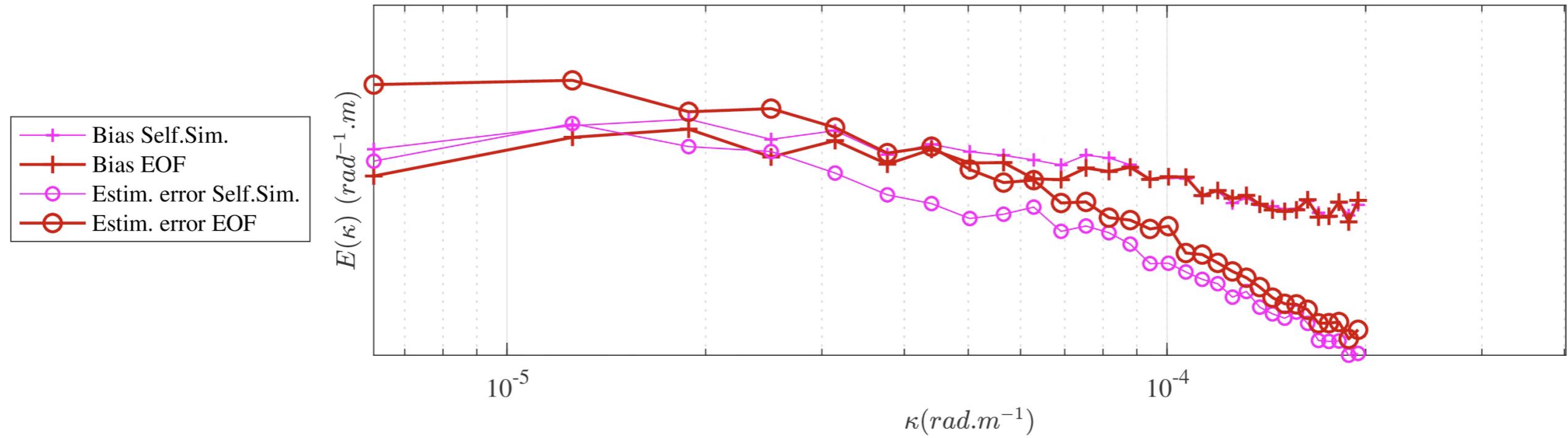
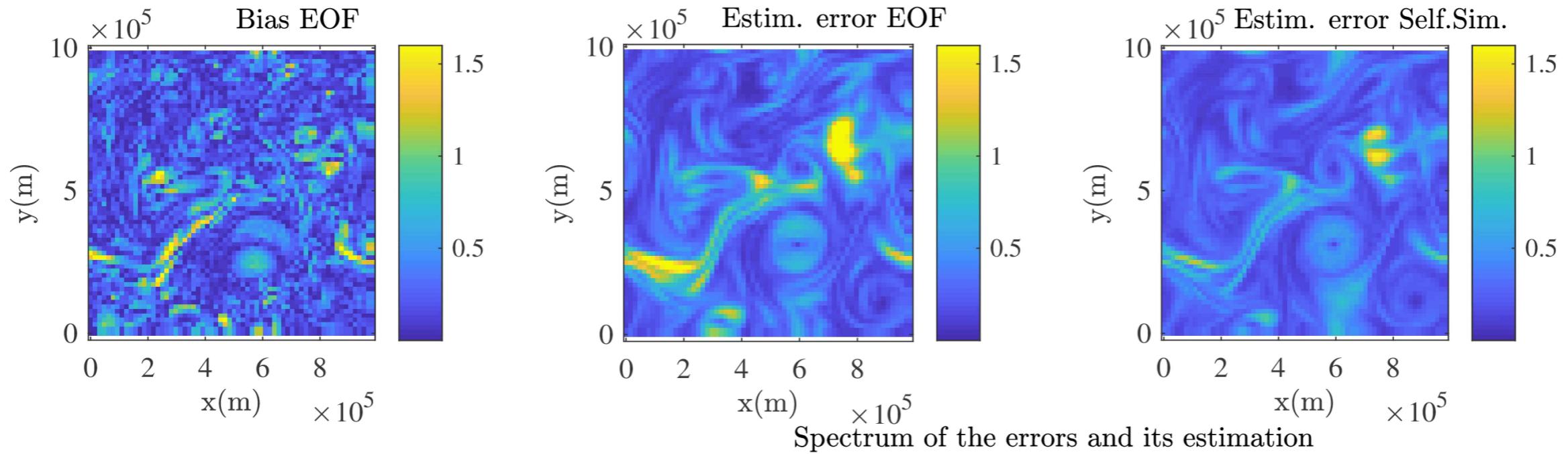


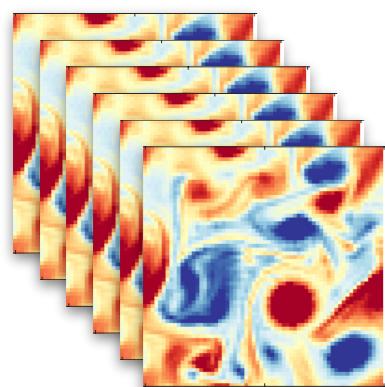
Ensemble : uncertainty quantification





Ensemble : uncertainty quantification

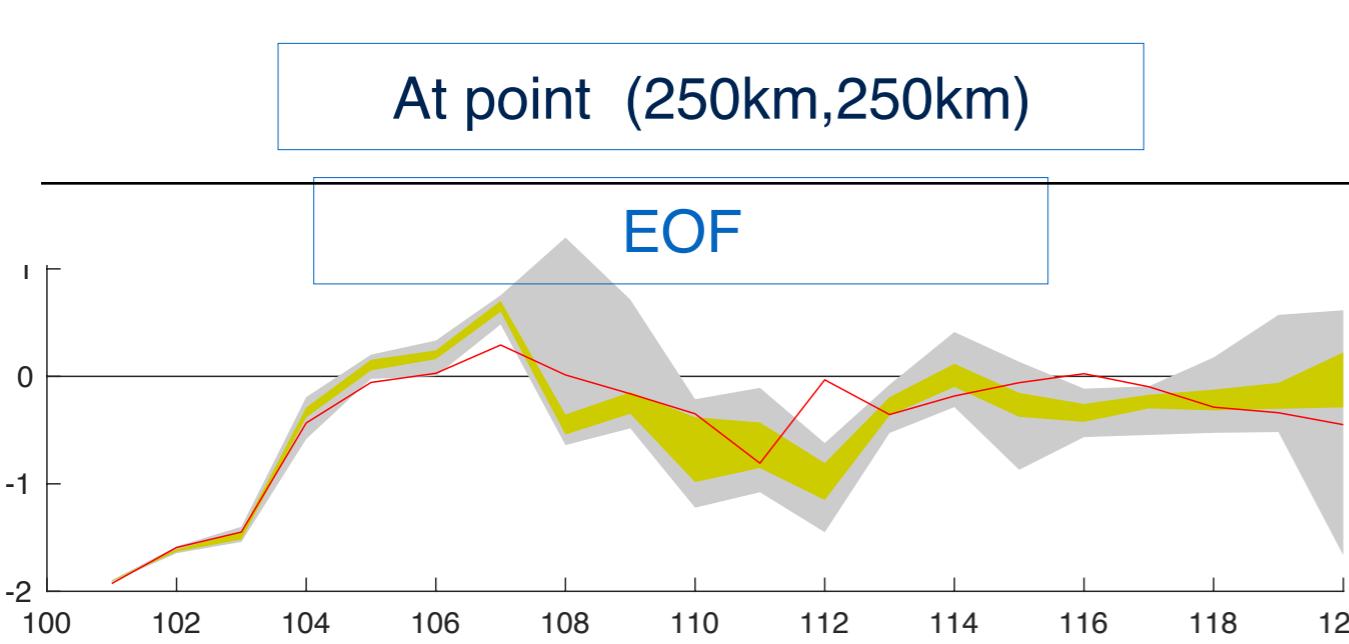




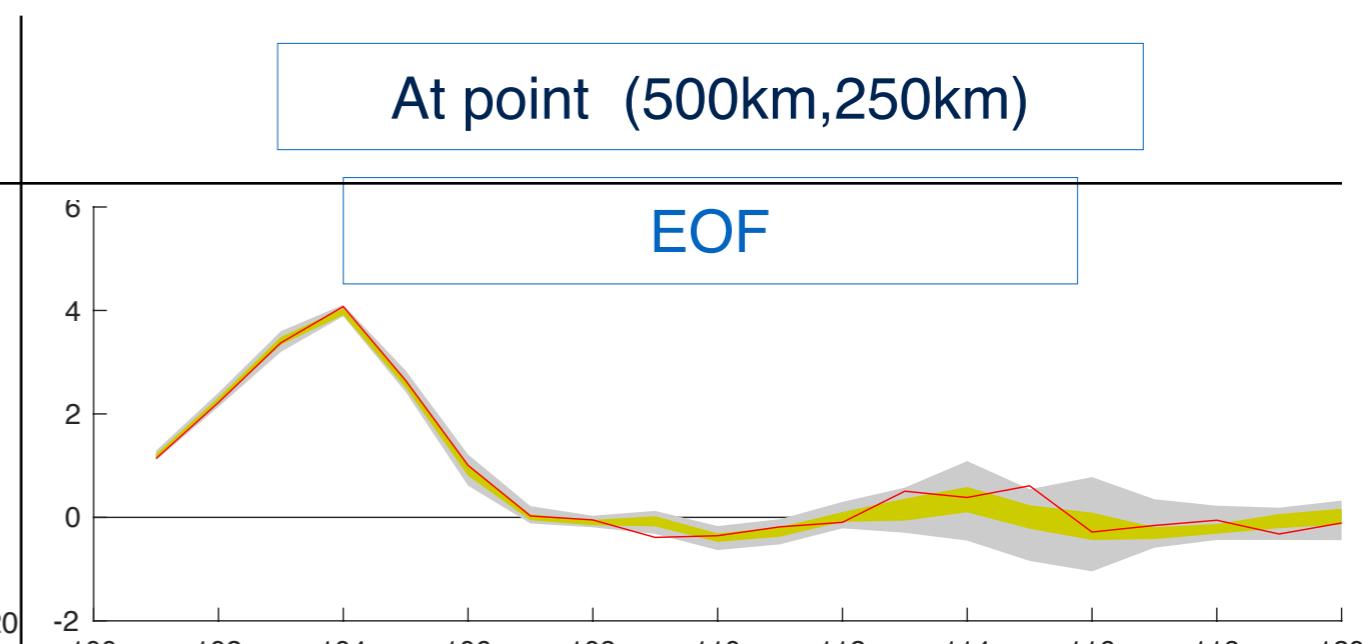
20
realisations

Ensemble : uncertainty quantification

At point (250km,250km)



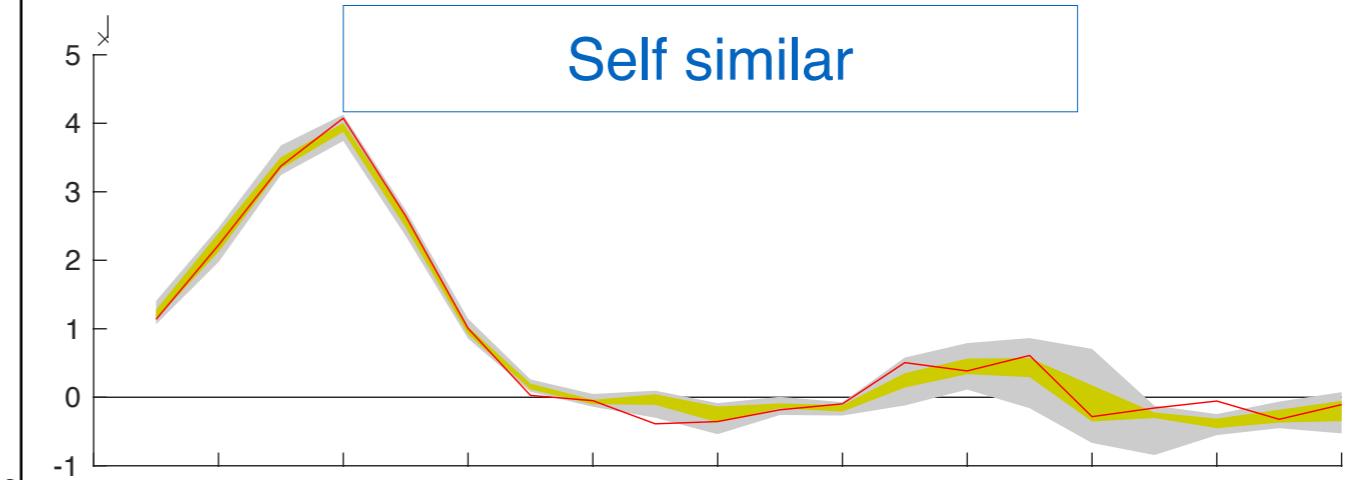
At point (500km,250km)

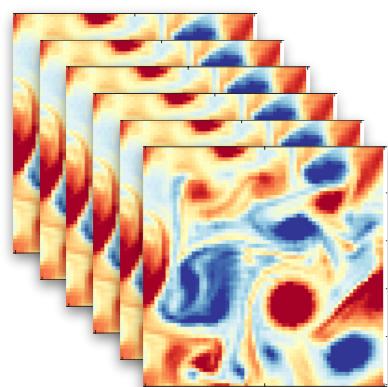


Self similar



Self similar





20
realisations

Ensemble : uncertainty quantification

At point (250km,500km)

EOF

Self similar

At point (500km,500km)

EOF

Self similar

Part IV

Unresolved velocity non-stationary heterogeneity

$t = 0$ day

SQG

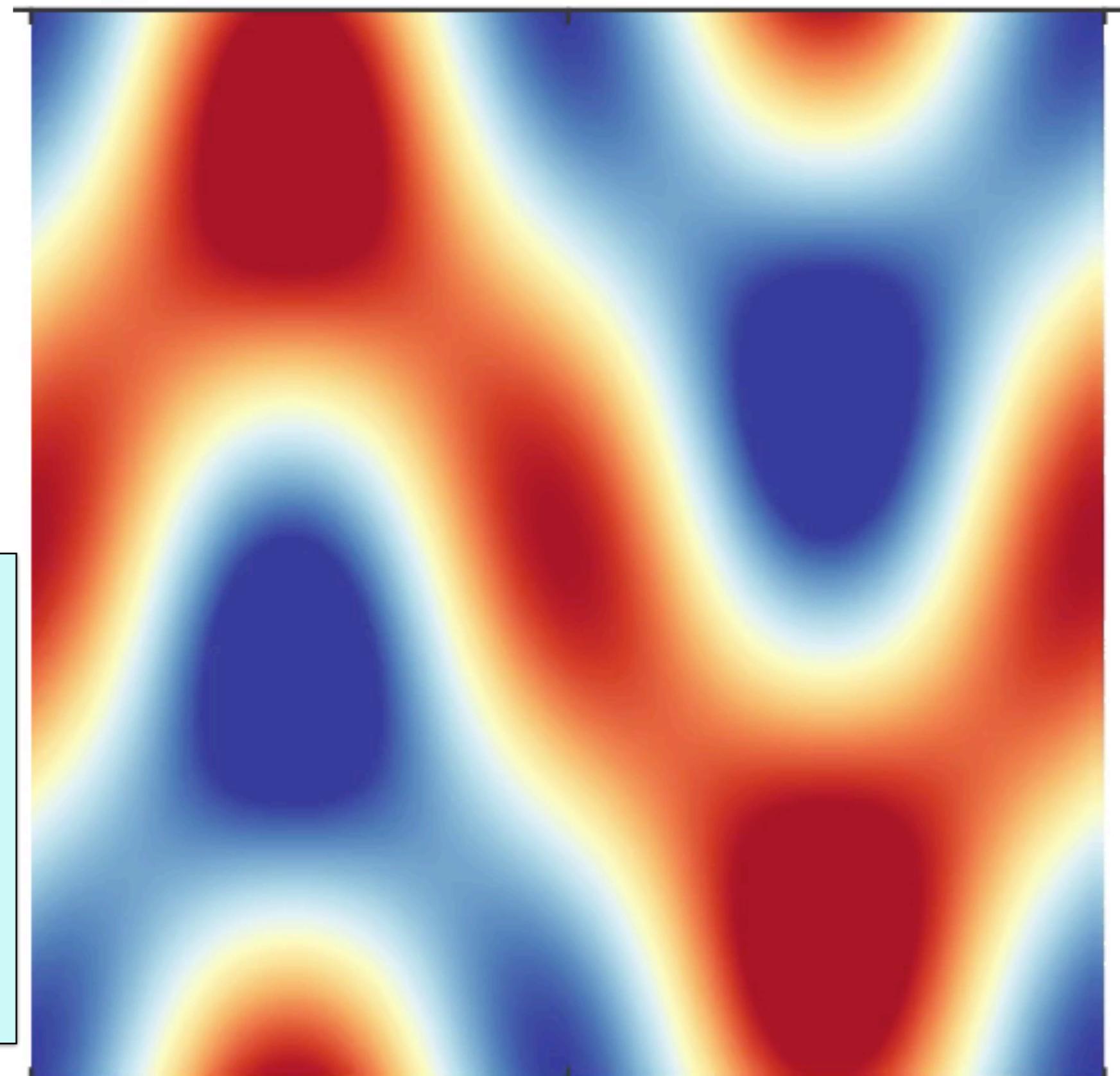
$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

$$\mathbf{u} = \left(\text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic SQG
1024 x 1024

Constantin et al.
(1994, 1999, 2012)



$t = 0$ day

SQG

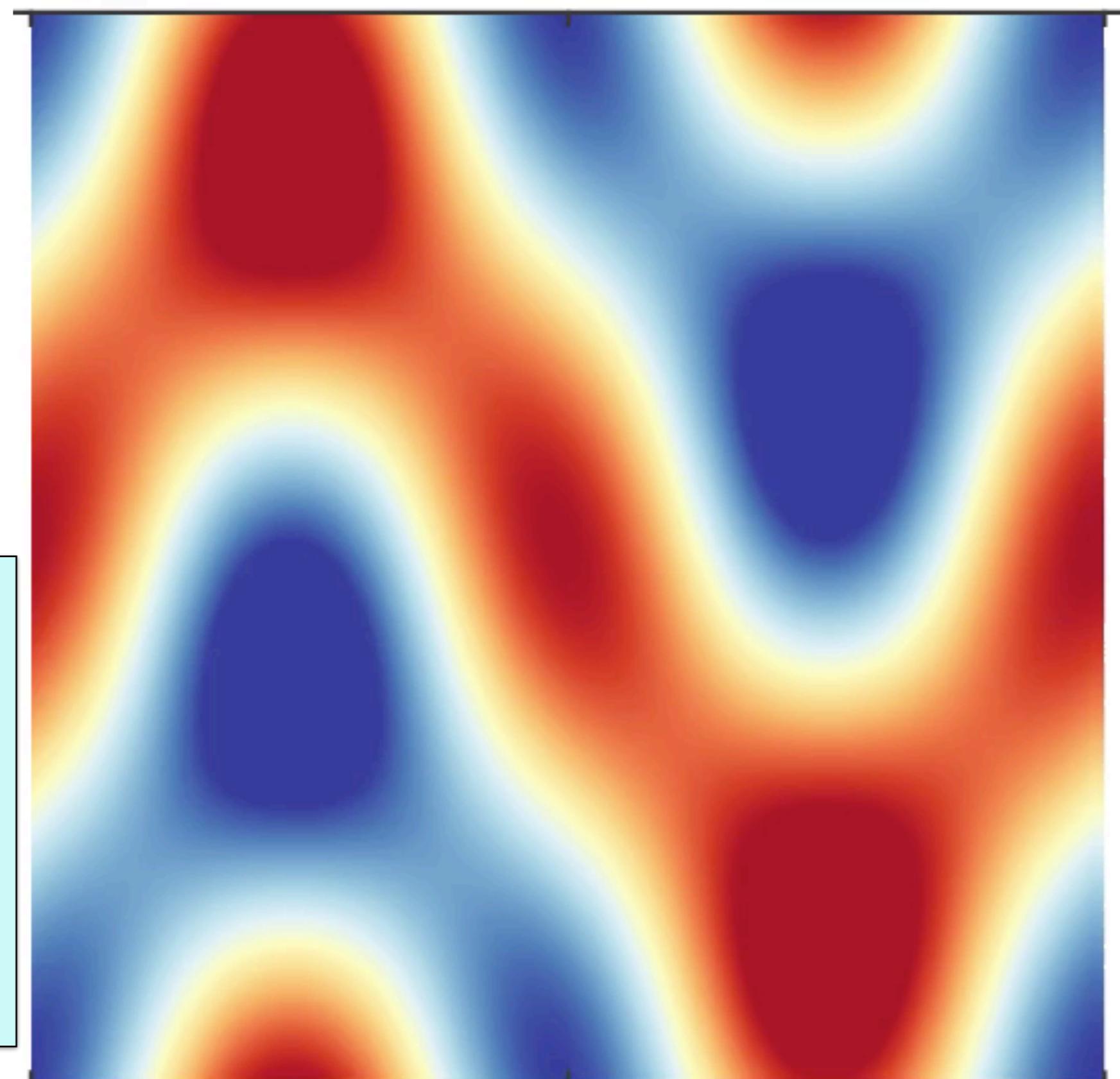
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Constantin et al.
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$t = 15$ day

SQG

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

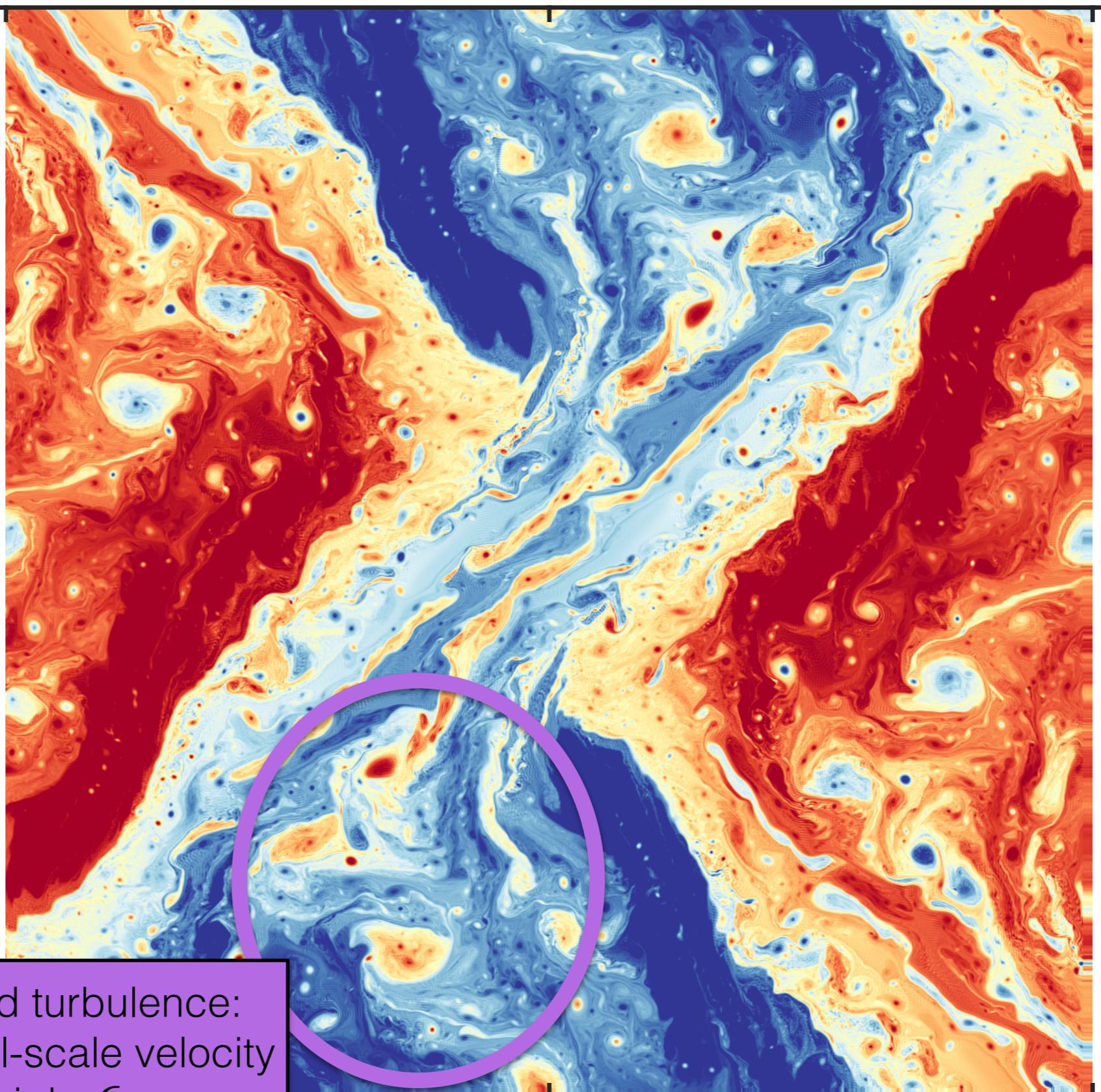
$$\mathbf{u} = \left(\text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic SQG
1024 x 1024

Constantin et al.
(1994, 1995)

Developed turbulence:
strong small-scale velocity
high ϵ



$t = 15$ day

SQG

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

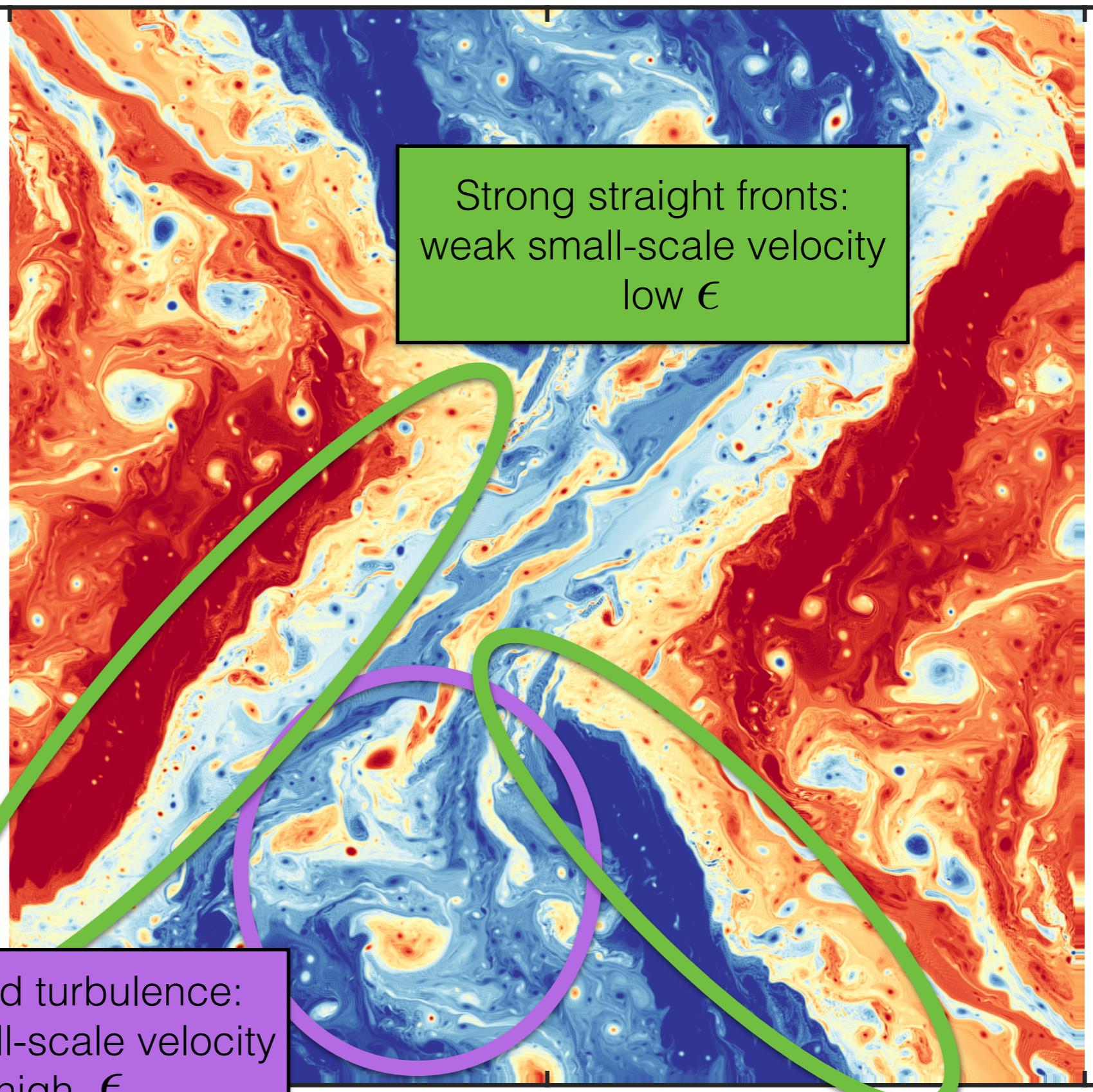
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Constantin et al.
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Developed turbulence:
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Large scales:

$$w$$

Small scales:

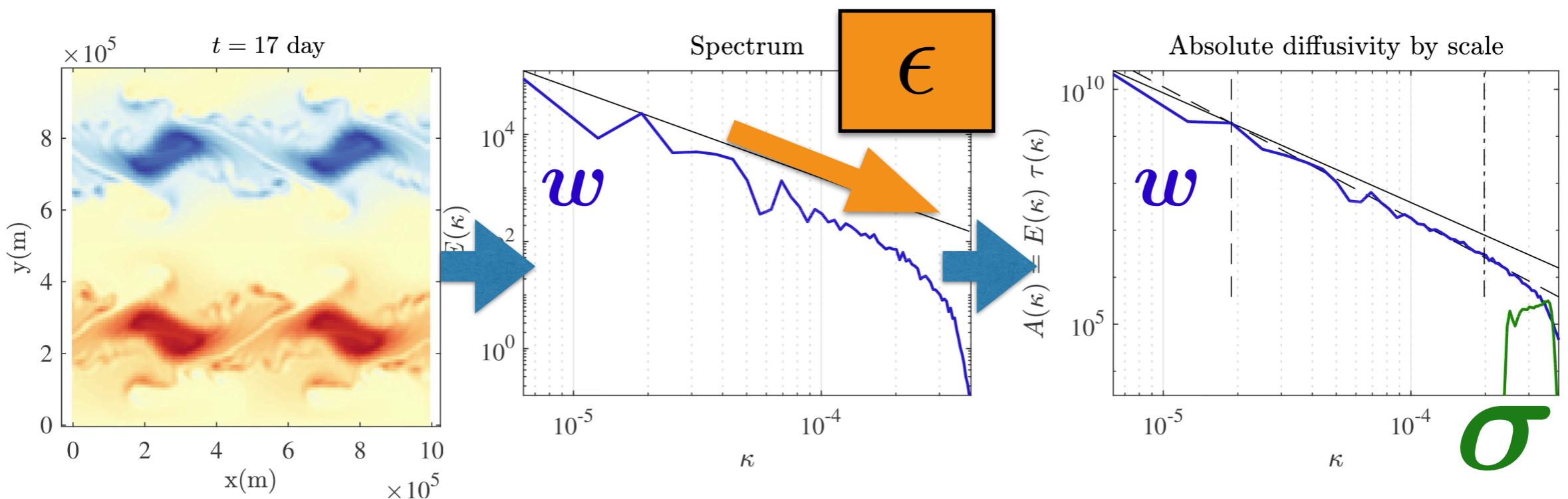
$$\sigma \dot{B}$$

Variance tensor:

$$\mathbf{a} = \mathbf{a}(x, x) = \frac{\mathbb{E}\{\boldsymbol{\sigma} dB (\boldsymbol{\sigma} dB)^T\}}{dt}$$

Heterogeneous modulation of σ

(for Resseguier et al. 2017b)



Large scales:

$$w$$

Small scales:

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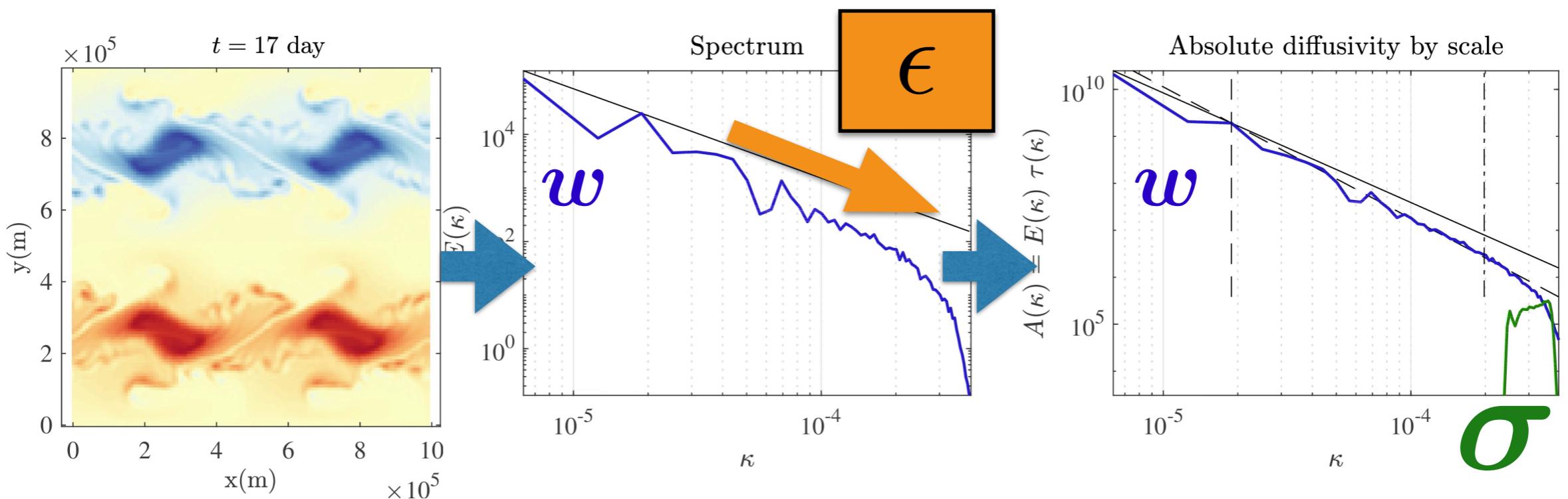
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Variance tensor:

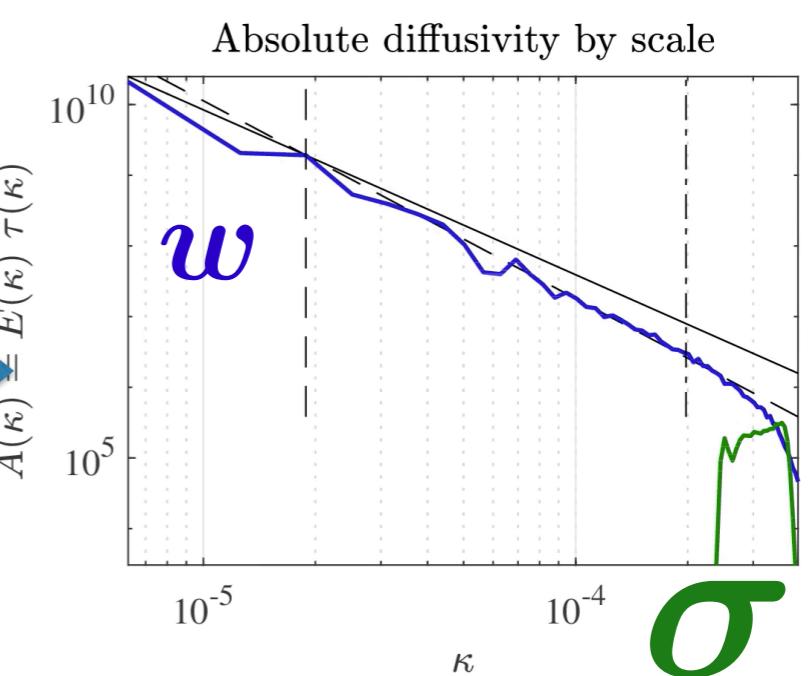
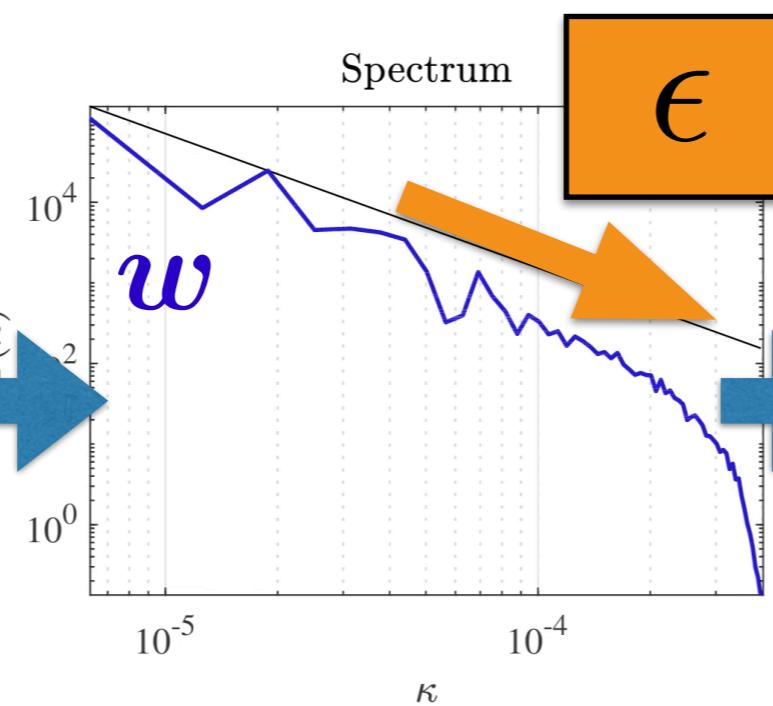
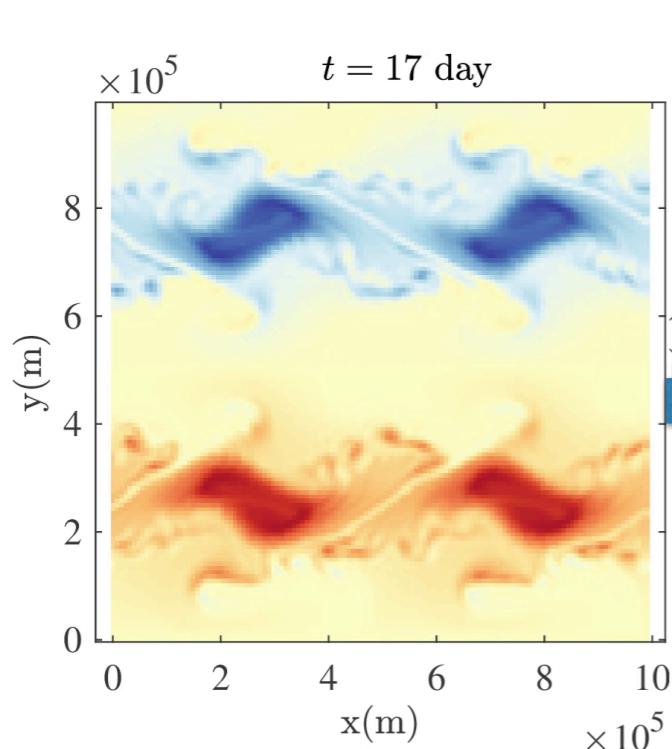
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Heterogeneous

modulation of σ

(for Resseguier et al. 2017b)

$$A(\kappa, x) = E(\kappa, x)\tau(\kappa, x) = \kappa^{-3/2}E^{1/2}(\kappa, x) = \text{cst. } \epsilon^p(x) \kappa^{-q}$$



Large scales:
 w

Small scales:
 $\sigma \dot{B}$

Variance
tensor:

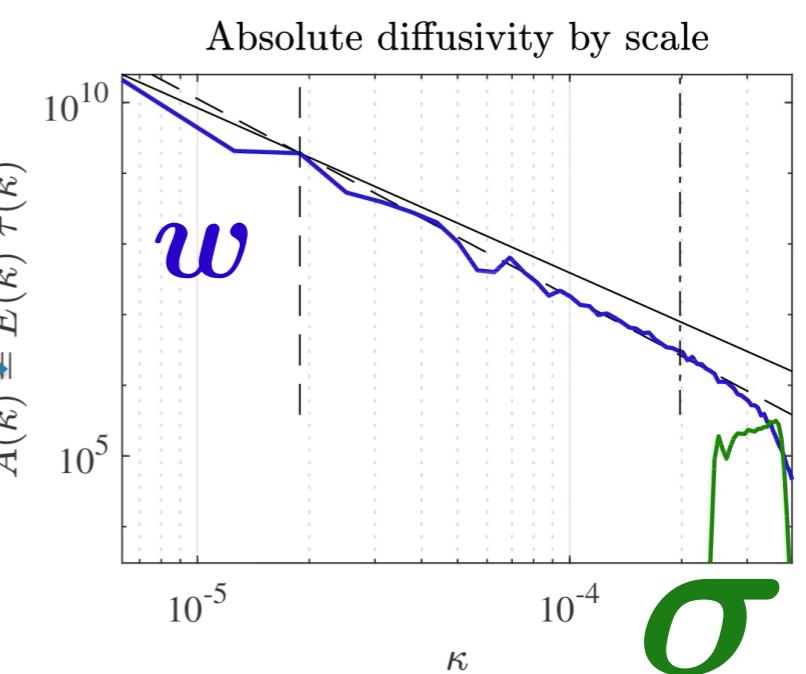
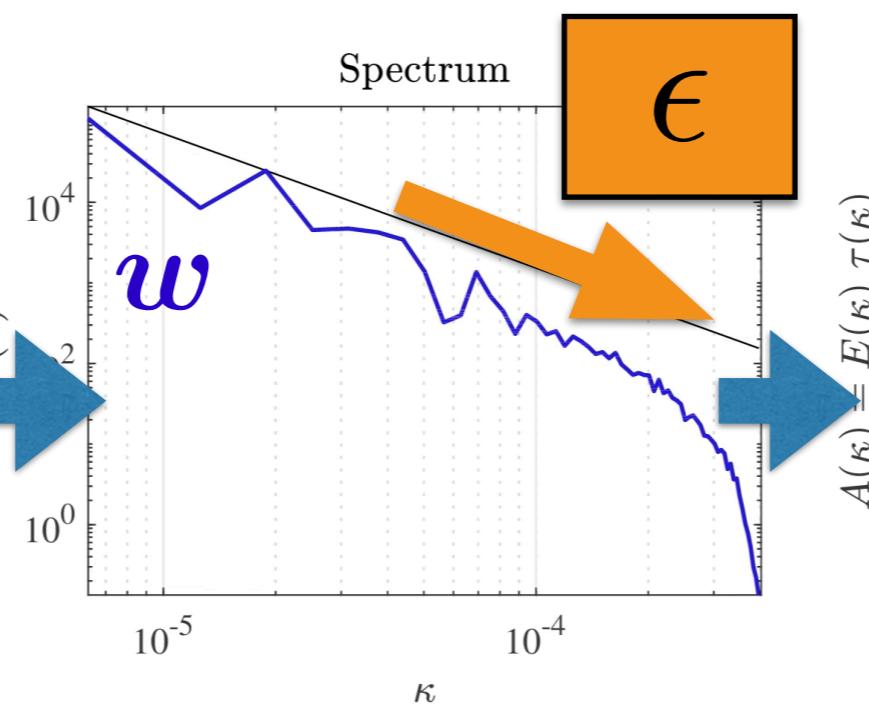
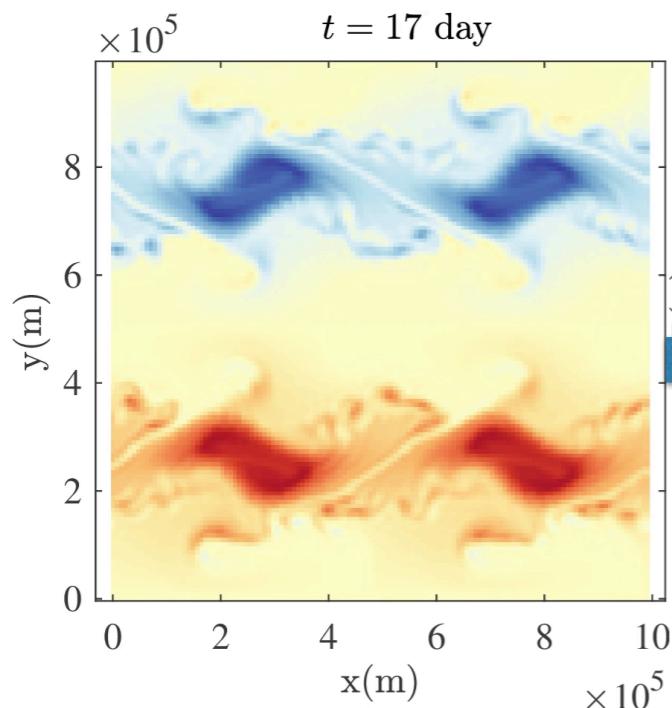
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

Heterogeneous modulation of σ

(for Resseguier et al. 2017b)

Dynamics
(1/3 for SQG)

$$A(\kappa, x) = E(\kappa, x)\tau(\kappa, x) = \kappa^{-3/2}E^{1/2}(\kappa, x) = \text{cst. } \epsilon^p(x) \kappa^{-q}$$



Large scales:
 w

Small scales:
 $\sigma \dot{B}$

Variance
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 $a = a(x, x) =$

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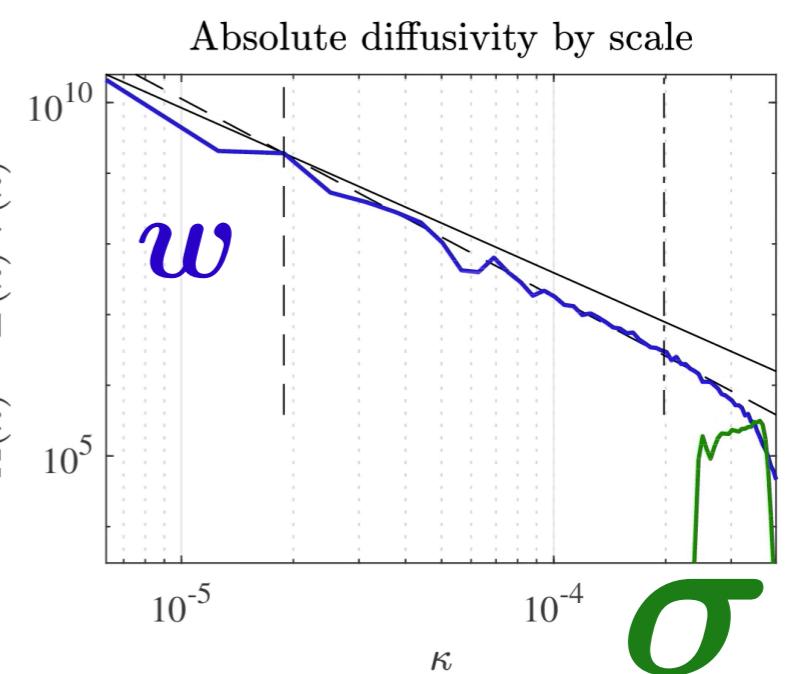
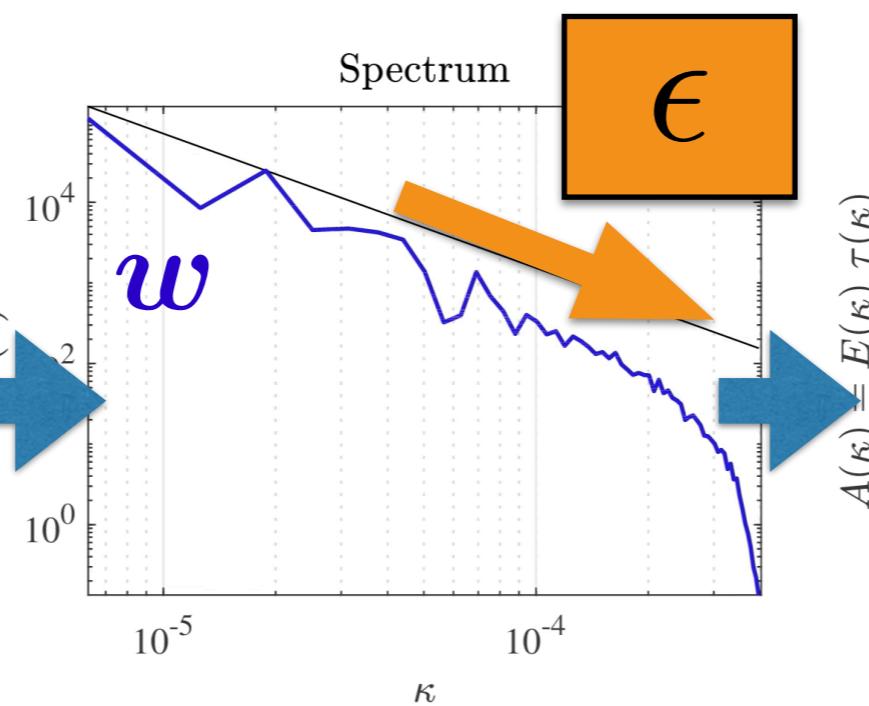
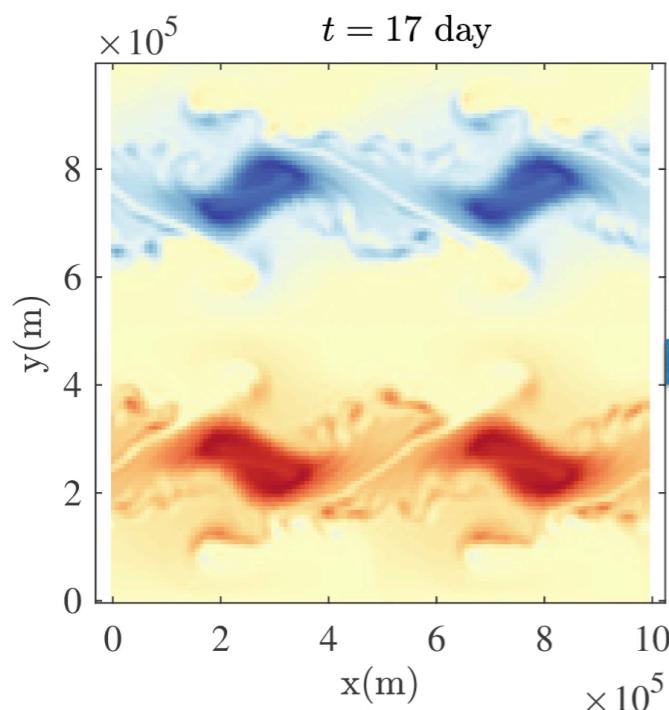
Heterogeneous modulation of σ

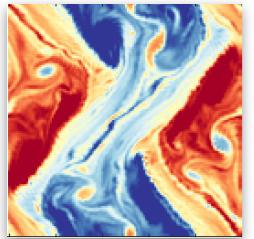
(for Resseguier et al. 2017b)

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Dynamics
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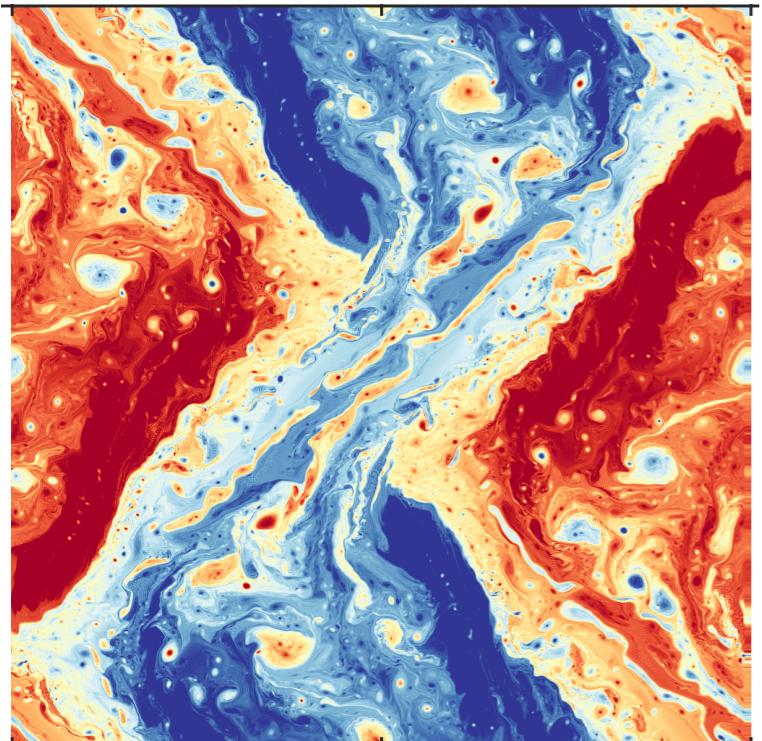
Learned on
large scales at each time t



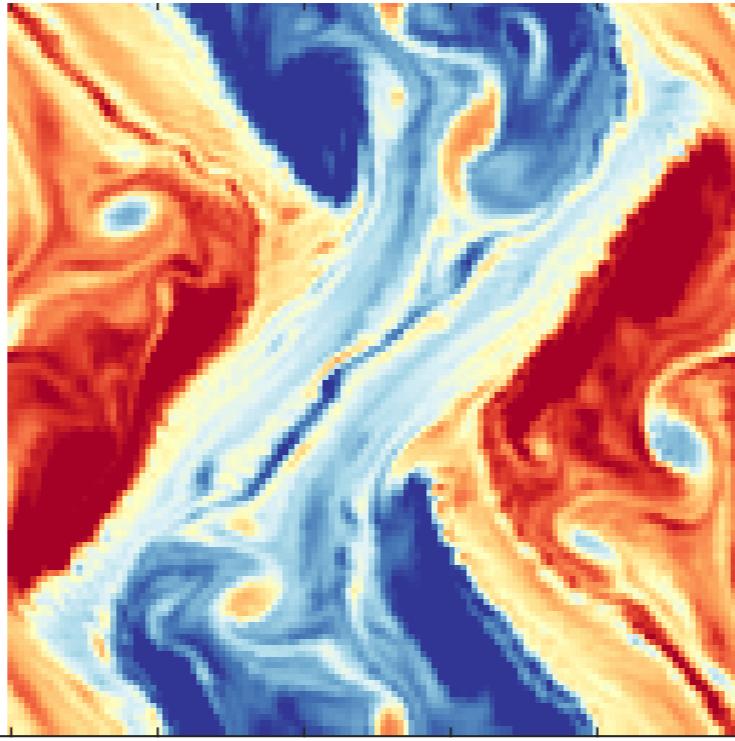


Heterogenous modulation

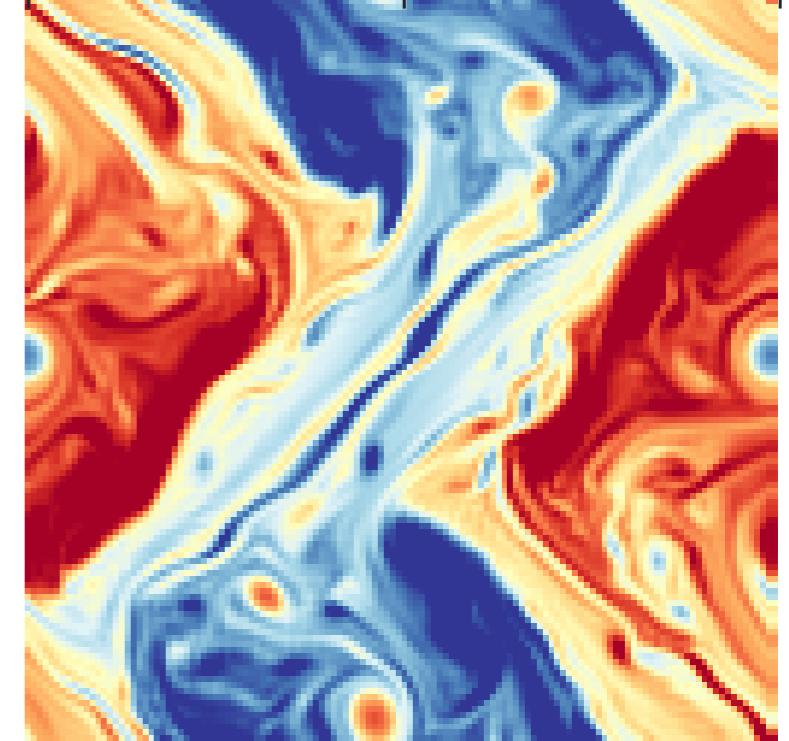
Deterministic 1024 x 1024

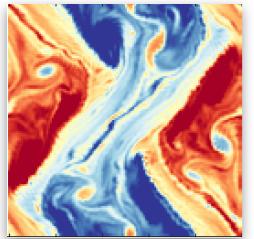


Stochastic 128 x128
with Homogeneous
small-scale velocity



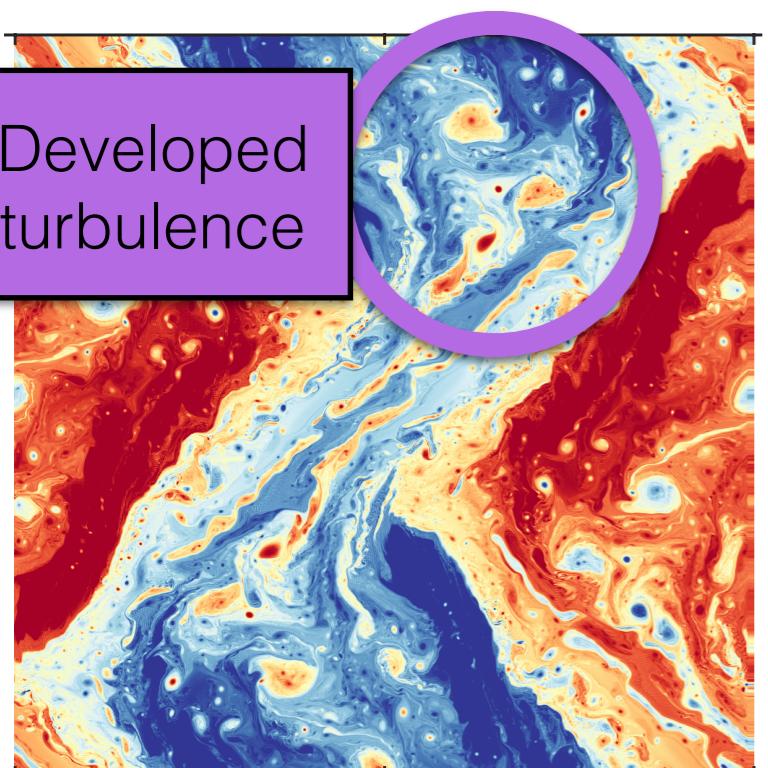
Stochastic 128 x128
with Energy-flux-based
modulation



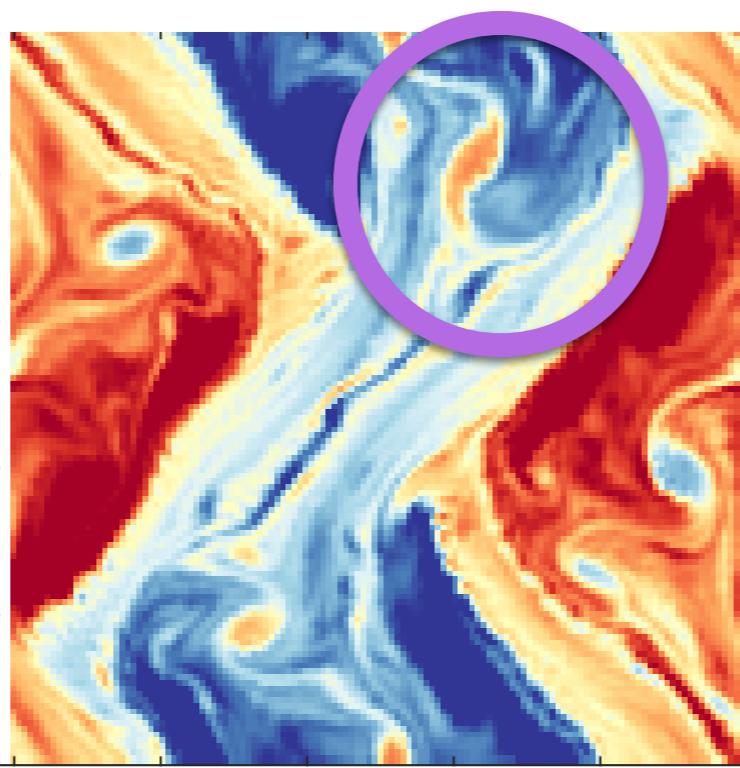


Heterogenous modulation

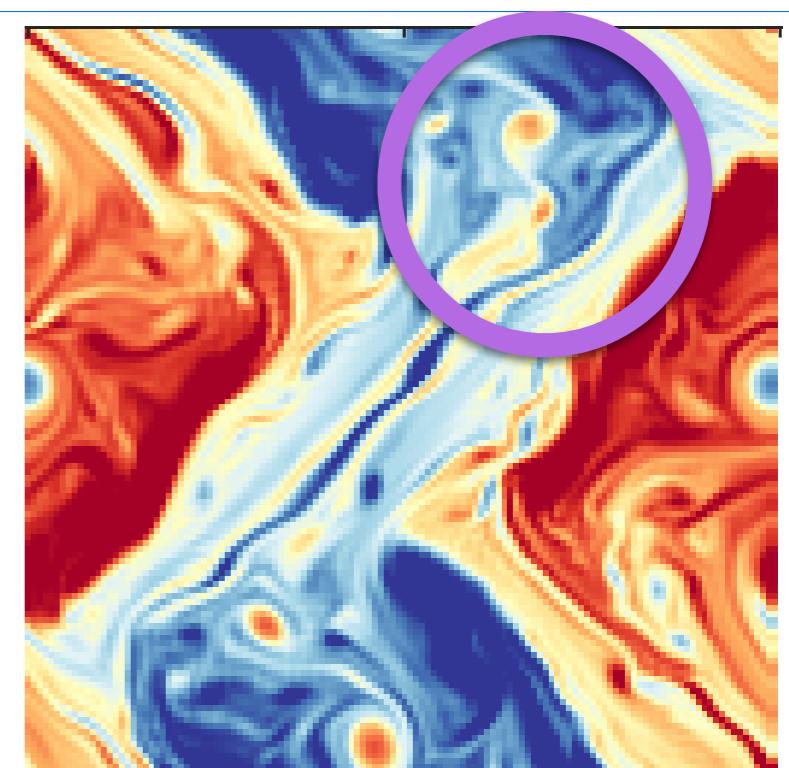
Deterministic 1024 x 1024

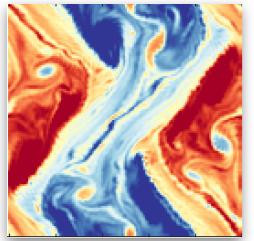


Stochastic 128 x128
with Homogeneous
small-scale velocity



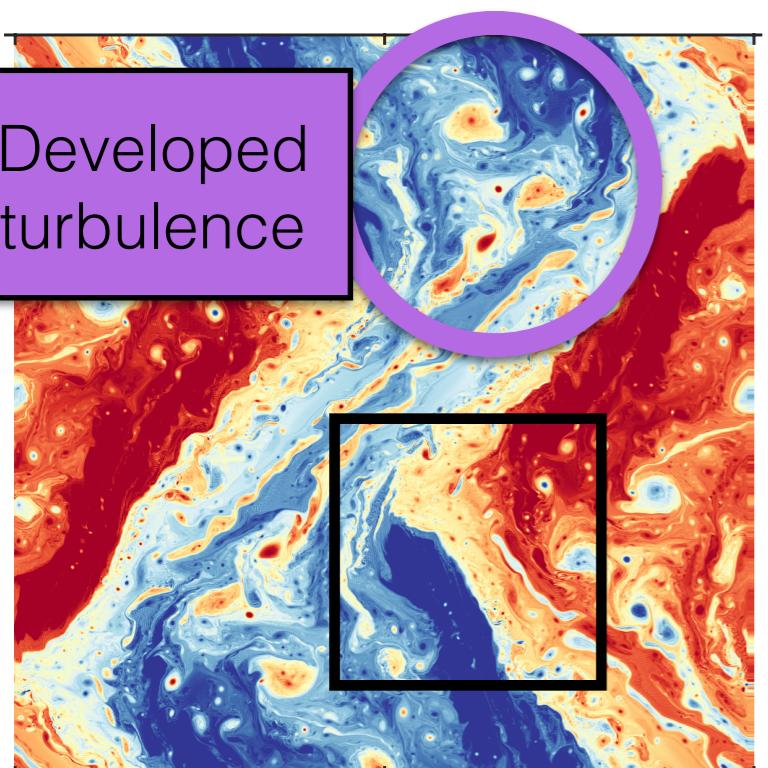
Stochastic 128 x128
with Energy-flux-based
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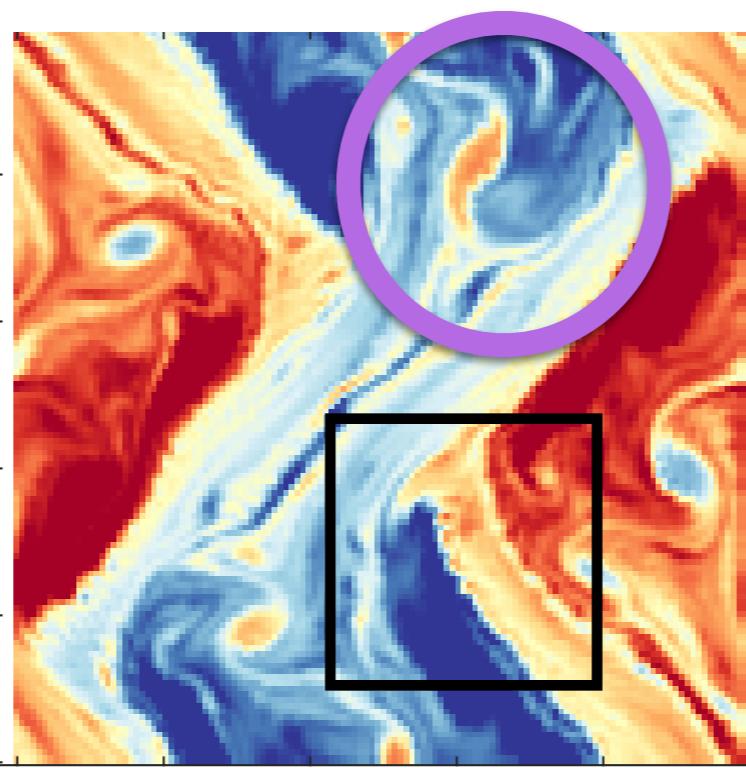


Heterogenous modulation

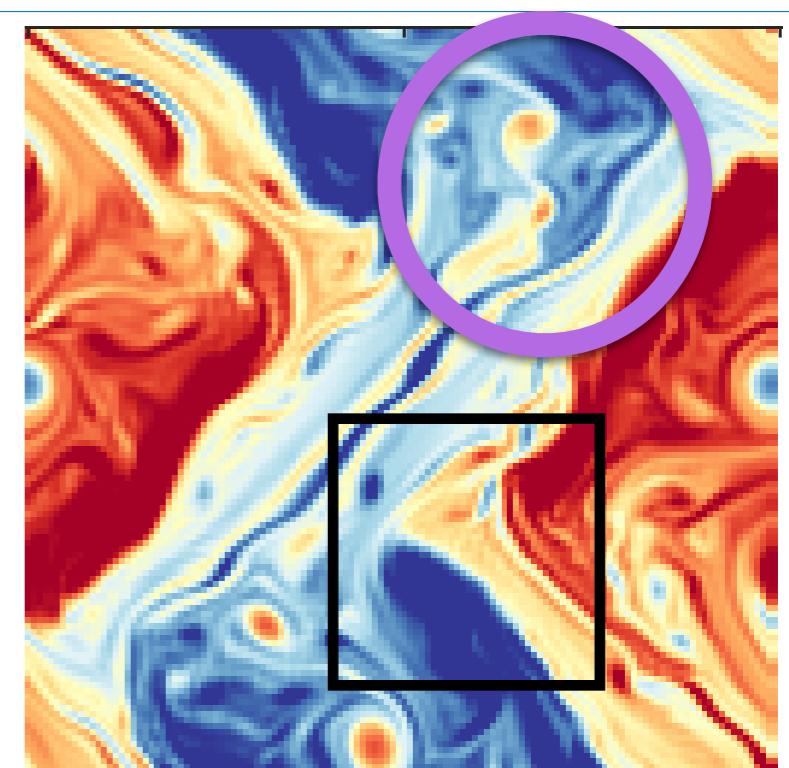
Deterministic 1024 x 1024

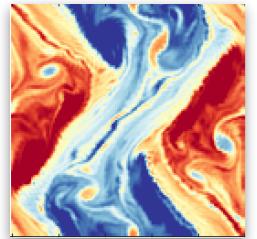


Stochastic 128 x128
with Homogeneous
small-scale velocity



Stochastic 128 x128
with Energy-flux-based
modulation



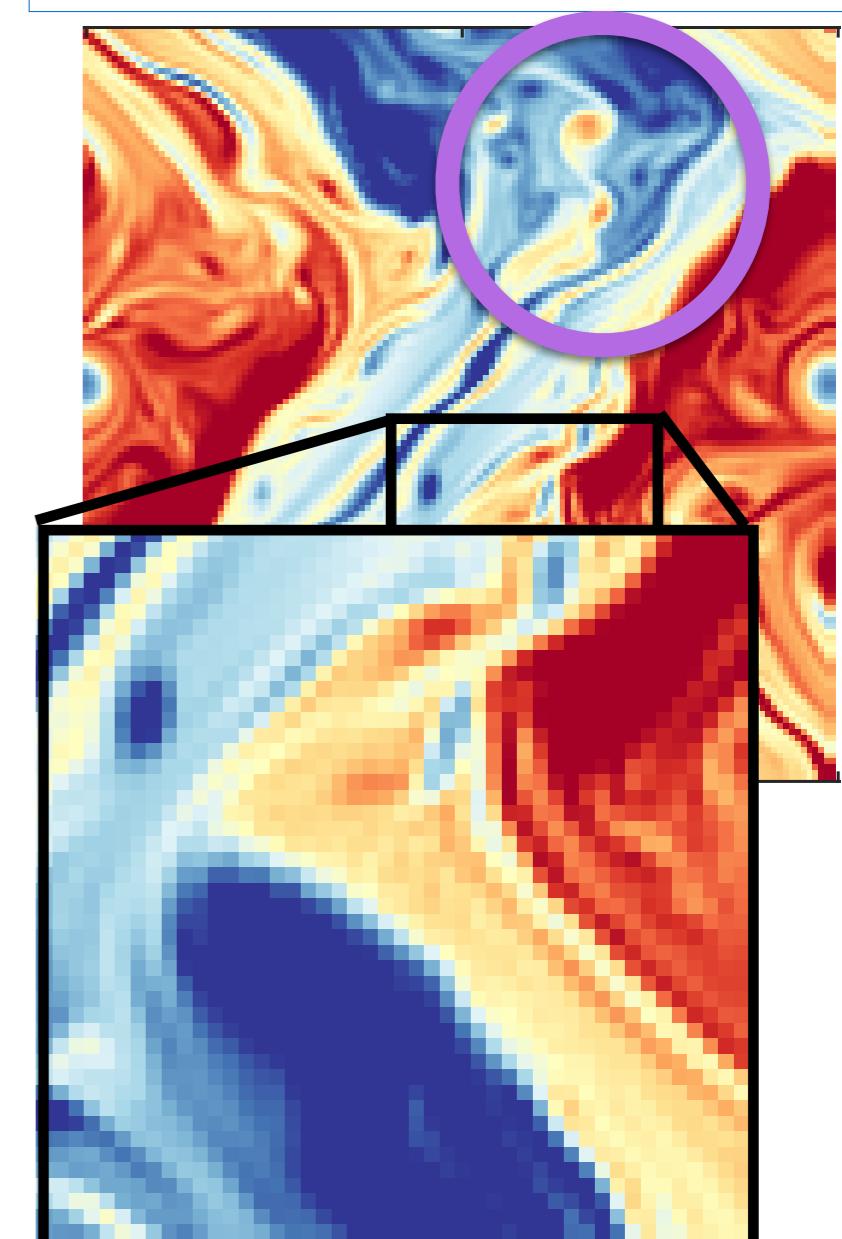
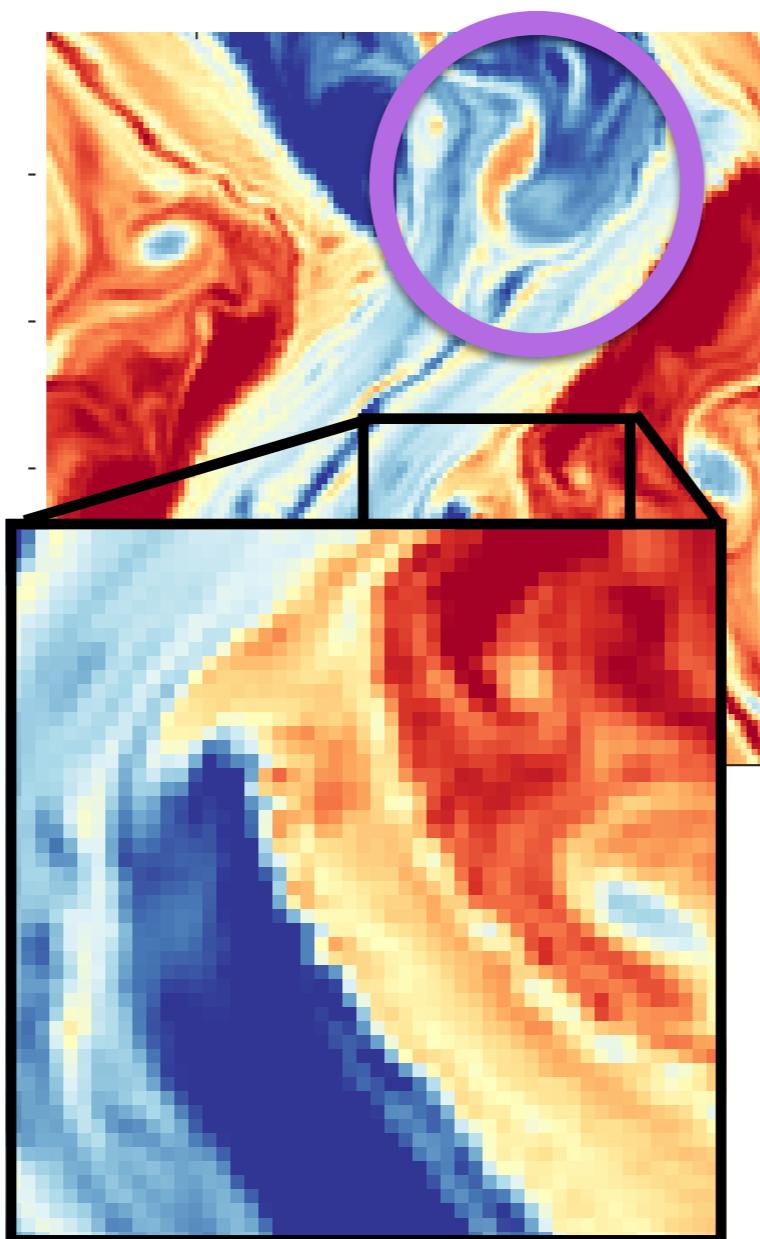
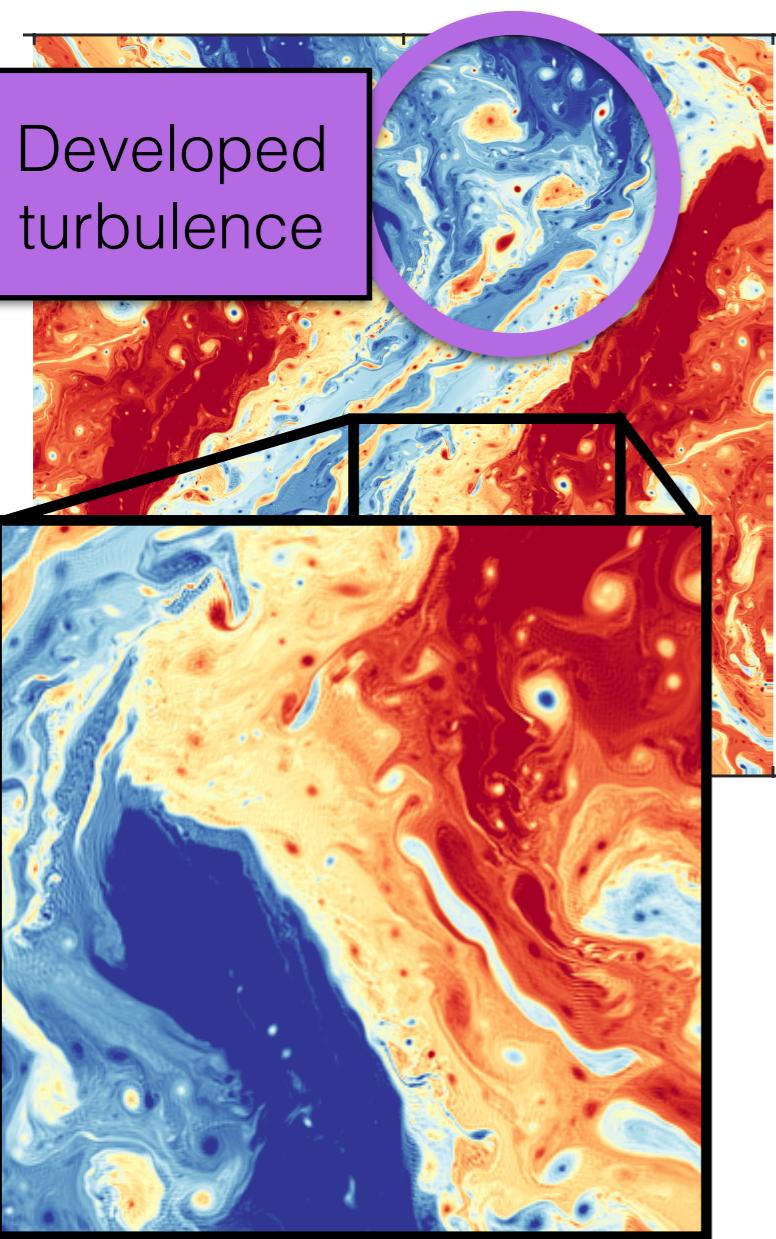


Heterogenous modulation

Deterministic 1024 x 1024

Stochastic 128 x128
with Homogeneous
small-scale velocity

Stochastic 128 x128
with Energy-flux-based
modulation



Conclusion

Conclusion

- Stochastic transport (in LU & SALT) blindly describes unresolved triades
- LU conserves kinetic energy / SALT conserves helicity
- For the small-scale velocity parametrisation, both the data-driven (Cotter et al. 2018b) and the self-similar method (Resseguier et al. 2017b) lead to accurate uncertainty quantification (to address filter divergence)
- Energy-flux modulation improves the simulations

Bibliography

- Cotter, C. J., Crisan, D., Holm, D. D., Pan, W., & Shevchenko, I. (2018). Numerically modelling stochastic lie transport in fluid dynamics. *arXiv preprint arXiv: 1801.09729*.
- Kunita, H., Stochastic Flows and Stochastic Differential Equations, 1990. (Cambridge: Cambridge University Press).
- Resseguier, V., Mémin, E., & Chapron, B. (2017). Geophysical flows under location uncertainty, Part II Quasi-geostrophy and efficient ensemble spreading. *Geophysical & Astrophysical Fluid Dynamics*, 111(3), 177-208.