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# Stochastic Advection by Lie Transport and Location Uncertainty: a common ground for uncertainty quantification in fluid dynamics

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► **To cite this version:**

Baylor Fox-Kemper, Darryl D Holm, Wei Pan, Valentin Resseguier. Stochastic Advection by Lie Transport and Location Uncertainty: a common ground for uncertainty quantification in fluid dynamics. American Institute of Mathematical Sciences, Jul 2018, Taipei, Taiwan. hal-01891177

**HAL Id: hal-01891177**

**<https://hal.science/hal-01891177>**

Submitted on 9 Oct 2018

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# Stochastic Advection by Lie Transport and Location Uncertainty: a common ground for uncertainty quantification in fluid dynamics

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

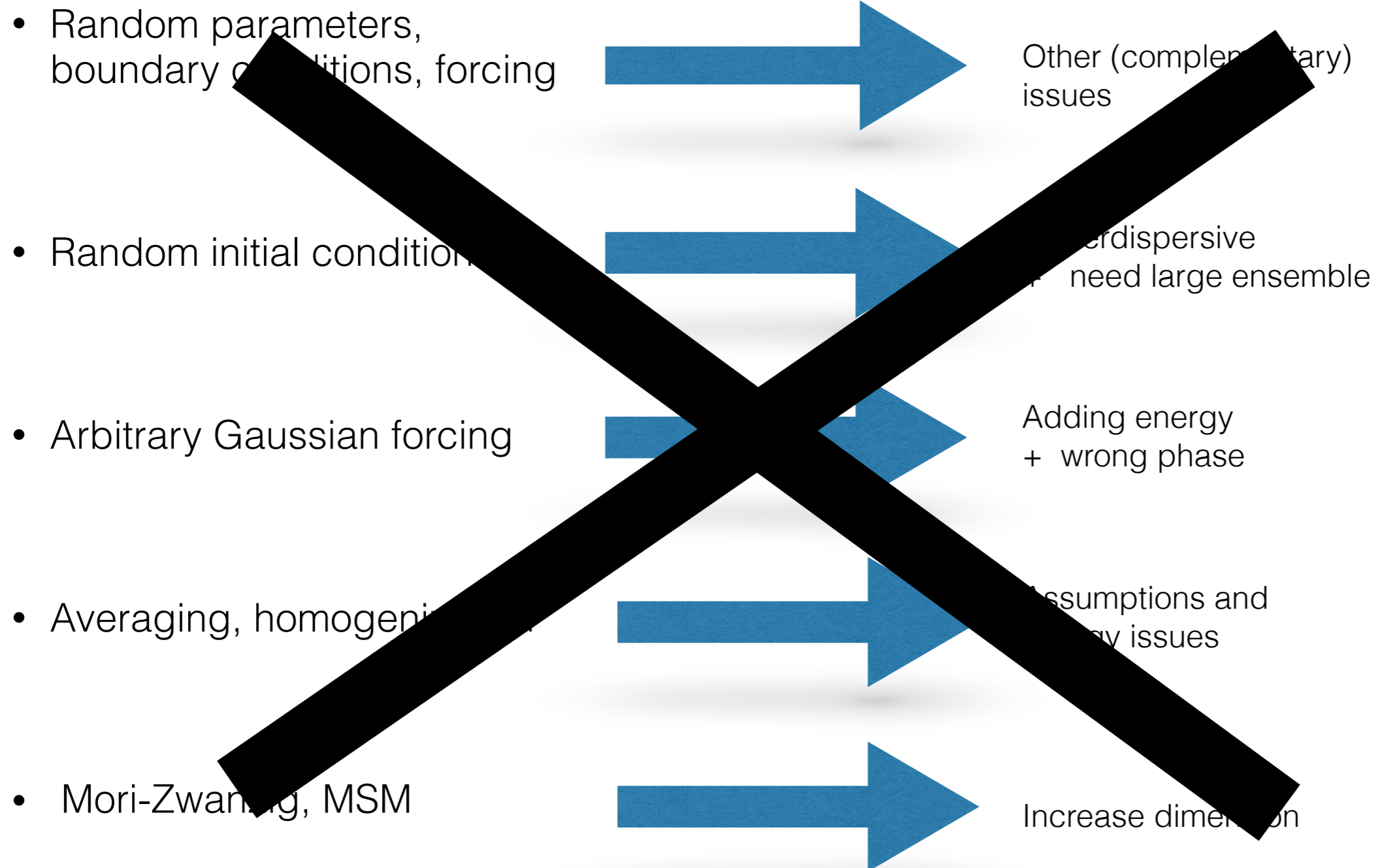
# Part I

## Stochastic transport

# Usual random CFD

- Random parameters, boundary conditions, forcing  Other (complementary) issues
- Random initial conditions  Underdispersive  
+ need large ensemble
- Arbitrary Gaussian forcing  Adding energy  
+ wrong phase
- Averaging, homogenization  Assumptions and energy issues
- Mori-Zwanzig, MSM  Increase dimension

# Usual random CFD



SALT & LU :  
Adding random velocity

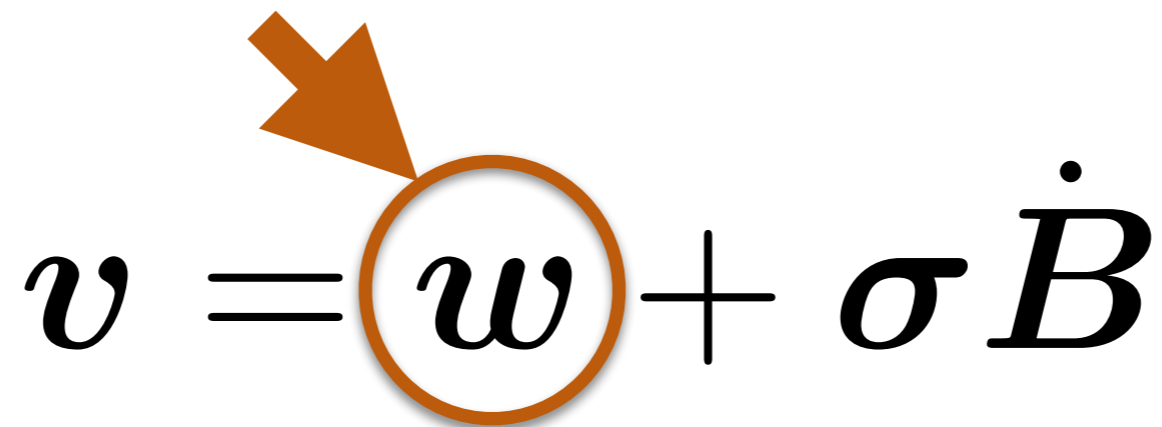
$$\mathbf{v} = \mathbf{w} + \sigma \dot{\mathbf{B}}$$



# SALT & LU :

## Adding random velocity

Resolved  
large scales

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


# SALT & LU :

## Adding random velocity

Resolved  
large scales

White-in-time  
small scales

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

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

The diagram shows the equation  $v = w + \sigma \dot{B}$ . The term  $w$  is enclosed in a brown circle, and a brown arrow points to it from the text "Resolved large scales". The term  $\sigma \dot{B}$  is enclosed in a purple circle, and a purple arrow points to it from the text "White-in-time small scales".

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

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

# SALT & LU :

Adding random velocity

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

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

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

# SALT & LU :

## Adding random velocity

Stratonovich

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

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

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

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

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

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

Stratonovich

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

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

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:

$\sigma \dot{B}$

Variance tensor:

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

# SALT & LU :

## Adding random velocity

Stratonovich

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

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

Ito

References : Mikulevicius and Rozovskii, 2004  
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:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

# Advection of tracer $\Theta$

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

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

# Advection of tracer $\Theta$

Ito-Wentzell  
formula  
(Kunita 1990)

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

# Advection of tracer $\Theta$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

Ito-Wentzell  
formula  
(Kunita 1990)

# Advection of tracer $\Theta$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

Ito-Wentzell  
formula  
(Kunita 1990)

Stratonovich notations:

# Advection of tracer $\Theta$

Ito-Wentzell  
formula  
(Kunita 1990)

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

Stratonovich notations:

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

# Advection of tracer $\Theta$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

Ito-Wentzell  
formula  
(Kunita 1990)

Ito notations:

# Advection of tracer $\Theta$

Ito-Wentzell  
formula  
(Kunita 1990)

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

Ito notations:

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

# Advection of tracer $\Theta$

Ito-Wentzell  
formula  
(Kunita 1990)

Large scales:

$w$

Small scales:

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Variance  
tensor:

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(Kunita 1990)

Large scales:

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

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Ito-Wentzell  
formula  
(Kunita 1990)

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

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

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

Stratonovich drift : « Drift correction »  
in Ito notations

# Advection of tracer $\Theta$

Ito-Wentzell formula  
(Kunita 1990)

Large scales:  
 $w$   
Small scales:  
 $\sigma \dot{B}$   
Variance tensor:  
 $a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$

Multiplicative random forcing

Ito notations:

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

Advection Diffusion

Stratonovich drift : « Drift correction » in Ito notations

# Advection of tracer $\Theta$

Ito-Wentzell formula  
(Kunita 1990)

Large scales:  
 $w$

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

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

Multiplicative random forcing

Ito notations:

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

Stratonovich drift : « Drift correction »  
in Ito notations

# Advection of tracer $\Theta$

Ito-Wentzell formula  
(Kunita 1990)

Large scales:  
 $w$   
Small scales:  
 $\sigma \dot{B}$   
Variance tensor:  
 $a = a(x, x) = \frac{\mathbb{E}\{\sigma d\mathbf{B} (\sigma d\mathbf{B})^T\}}{dt}$

Multiplicative random forcing

Ito notations:

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

Stratonovich drift : « Drift correction » in Ito notations

# Advection of tracer $\Theta$

Ito-Wentzell formula  
(Kunita 1990)

Large scales:  
 $w$   
Small scales:  
 $\sigma \dot{B}$   
Variance tensor:  
 $a = a(x, x) = \frac{\mathbb{E}\{\sigma d\mathbf{B} (\sigma d\mathbf{B})^T\}}{dt}$

Multiplicative random forcing

Balanced energy exchanges

Ito notations:

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

Stratonovich drift : « Drift correction » in Ito notations

Large scales:

$$\mathbf{w}$$

Small scales:

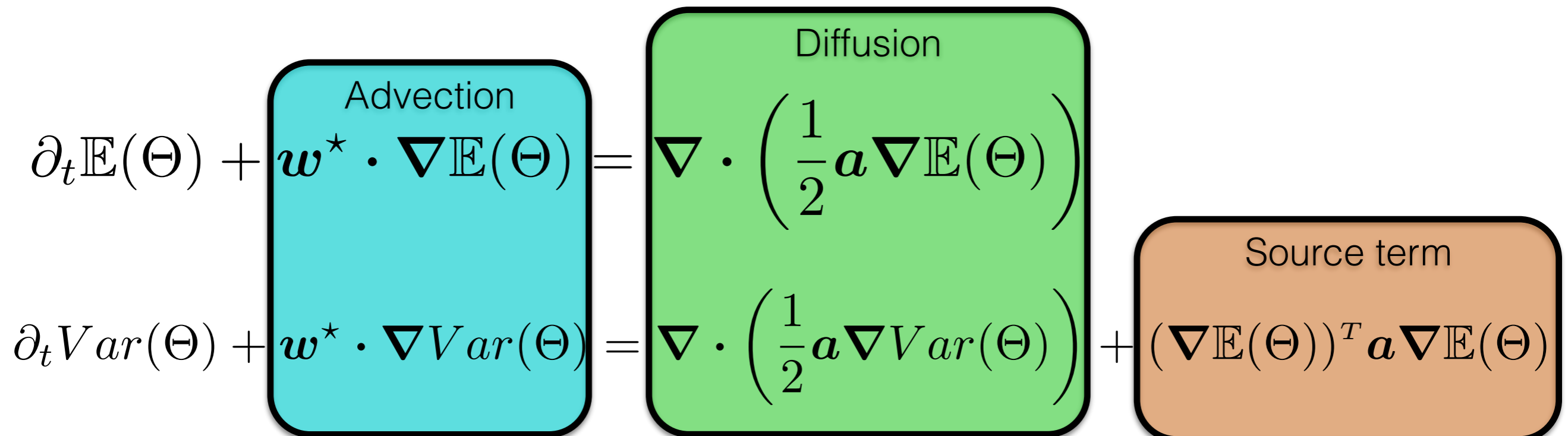
$$\sigma \dot{B}$$

Variance tensor:

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

# Moments of a passive tracer $\Theta$

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



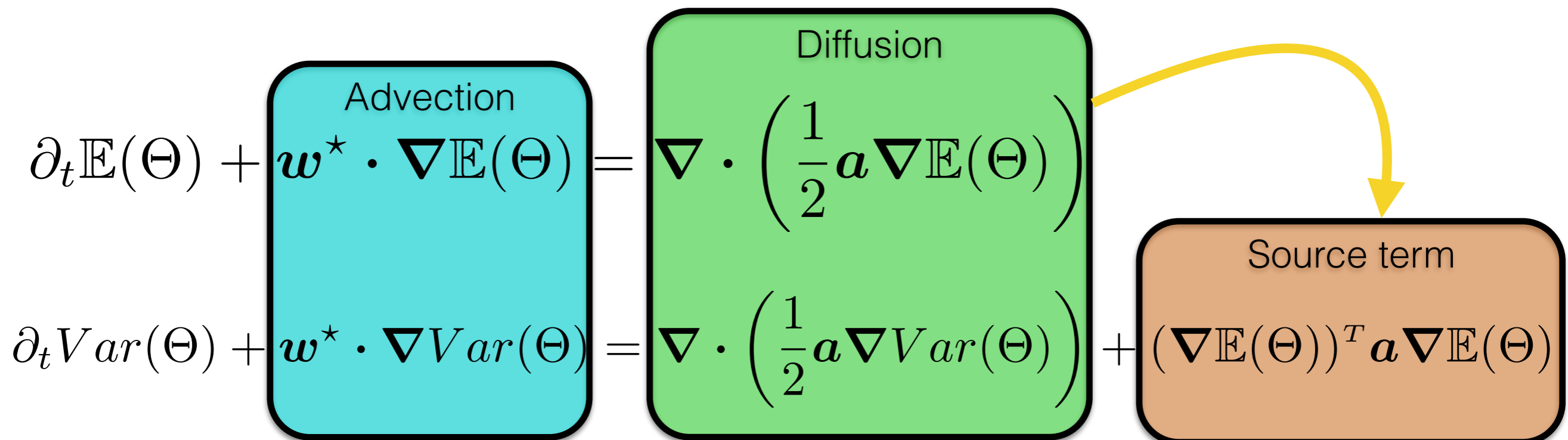
Large scales:  
 $w$

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

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

# Moments of a passive tracer $\Theta$

$$\Theta(\bullet, t') \perp\!\!\!\perp w(\bullet, t), \forall t' \leq t$$





Large scales:

$w$

Small scales:

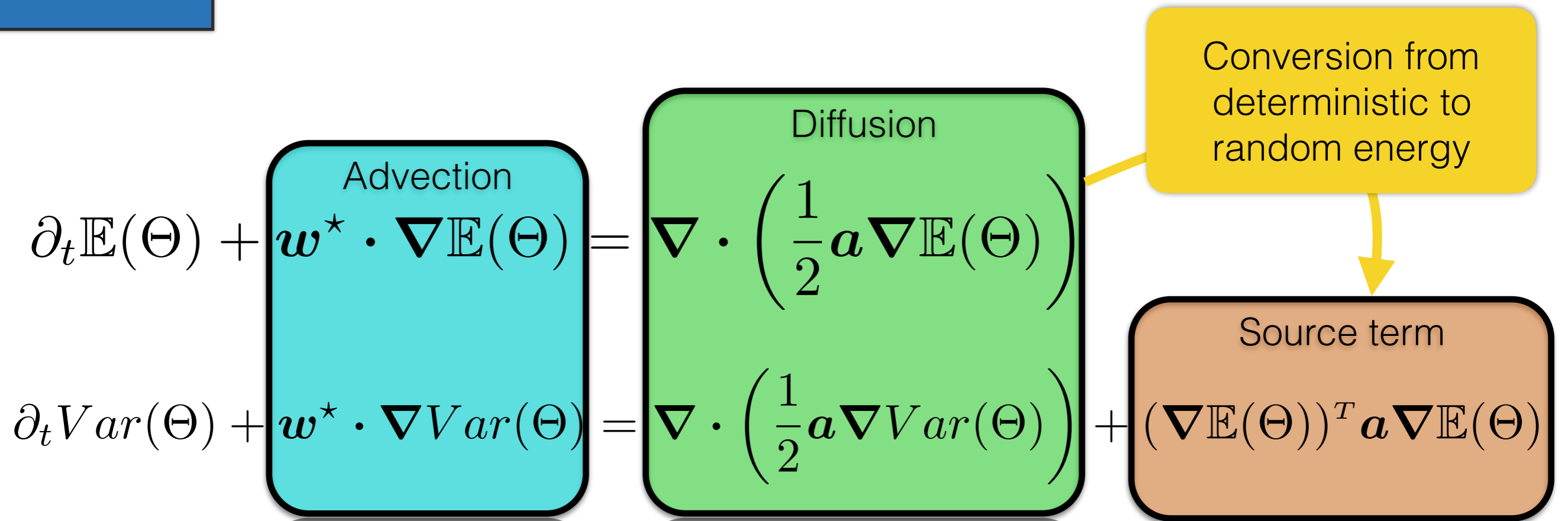
$\sigma \dot{B}$

Variance tensor:

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

# Moments of a passive tracer $\Theta$

$$\Theta(\bullet, t') \perp\!\!\!\perp w(\bullet, t), \forall t' \leq t$$



Large scales:

$$\mathbf{w}$$

Small scales:

$$\sigma \dot{B}$$

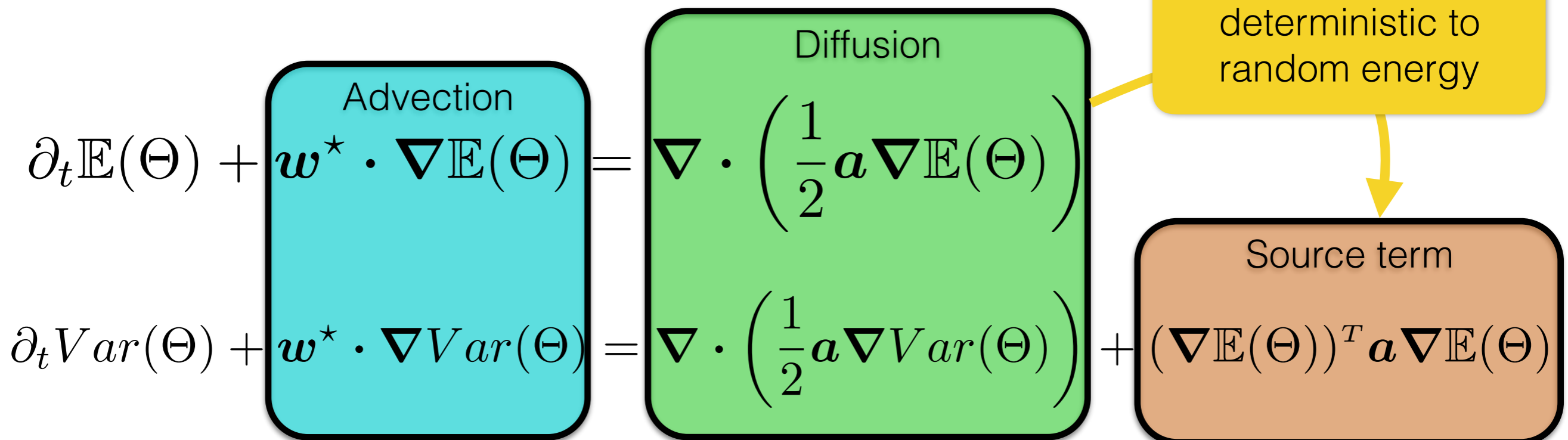
Variance tensor:

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

# Moments of a passive tracer $\Theta$

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

Conservation of the energy  $\frac{d}{dt} \int_{\Omega} \Theta^2 = 0$



# Part II

## Stochastic Navier-Stokes

### SALT vs LU

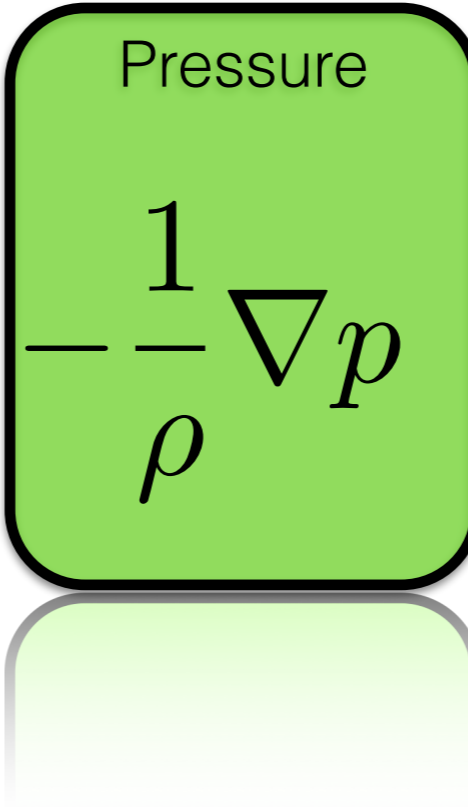
# Comparison

	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

$$\frac{Dw}{Dt} = \text{Pressure} \left[ -\frac{1}{\rho} \nabla p \right]$$


# Navier-Stokes LU

Stochastic  
transport  
of Itô drift

$$\frac{Dw}{Dt} =$$

Pressure

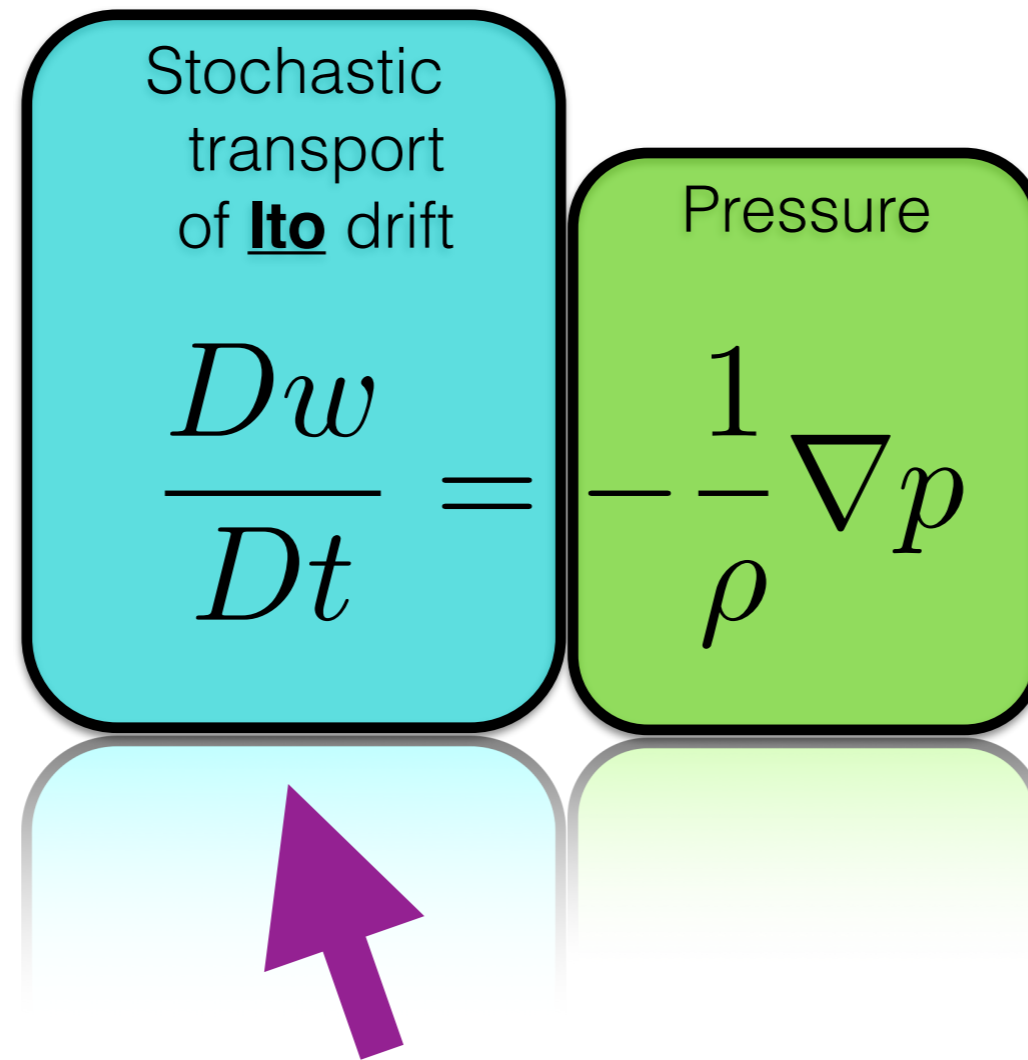
$$-\frac{1}{\rho} \nabla p$$

# Navier-Stokes LU

Stochastic transport of Ito drift

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

Pressure



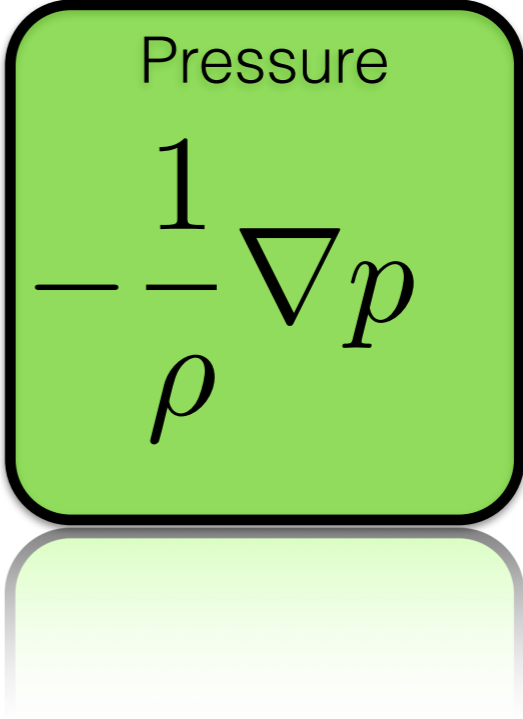
Conserve (Ito) kinetic energy



# Navier-Stokes SALT

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

# Navier-Stokes SALT

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


# Navier-Stokes SALT

Stochastic  
transport of  
**Stratonovich** drift

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

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

Pressure

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

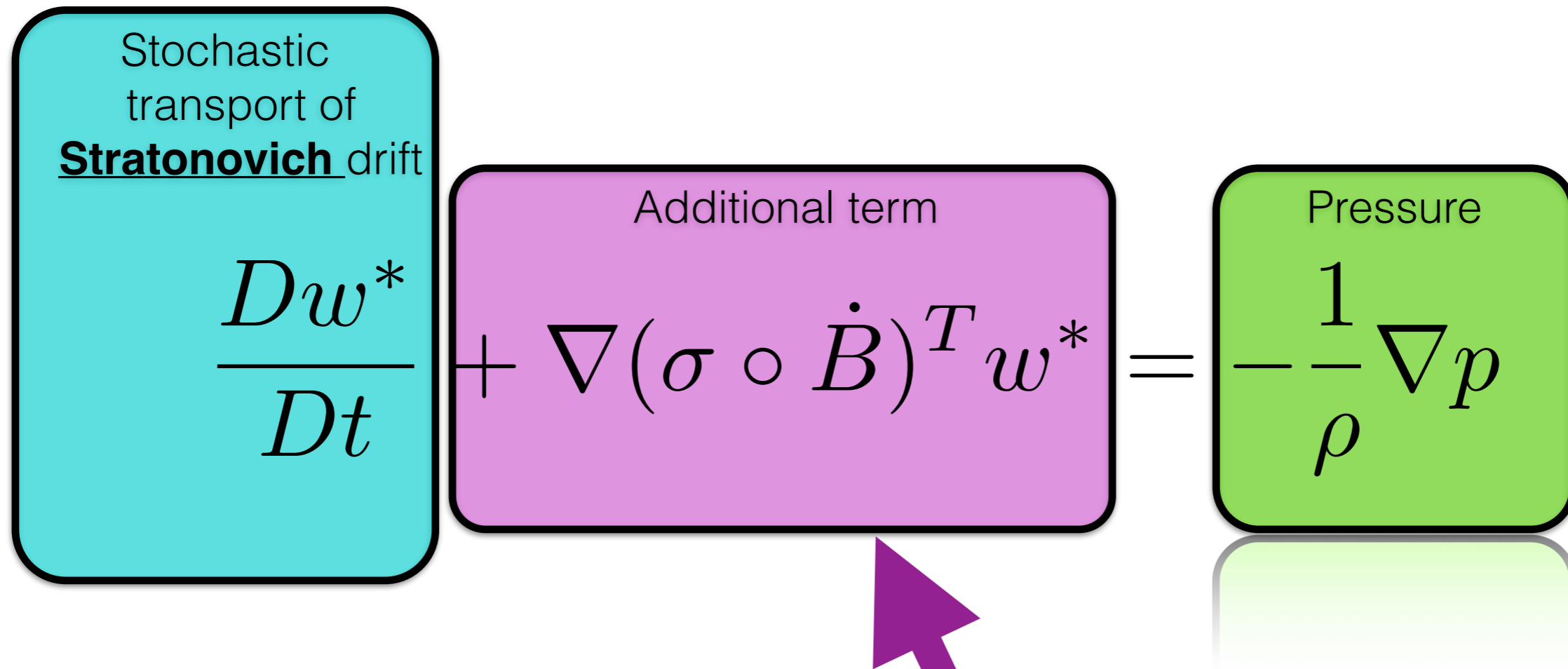
# Navier-Stokes SALT

Stochastic transport of **Stratonovich** drift

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

Additional term

Pressure



Conserve (Stratonovich) helicity

# Part III

## Unresolved velocity parametrisation

*Code available online*

Large scales:

$$w$$

Small scales:

$$\sigma \dot{B}$$

Variance tensor:

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

# Parameter-free models for $\sigma$

	Cotter et al. 2018b	Resseguier et al. 2017b
<b>Applicable to :</b>	LU & SALT	LU & SALT
<b>Homogeneous</b>	No	Yes
<b>Stationary</b>	Yes	Not anymore
<b>Self-similar assumption</b>	No	Yes
<b>Data-driven</b>	Yes	No

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)

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

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

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

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iid  
Brownian  
motion



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Small scales:

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Variance  
tensor:

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

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

weighted  
EOFs learned  
on data

iid  
Brownian  
motion

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

Large scales:

$w$

Small scales:

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tensor:

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

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

iid  
Brownian  
motion

weighted  
EOFs learned  
on data

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

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

Large scales:

$w$

Small scales:

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tensor:

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iid  
Brownian  
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EOFs learned  
on data

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

Flow from **high-**  
resolution velocity

Flow from **low-**  
resolution velocity

(improvement of)  
 Resseguier et al. 2017b:  
 Self-similar model

Large scales:

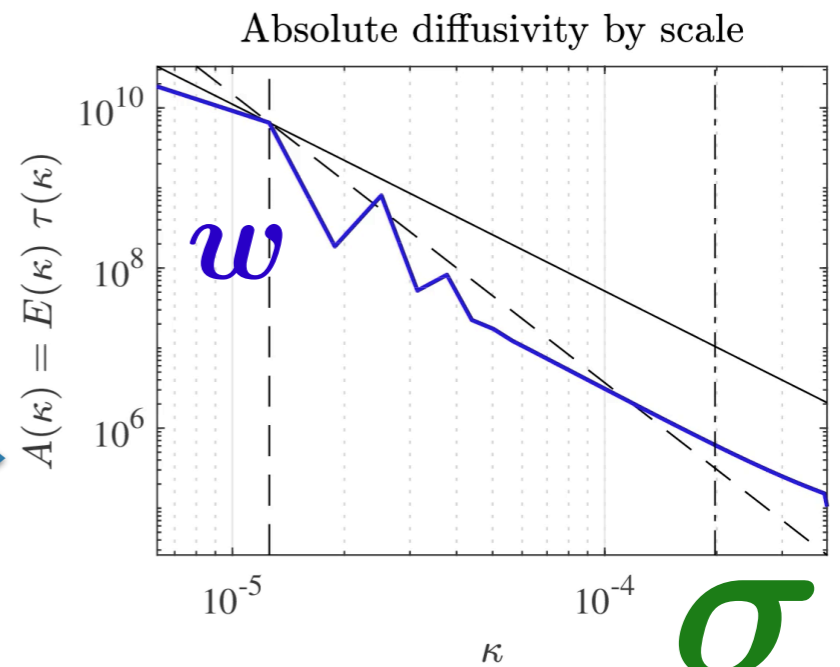
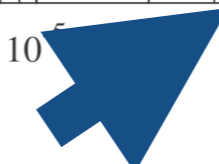
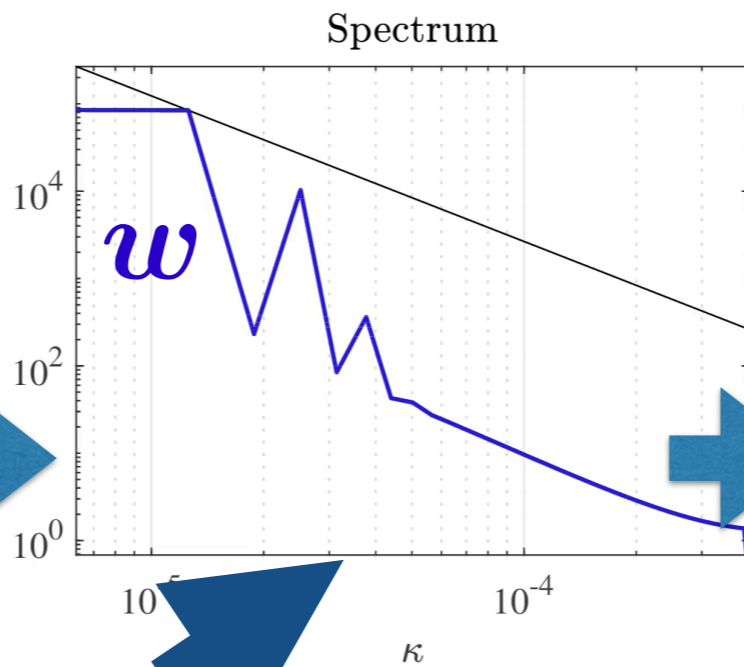
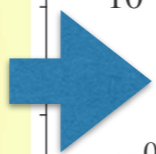
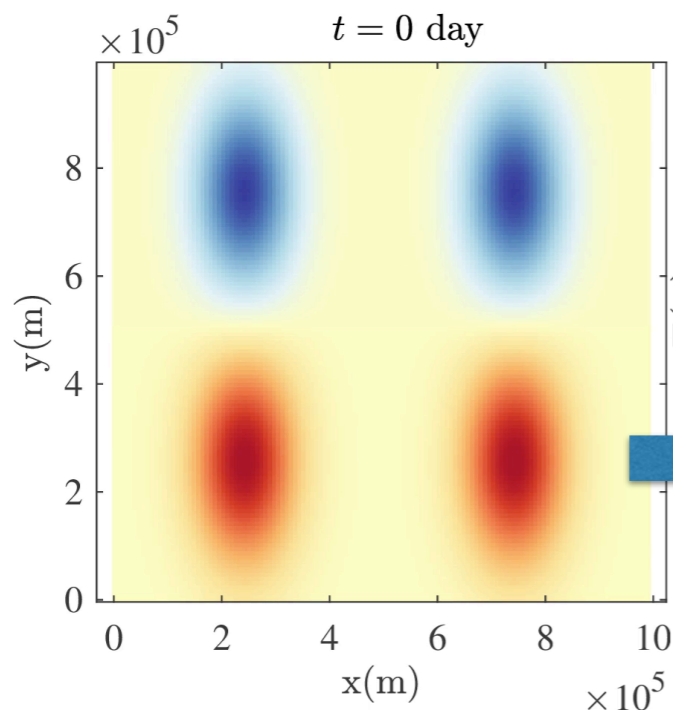
$w$

Small scales:

$\sigma \dot{B}$

Variance  
 tensor:

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



Kinetic energy  
 Spectrum

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

(improvement of)  
 Resseguier et al. 2017b:  
 Self-similar model

Large scales:

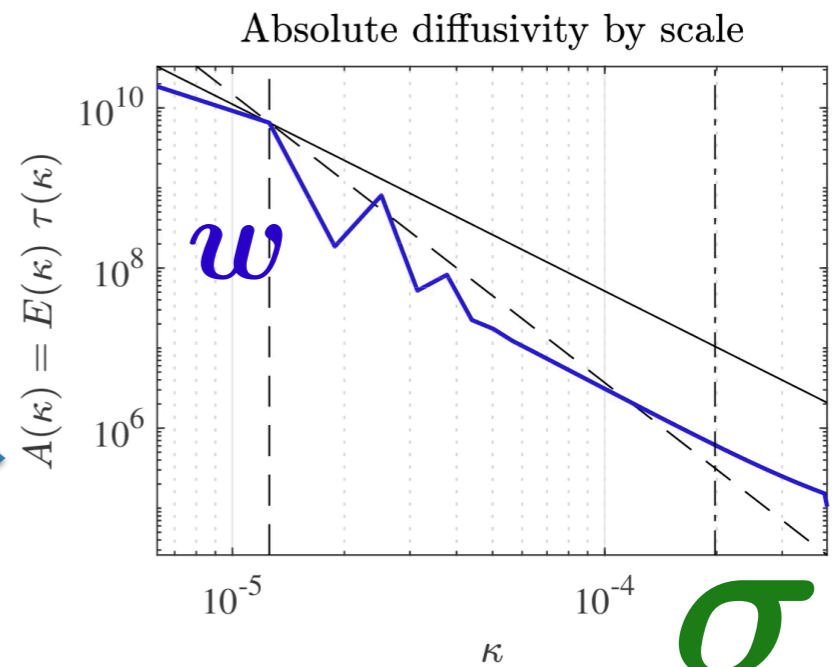
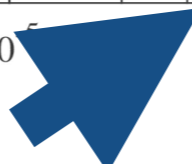
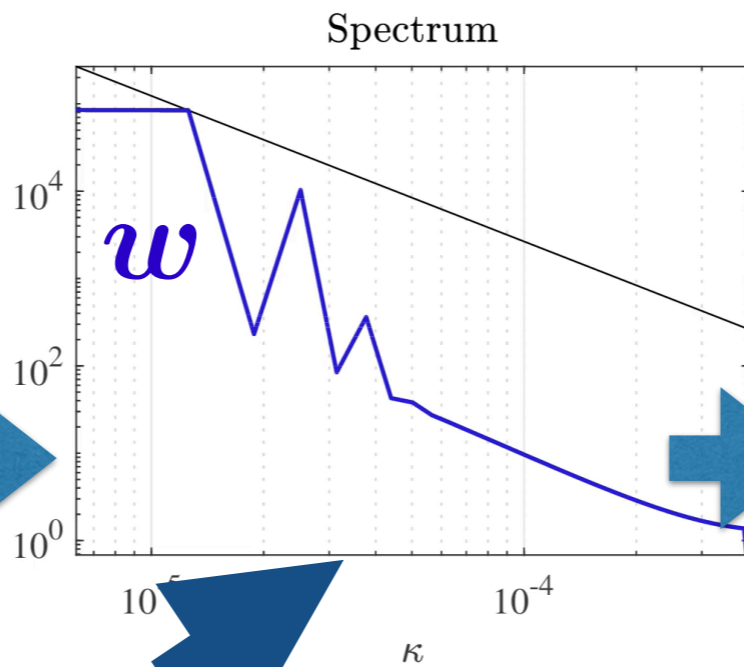
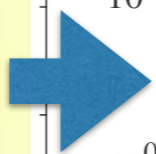
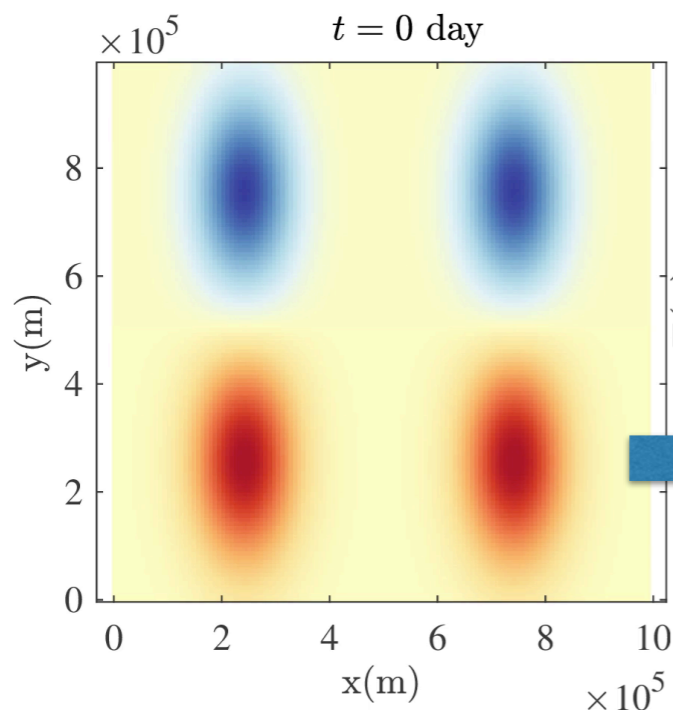
$w$

Small scales:

$\sigma \dot{B}$

Variance  
 tensor:

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



Kinetic energy  
 Spectrum

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

# (improvement of) Resseguier et al. 2017b: Self-similar model

Large scales:

$w$

Small scales:

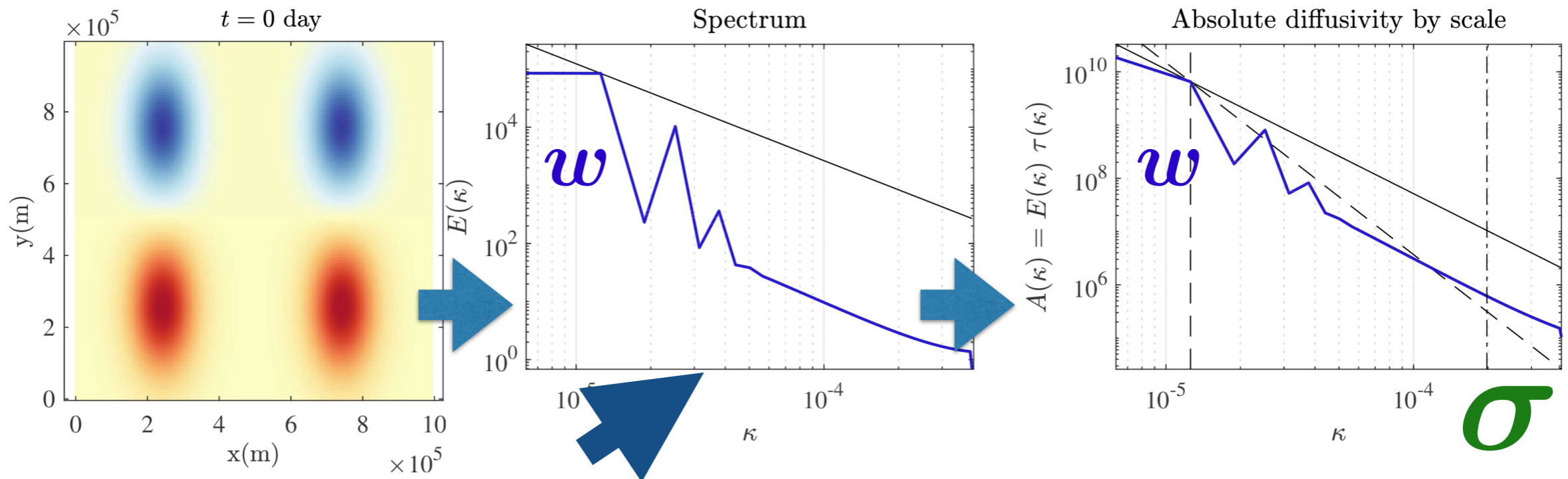
$\sigma \dot{B}$

Variance  
tensor:

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

Absolute Diffusivity  
Spectral Density

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



Kinetic energy  
Spectrum

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

$t = 100$  day

## SQG

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

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

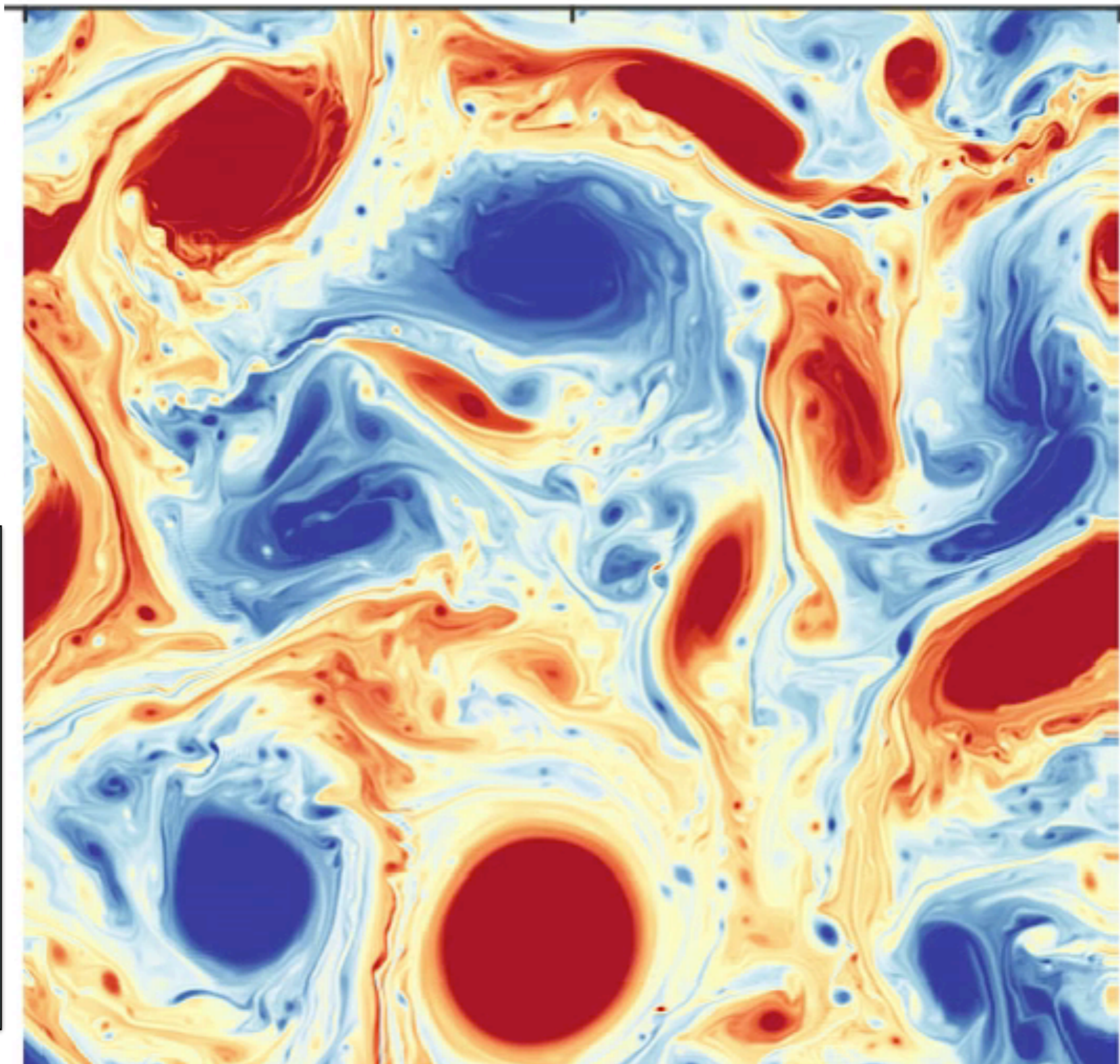
$$u = \left( \text{cst.} \nabla^\perp \Delta^{-1/2} \right) b$$

Reference flow:

deterministic

SQG

512 x 512



$t = 100$  day

## SQG

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

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

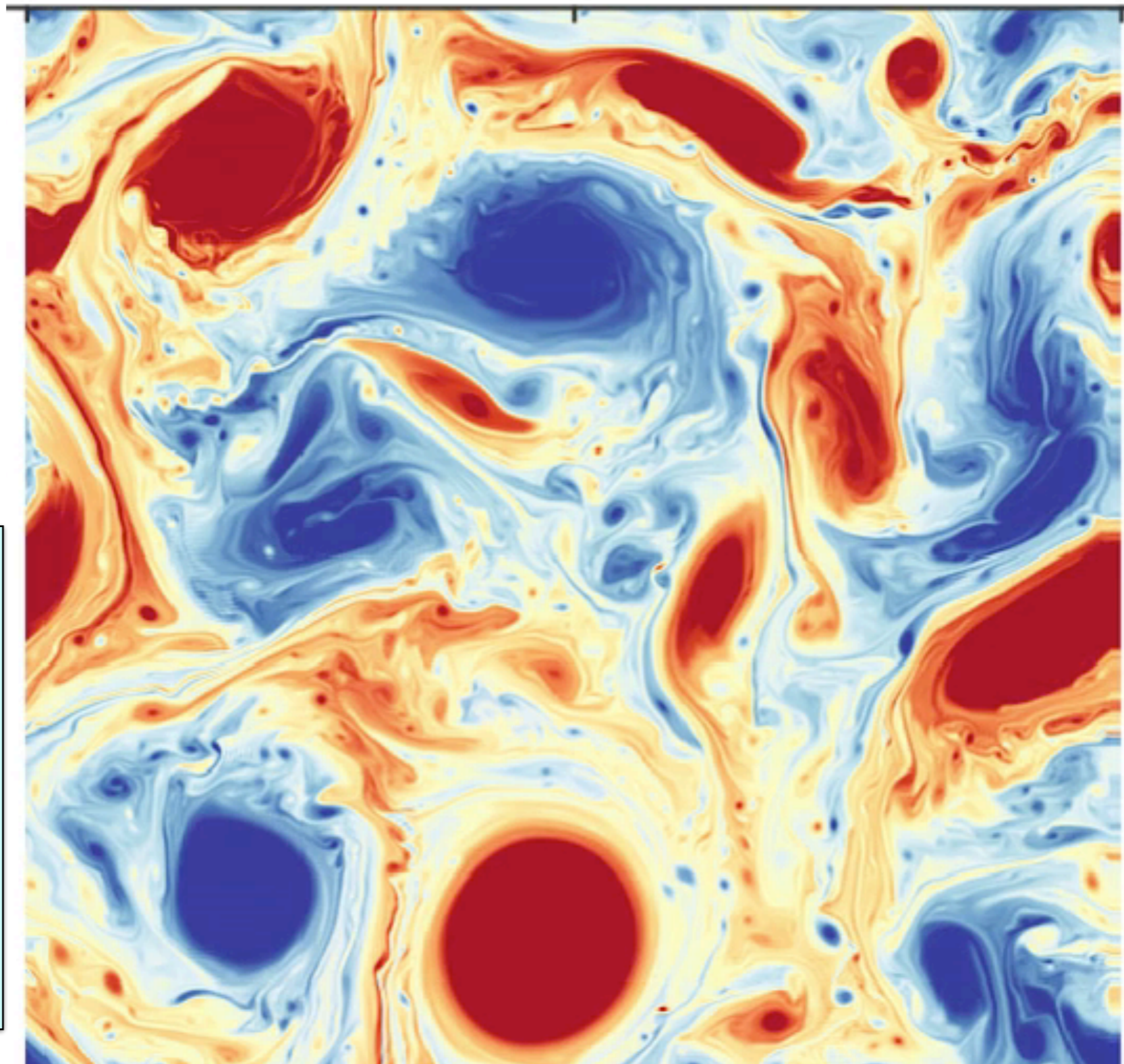
$$u = \left( \text{cst.} \nabla^\perp \Delta^{-1/2} \right) b$$

Reference flow:

deterministic

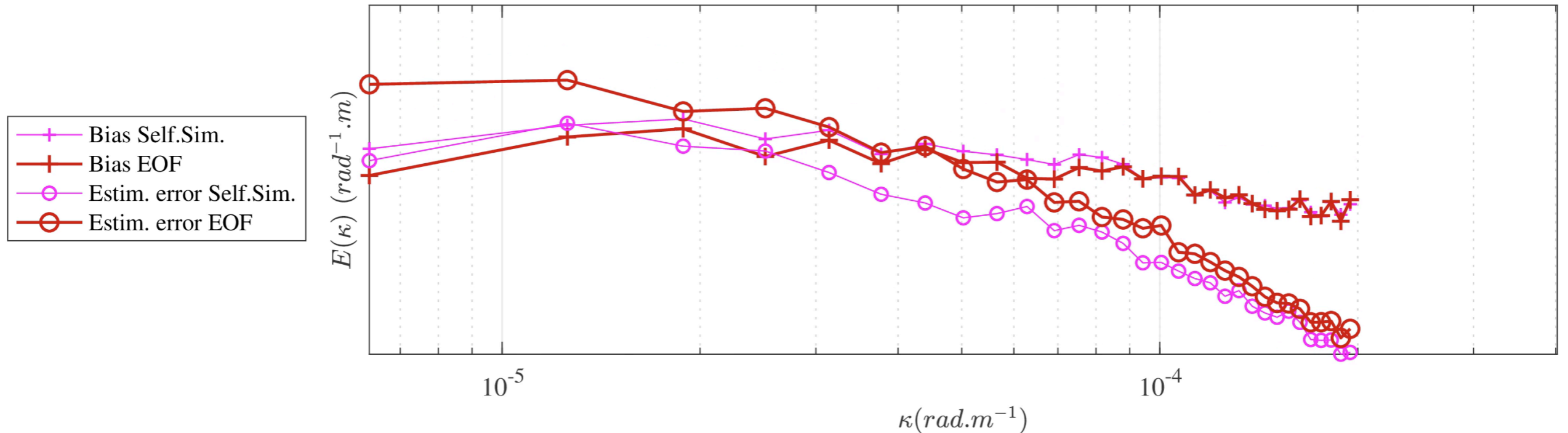
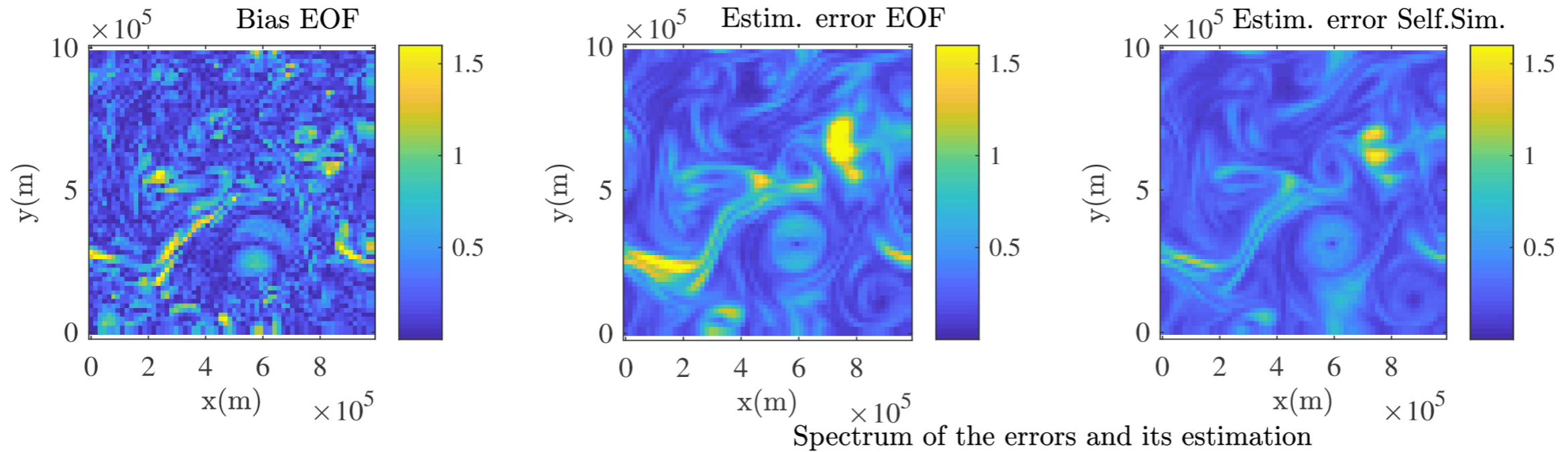
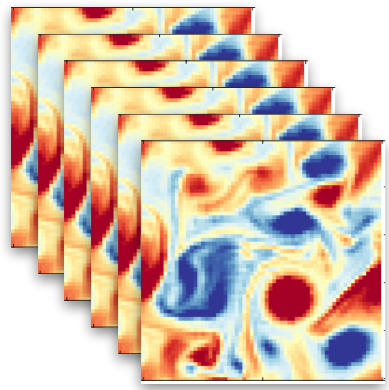
SQG

512 x 512

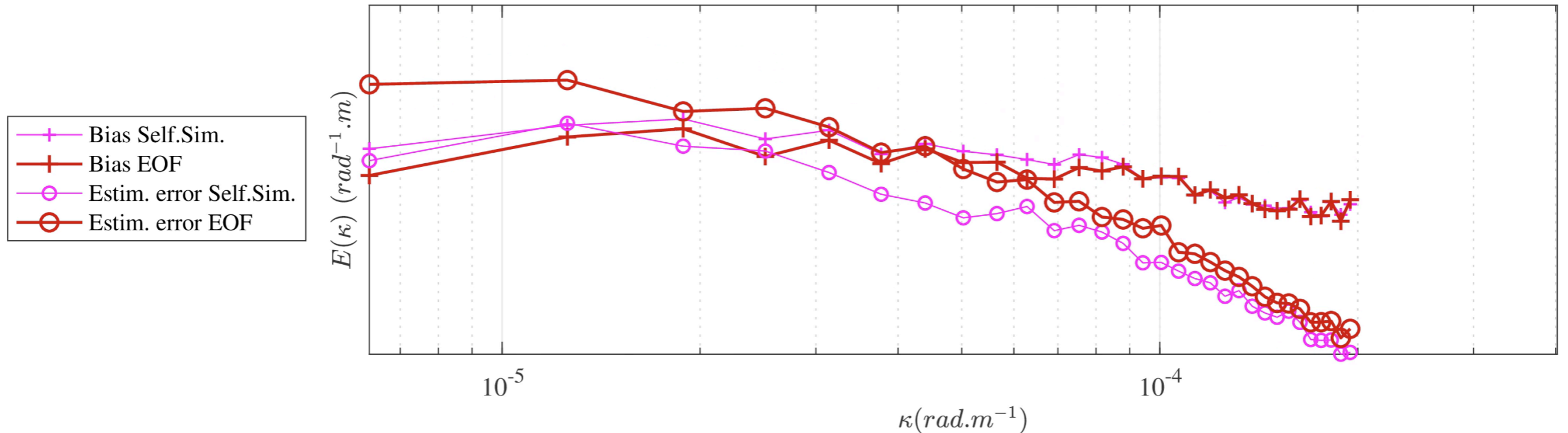
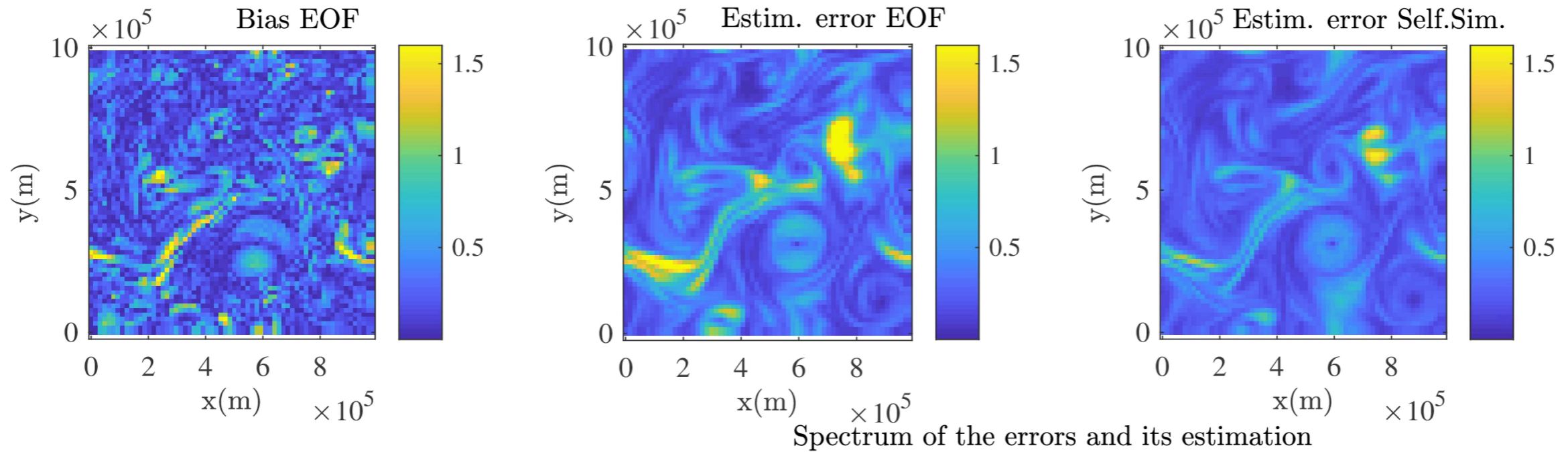
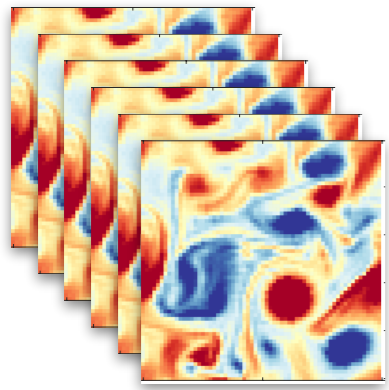




# Ensemble : uncertainty quantification



# Ensemble : uncertainty quantification



# Part IV

Unresolved velocity  
non-stationary heterogeneity

$t = 0$  day

## SQG

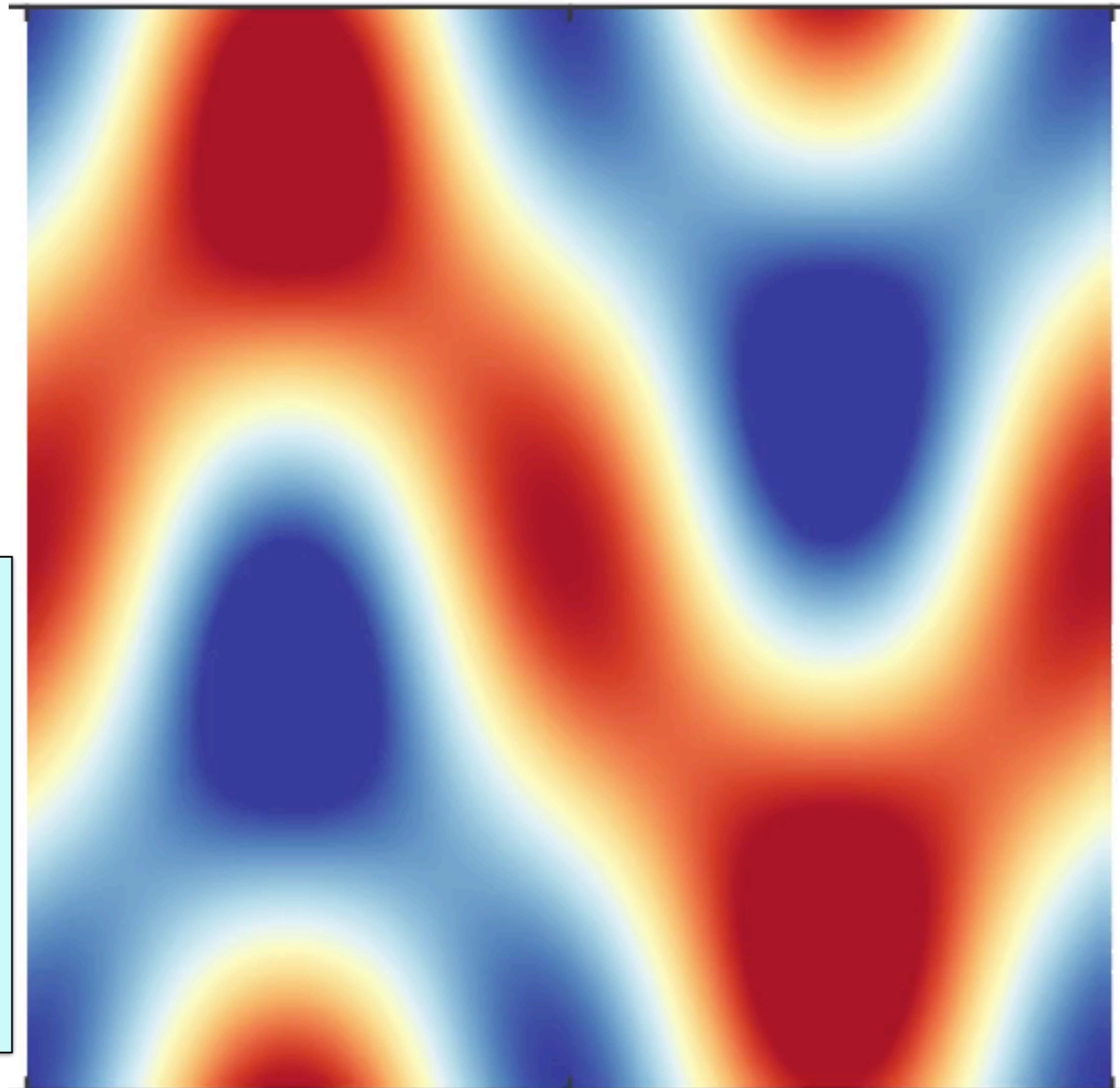
$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-} \\ \text{viscosity}$$

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

Reference flow:

deterministic SQG  
1024 x 1024

Constantin et al.  
(1994, 1999, 2012)



$t = 0$  day

## SQG

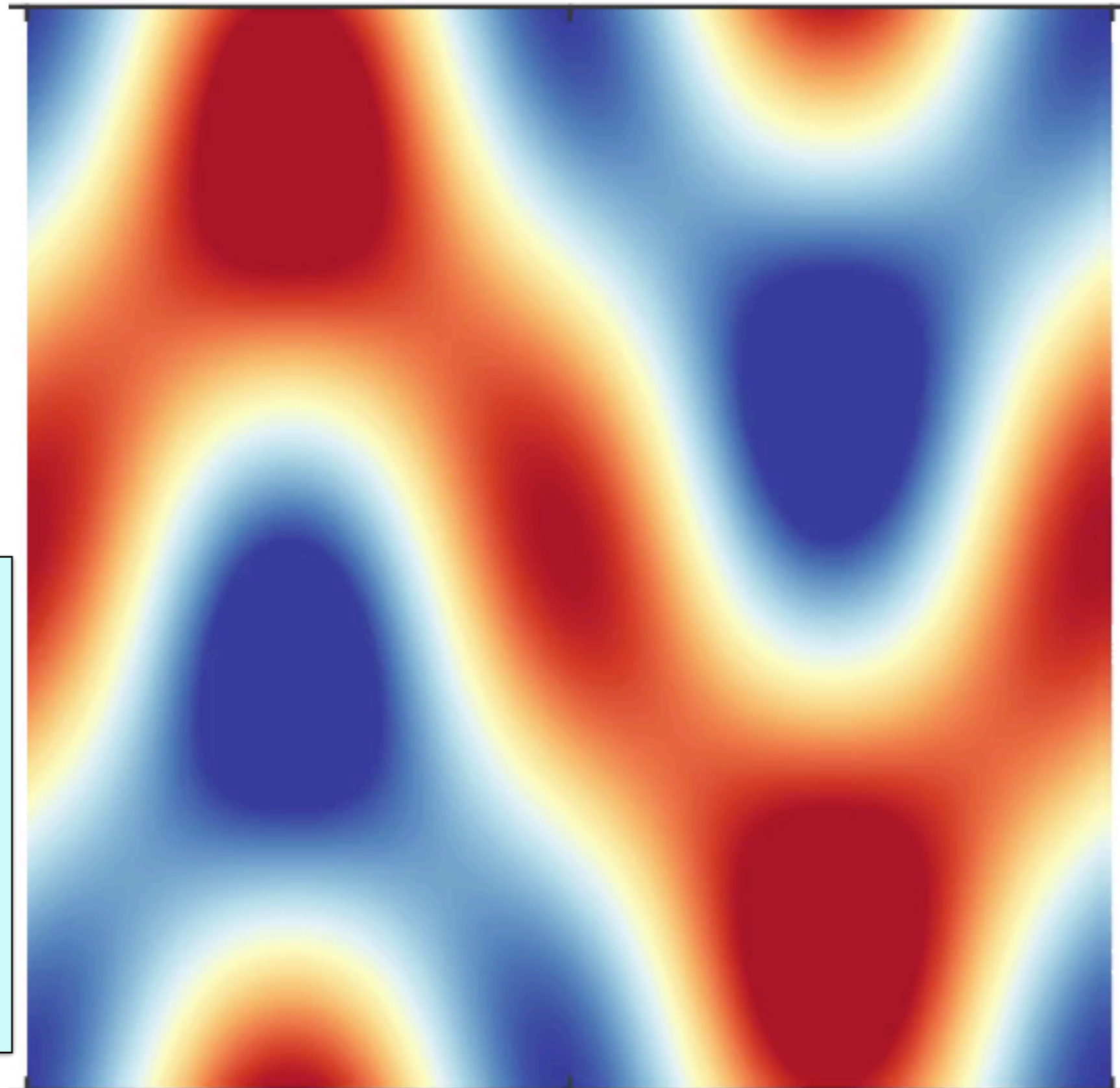
$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

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

Reference flow:

deterministic SQG  
1024 x 1024

Constantin et al.  
(1994, 1999, 2012)



$t = 15 \text{ day}$

## SQG

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

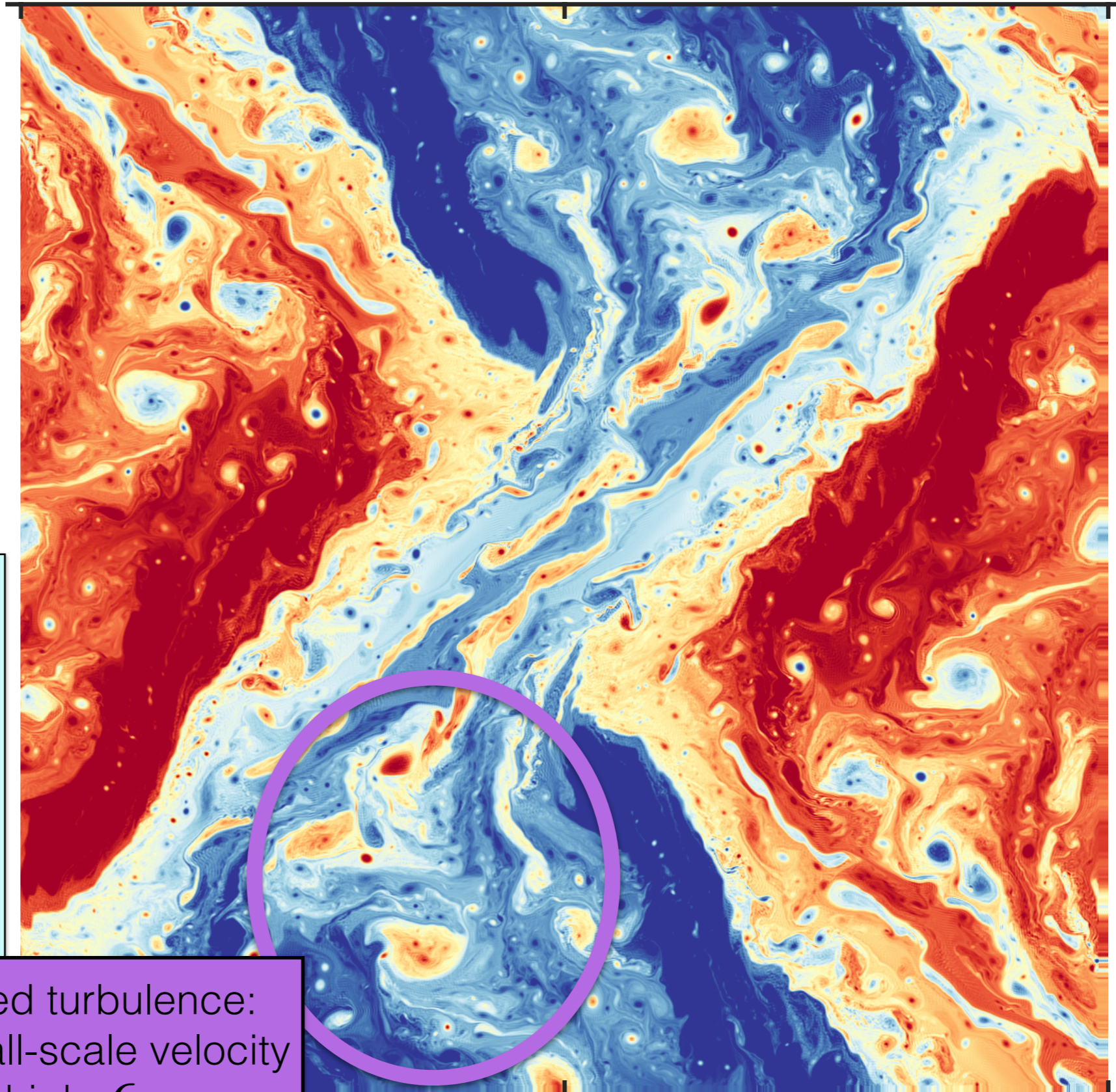
$$\mathbf{u} = \left( \text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic SQG  
1024 x 1024

Constantin et al.  
(1994, 1999)

Developed turbulence:  
strong small-scale velocity  
high  $\epsilon$



$t = 15$  day

## SQG

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

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

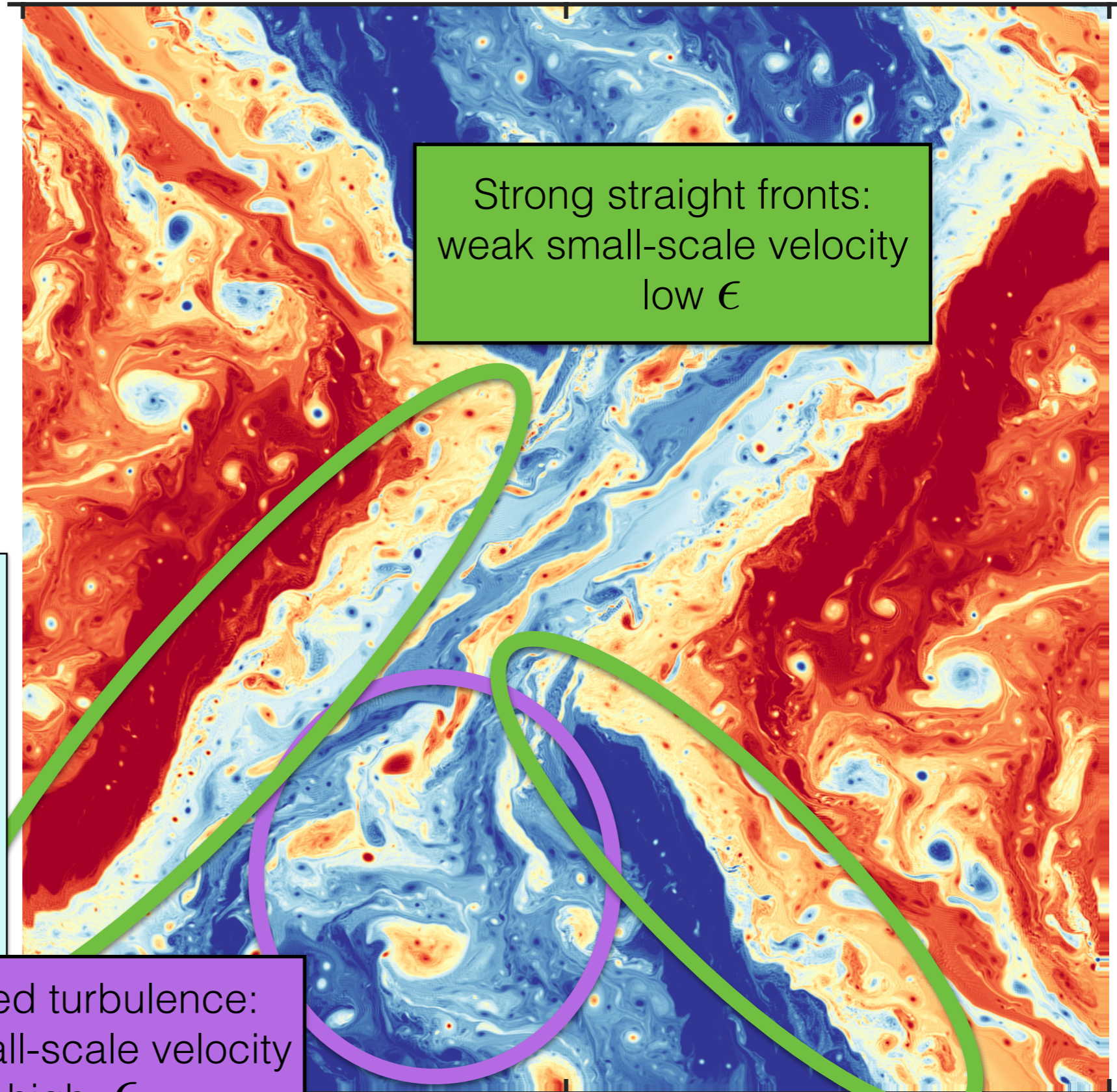
Reference flow:

deterministic SQG  
1024 x 1024

Constantin et al.  
(1994, 1999)

Developed turbulence:  
strong small-scale velocity  
high  $\epsilon$

Strong straight fronts:  
weak small-scale velocity  
low  $\epsilon$



Large scales:

$w$

Small scales:

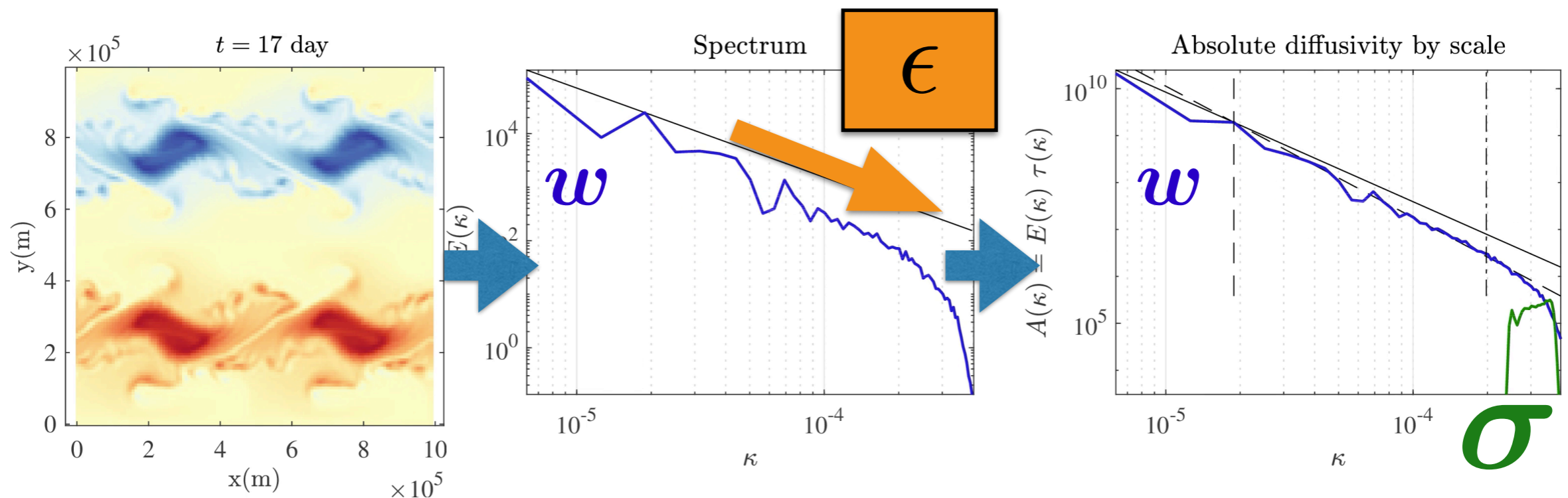
$\sigma \dot{B}$

Variance tensor:

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

# Heterogeneous modulation of $\sigma$

(for Resseguier et al. 2017b)





Large scales:

$w$

Small scales:

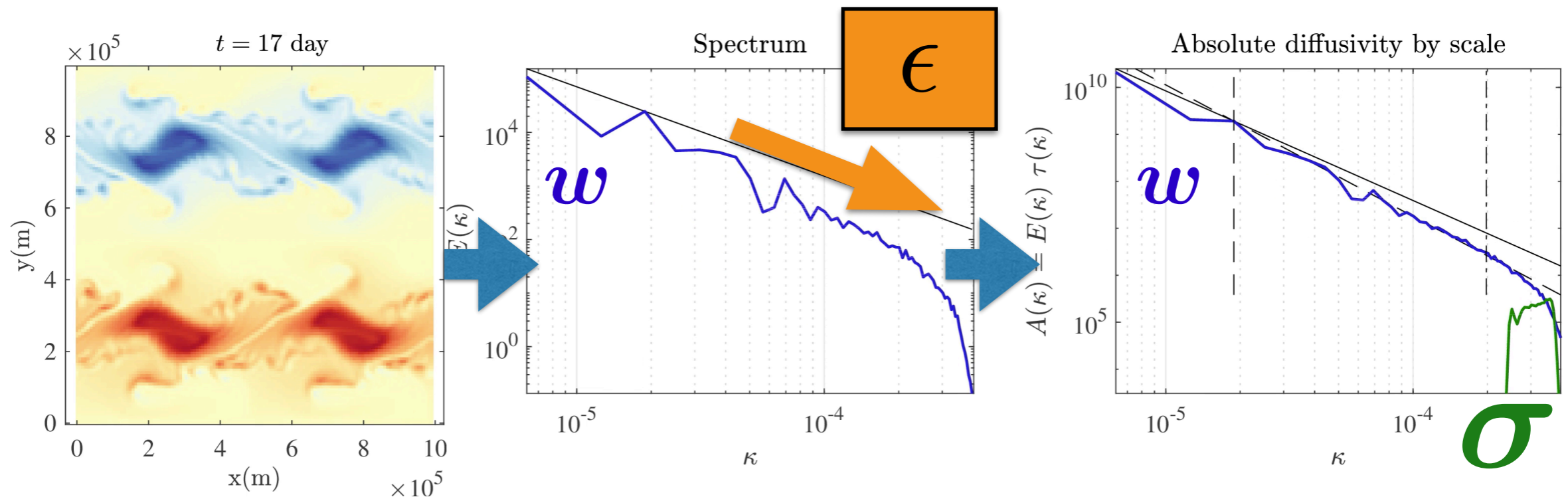
$\sigma \dot{B}$

Variance tensor:

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

# Heterogeneous modulation of $\sigma$

(for Resseguier et al. 2017b)



Large scales:

$w$

Small scales:

$\sigma \dot{B}$

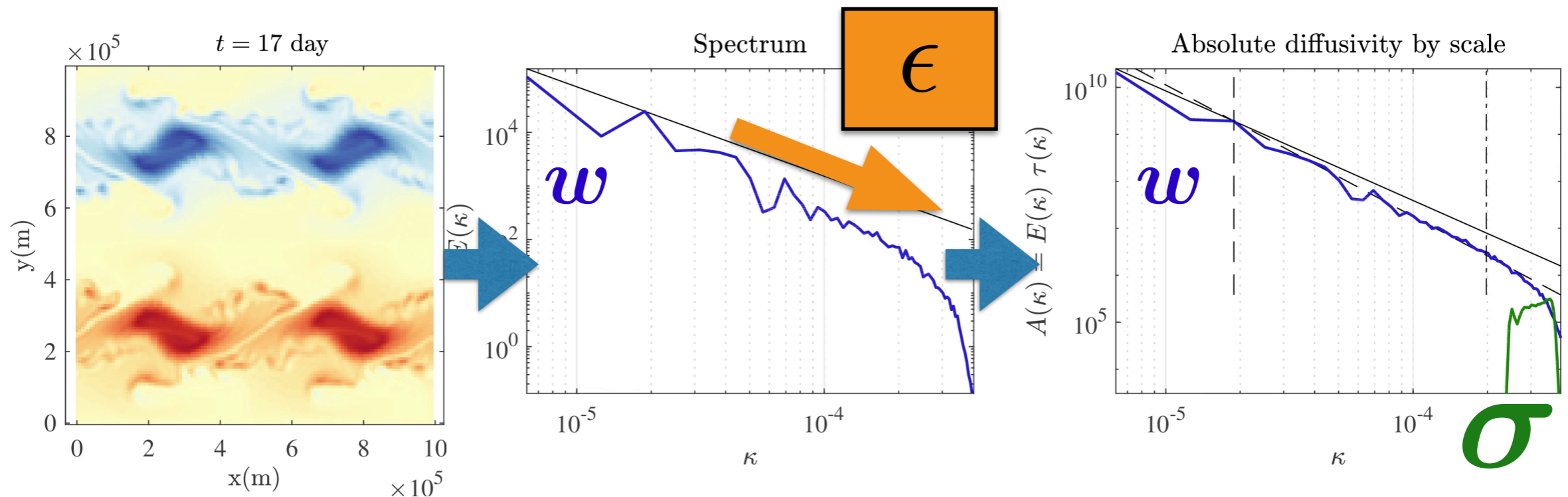
Variance tensor:

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

# Heterogeneous modulation of $\sigma$

(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:

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

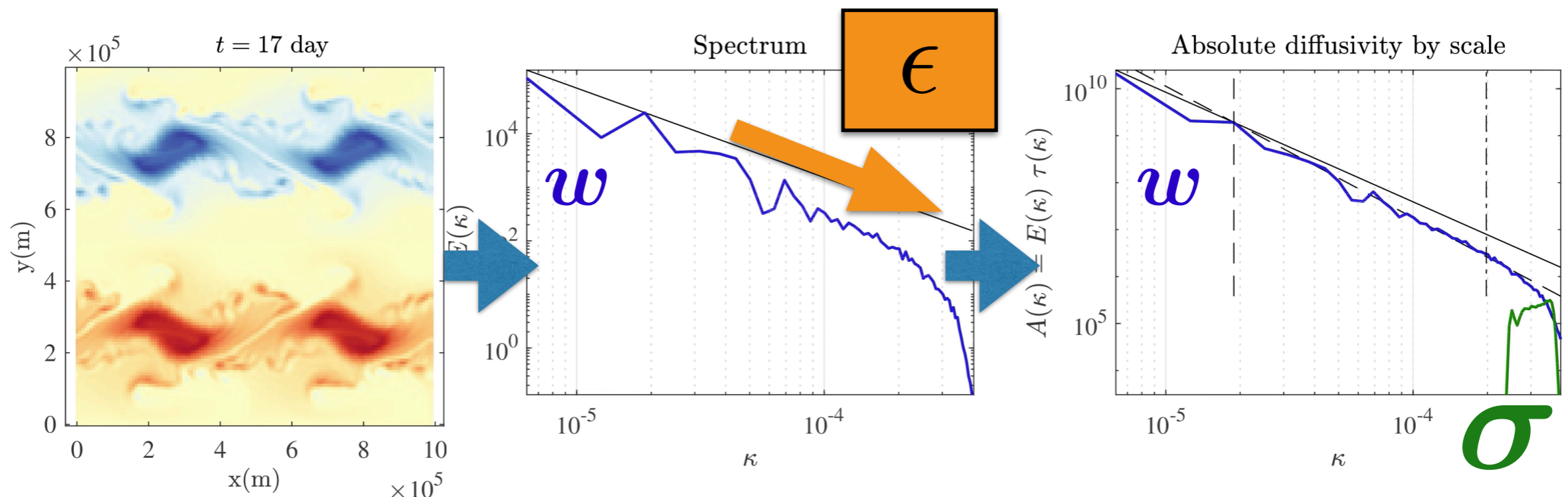
# Heterogeneous

# modulation of $\sigma$

(for Resseguier et al. 2017b)

Dynamics  
(1/3 for SQG)

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



# Heterogeneous modulation of $\sigma$

(for Resseguier et al. 2017b)

Large scales:  
 $w$

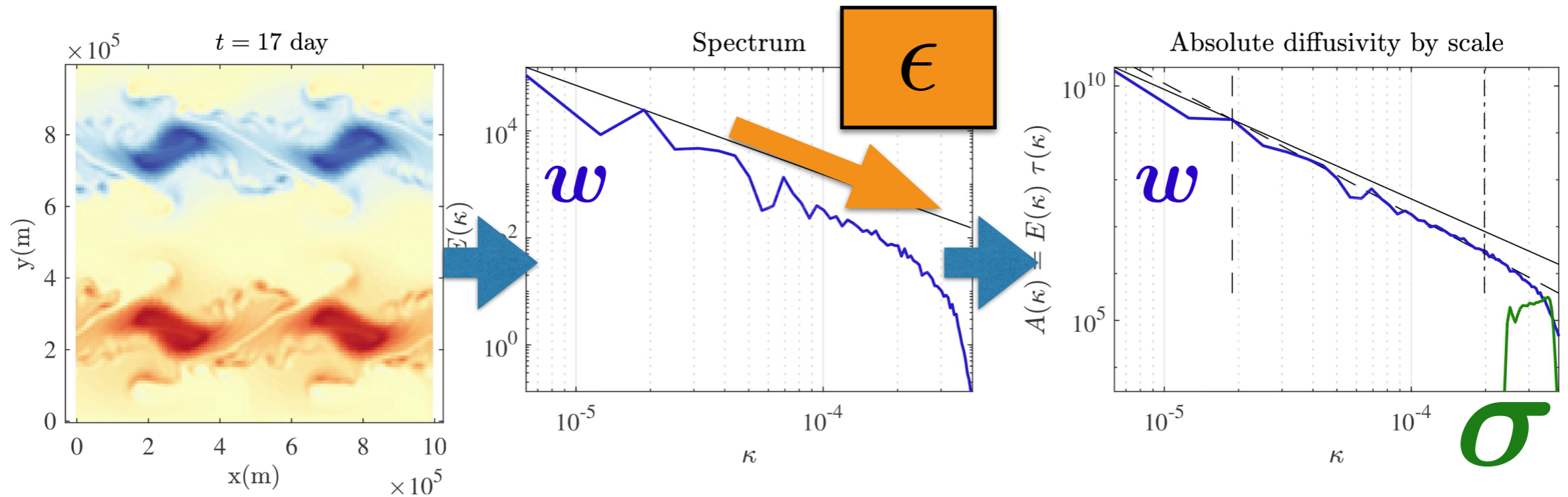
Small scales:  
 $\sigma \dot{B}$

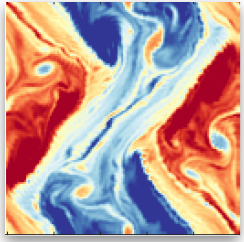
Variance tensor:  
 $a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$

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

Dynamics  
(1/3 for SQG)

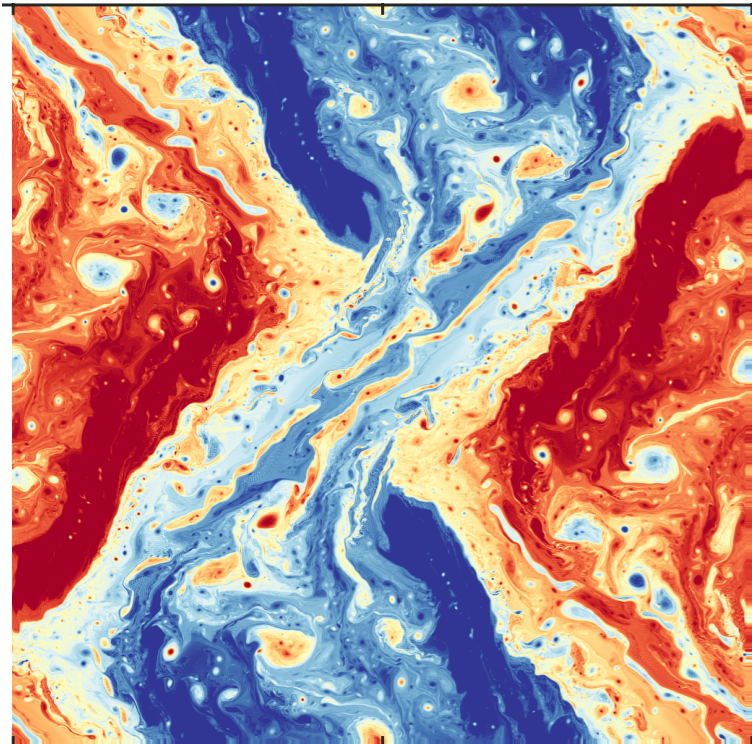
Learned on large scales at each time  $t$



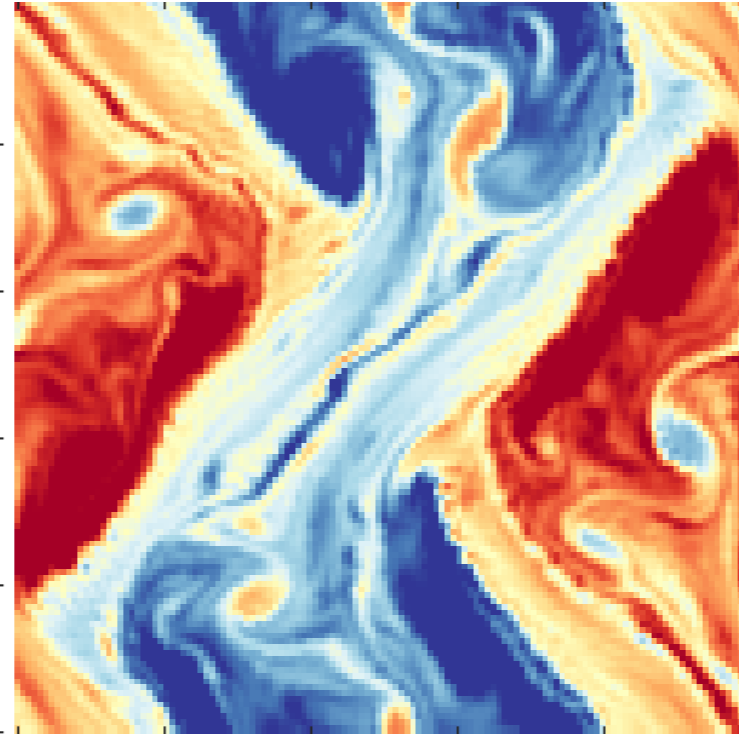


# Heterogenous modulation

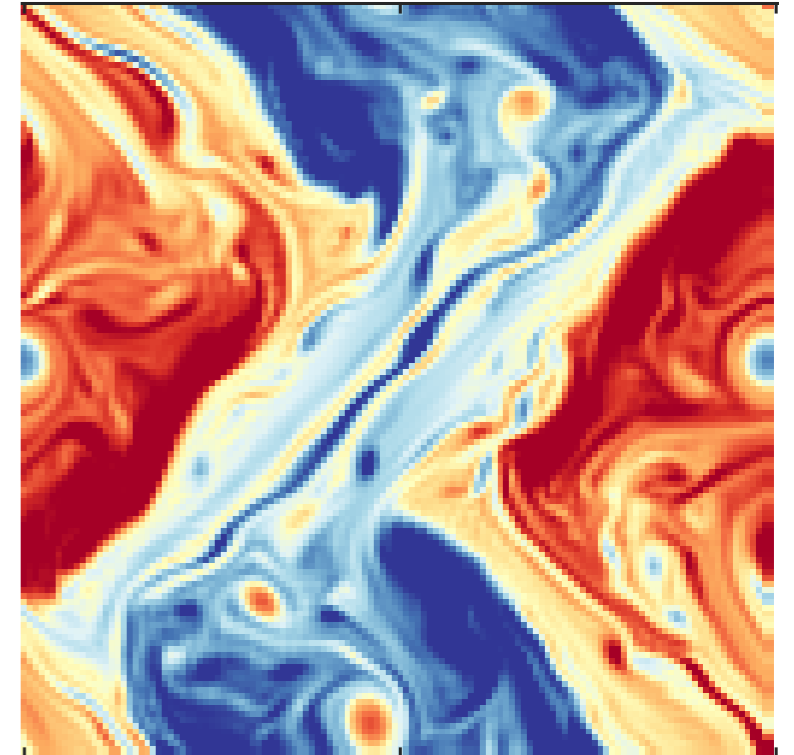
Deterministic 1024 x 1024

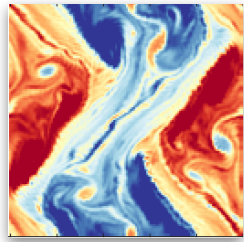


Stochastic 128 x128  
with Homogeneous  
small-scale velocity



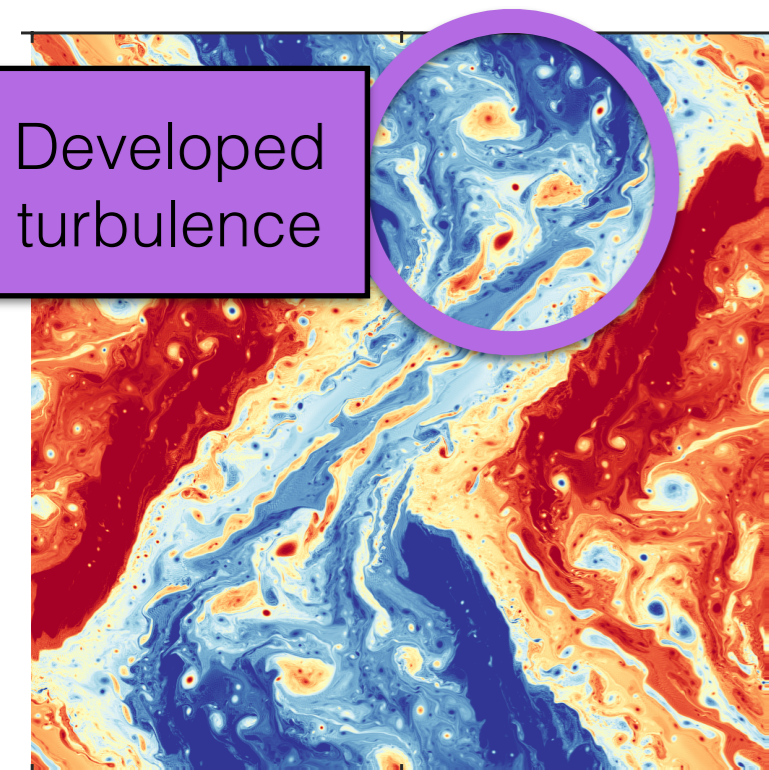
Stochastic 128 x128  
with Energy-flux-based  
modulation



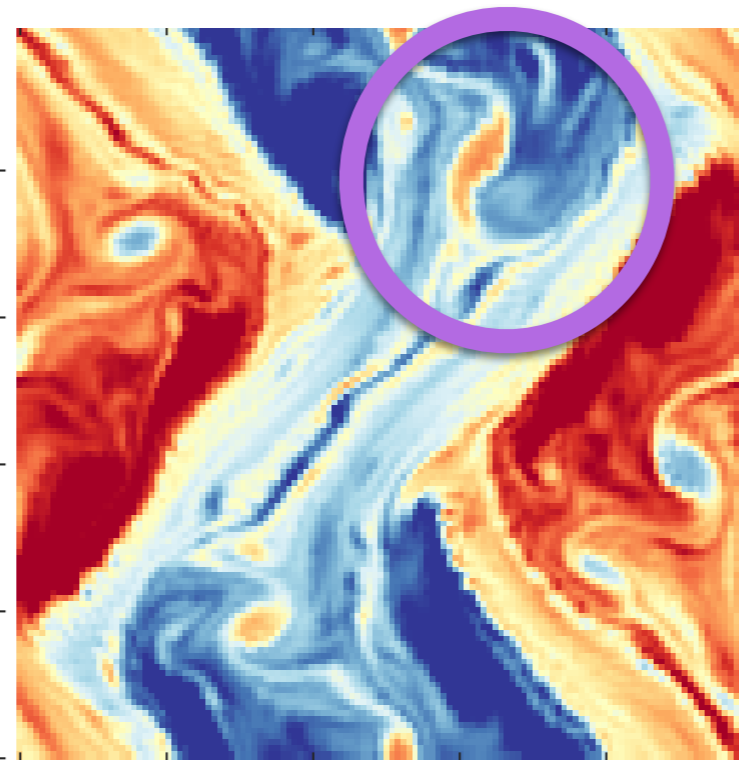


# Heterogenous modulation

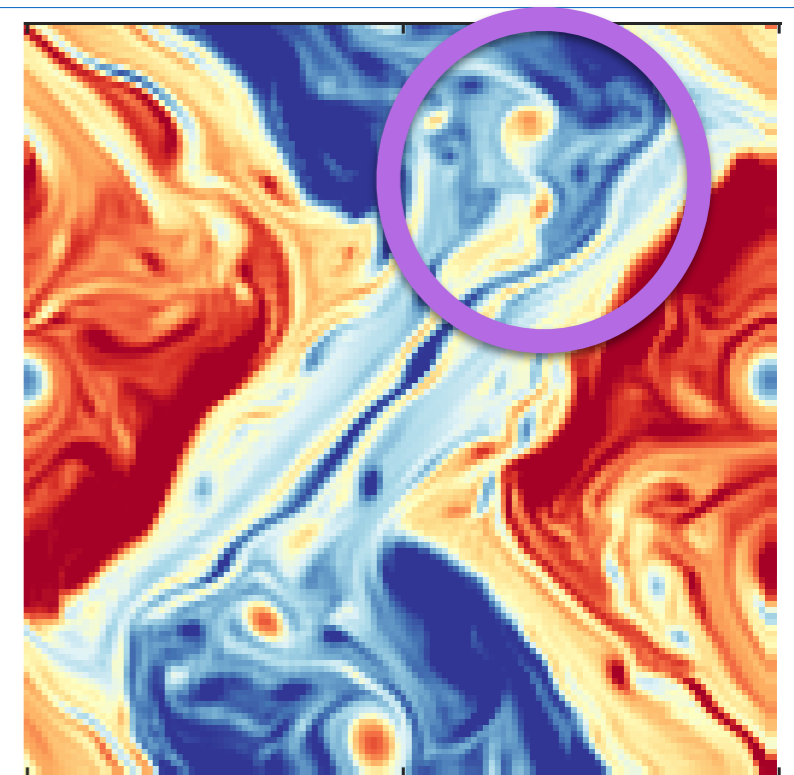
Deterministic 1024 x 1024

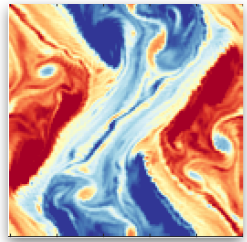


Stochastic 128 x 128  
with Homogeneous  
small-scale velocity



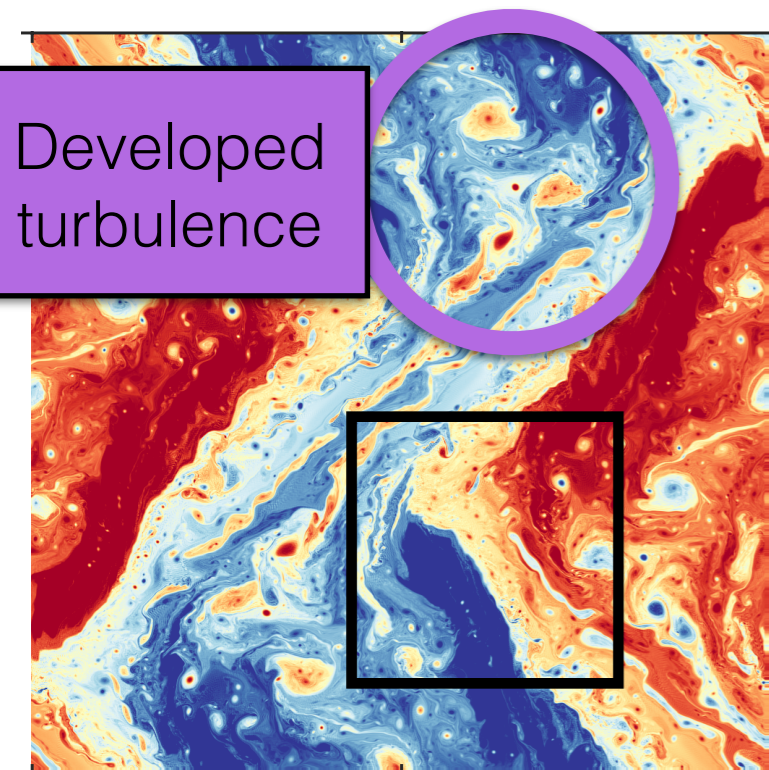
Stochastic 128 x 128  
with Energy-flux-based  
modulation



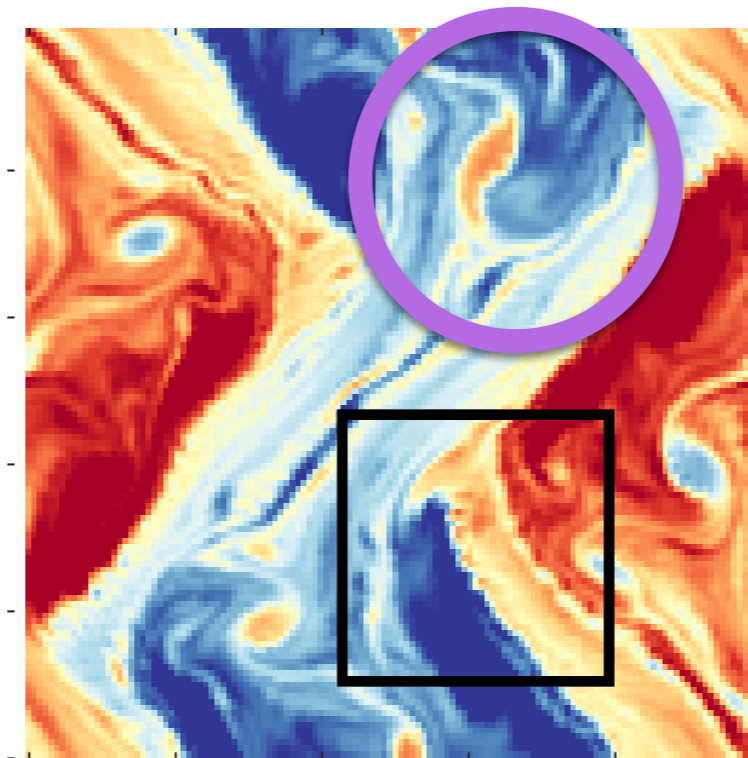


# Heterogenous modulation

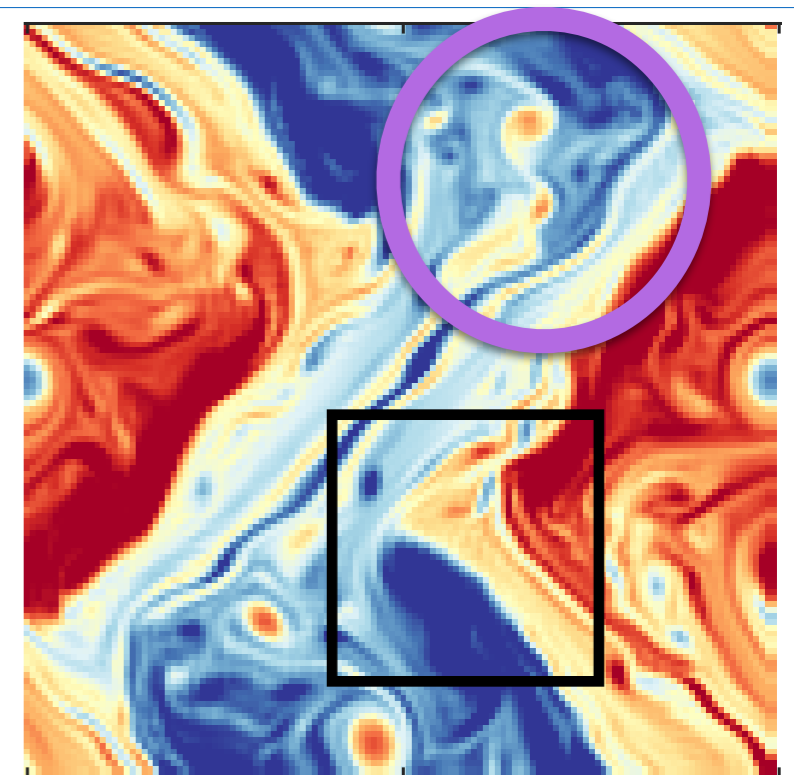
Deterministic 1024 x 1024

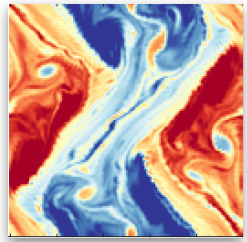


Stochastic 128 x 128  
with Homogeneous  
small-scale velocity



Stochastic 128 x 128  
with Energy-flux-based  
modulation





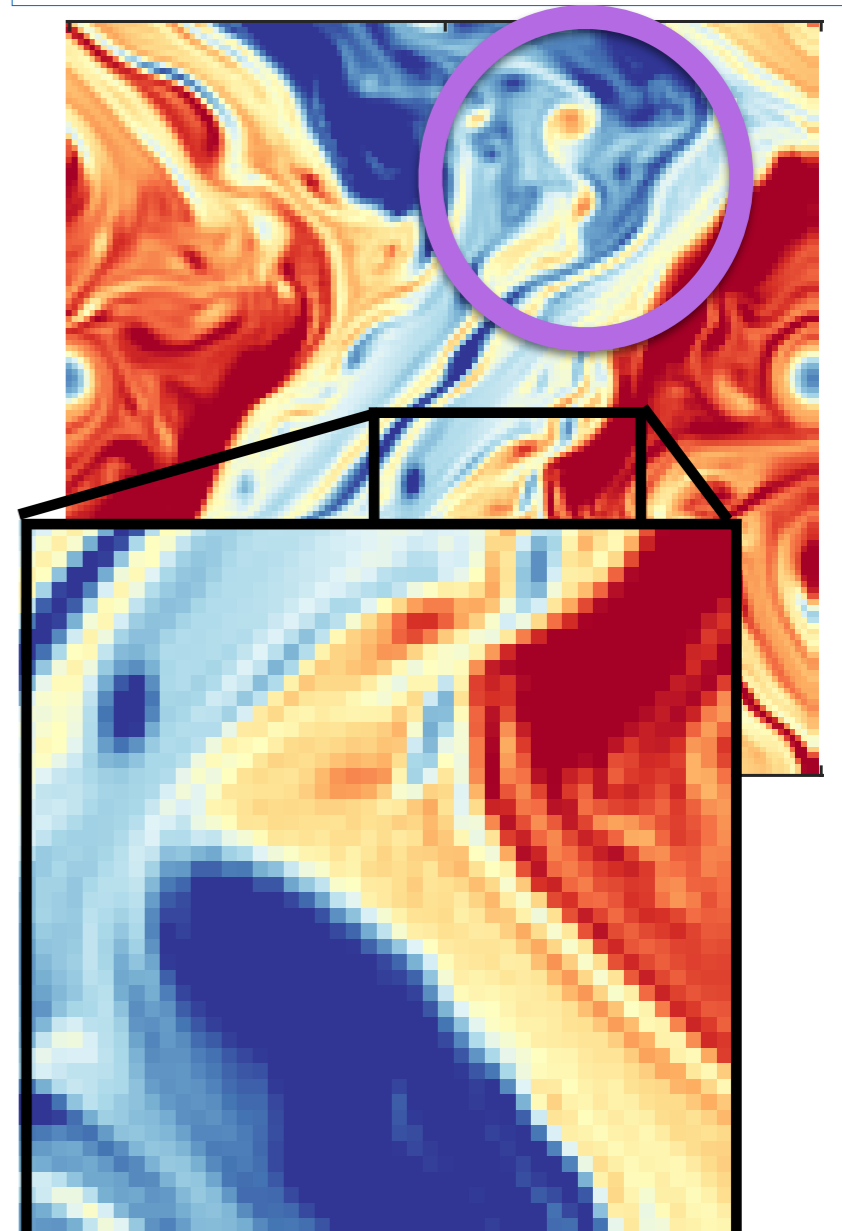
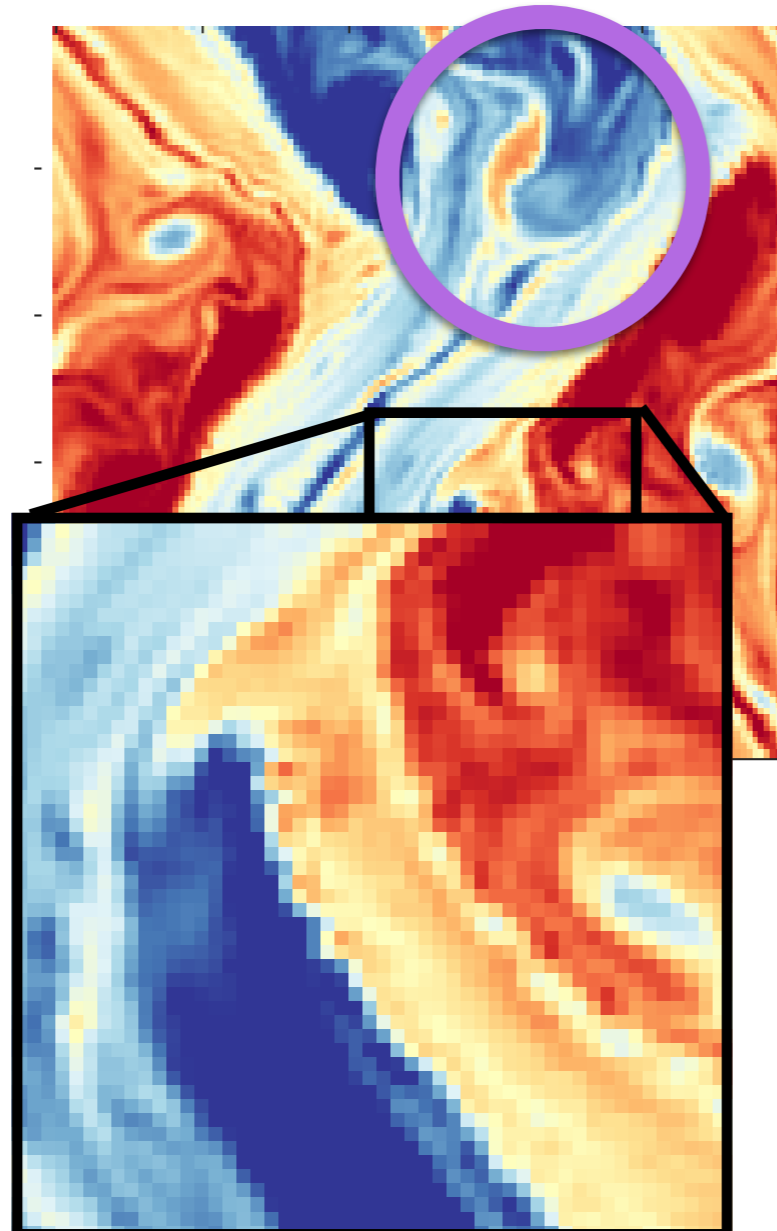
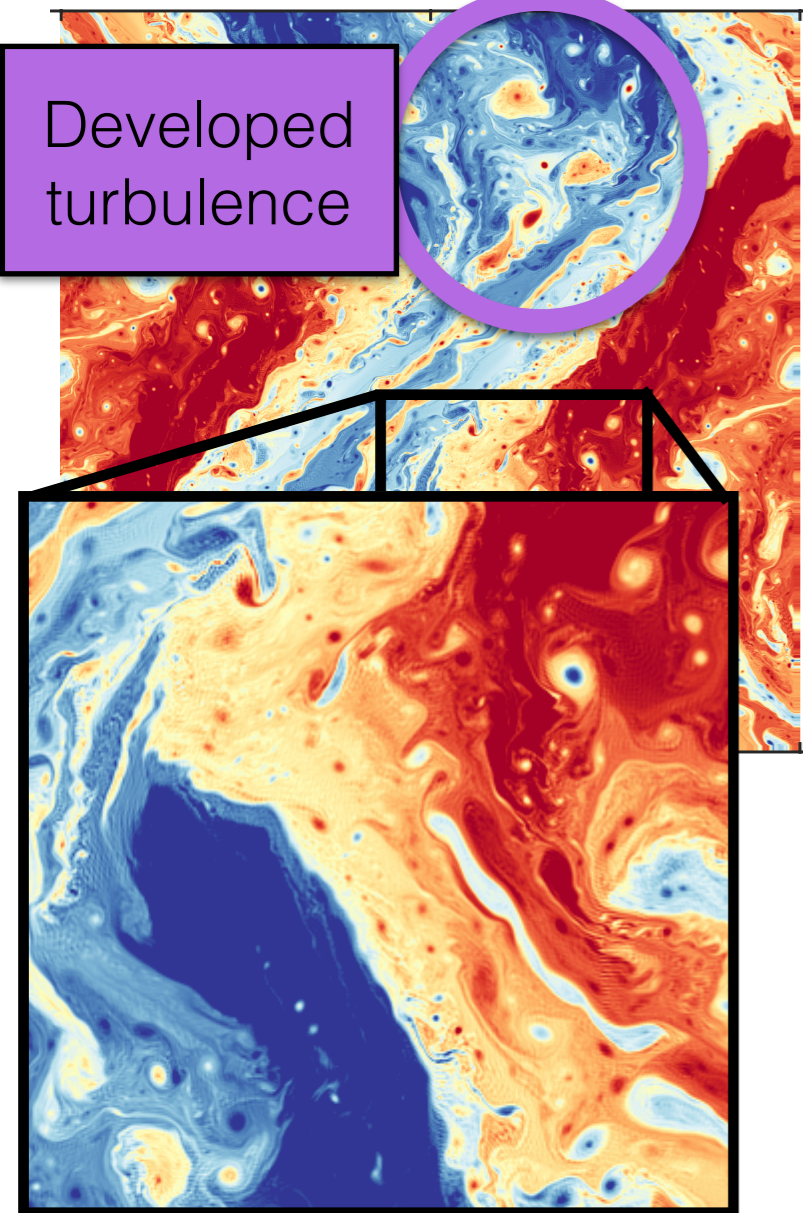
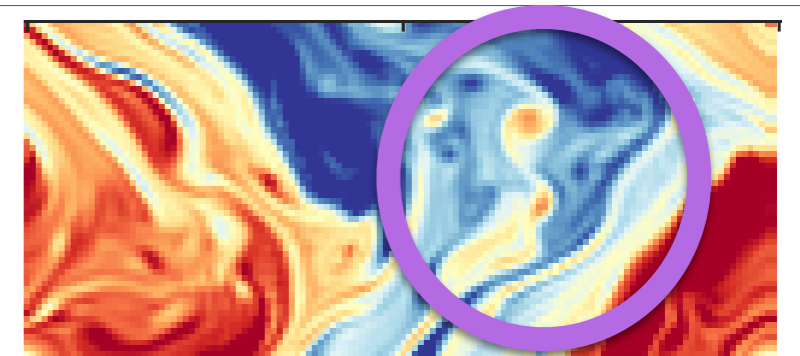
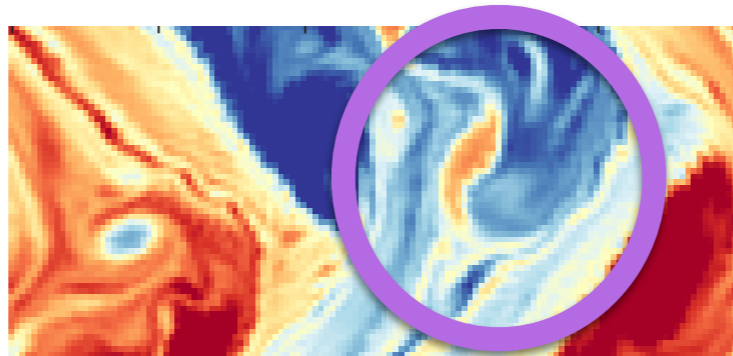
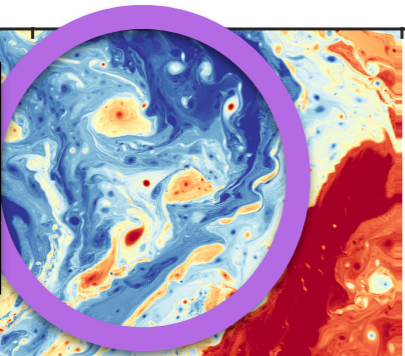
# Heterogenous modulation

Deterministic 1024 x 1024

Stochastic 128 x 128  
with Homogeneous  
small-scale velocity

Stochastic 128 x 128  
with Energy-flux-based  
modulation

Developed  
turbulence





# 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.