Structure Functions as a Tool for Measuring the Mesoscale

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Mesoscale Turbulence

• The dynamics of the ocean are heterogeneous and spatially intermittent, even at small scales



Pearson & Fox-Kemper Physical Review Letters (2018)



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- Mesoscale eddy forced at deformation radius
- Eddy breaks into child eddies conserving properties

Adapted from Meneveau & Sreenivasan, JFM (1991)



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Child eddies

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Child eddies take fraction β of initial property

This cascade to smaller scales repeats

- After *n* cascade steps then

 $c \neq n = \varepsilon \neq 0 \qquad log(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow n / \varepsilon \downarrow 0) = \sum_{i=1}^{i=1} 1 n \text{ log}(\varepsilon \downarrow$

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Spatial Distribution of Turbulence Statistics

• KE dissipation in Parallel Ocean Program (POP) 1/10*îo* model



Pearson & Fox-Kemper, Physical Review Letters (2018)

Spatial Distribution of Turbulence Statistics



Pearson & Fox-Kemper, Physical Review Letters (2018)

Spatial Distribution of Turbulence Statistics

- Modelled mesoscale turbulence has log-normal statistics
- How can we observe and diagnose the spatial variations in mesoscale spectral fluxes?



Pearson & Fox-Kemper, *Physical Review Letters* (2018)

• Structure functions depend on spatial differences in variables

 $\delta \phi = \phi(\mathbf{x} + \mathbf{r}) - \phi(\mathbf{x})$

• Exact laws relate structure functions to spectral fluxes in inertial cascades of isotropic, homogeneous turbulence

Spectral KE Flux=
$$\varepsilon$$
=-5/4 · $\delta u \downarrow L \delta u \downarrow L \delta u \downarrow L /r$

Kolmogorov (1941)

3D turbulence:



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Lindborg (2007)

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Anisotropic Structure Functions



• Requires complete prior knowledge of anisotropy to quantify $\mathcal{E} \downarrow q$

New Anisotropic Structure Functions

 $\varepsilon \downarrow q = -1/2 \, \delta q \, \delta A \downarrow q$ where $A \downarrow q = -\mathbf{u} \downarrow g \cdot \nabla q$

New statistic has the following benefits over $\varepsilon \downarrow q = -1/4 \ \nabla \cdot (\delta u \delta q \delta q)$:

- 1. Does not require integration
- 2. Can be evaluated without prior quantification of anisotropy
- 3. Has two, rather than three, differences (potential convergence benefits)

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However, it does require knowledge of local flow gradients (amenable to numerical models, and future data [i.e. SWOT])

Extension to other dynamical regimes

 $\varepsilon \downarrow q = -1/2 \ \delta q \delta A \downarrow q$ in the QG enstrophy cascade

Analogous relations for diagnosing spectral fluxes in the;

- 1. QG inverse energy cascade
- 2. Surface QG cascades
- 3. Two-dimensional cascades

Pearson, Pearson & Fox-Kemper In prep. (GRL)

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In isotropic limit, new and old structure functions converge

New comparable to old under anisotropic conditions



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- Log-normality allows us to accurately quantify observational data and develop more precise turbulence parameterizations
- New structure functions could diagnose spectral fluxes from patchy or irregular data
- New structure functions for **anisotropic QG**, **SQG**, and **2D** inertial cascades

Anisotropic Structure Functions

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Anisotropic QG:



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