
Large-scale Eigenfunctions and Mixing

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with

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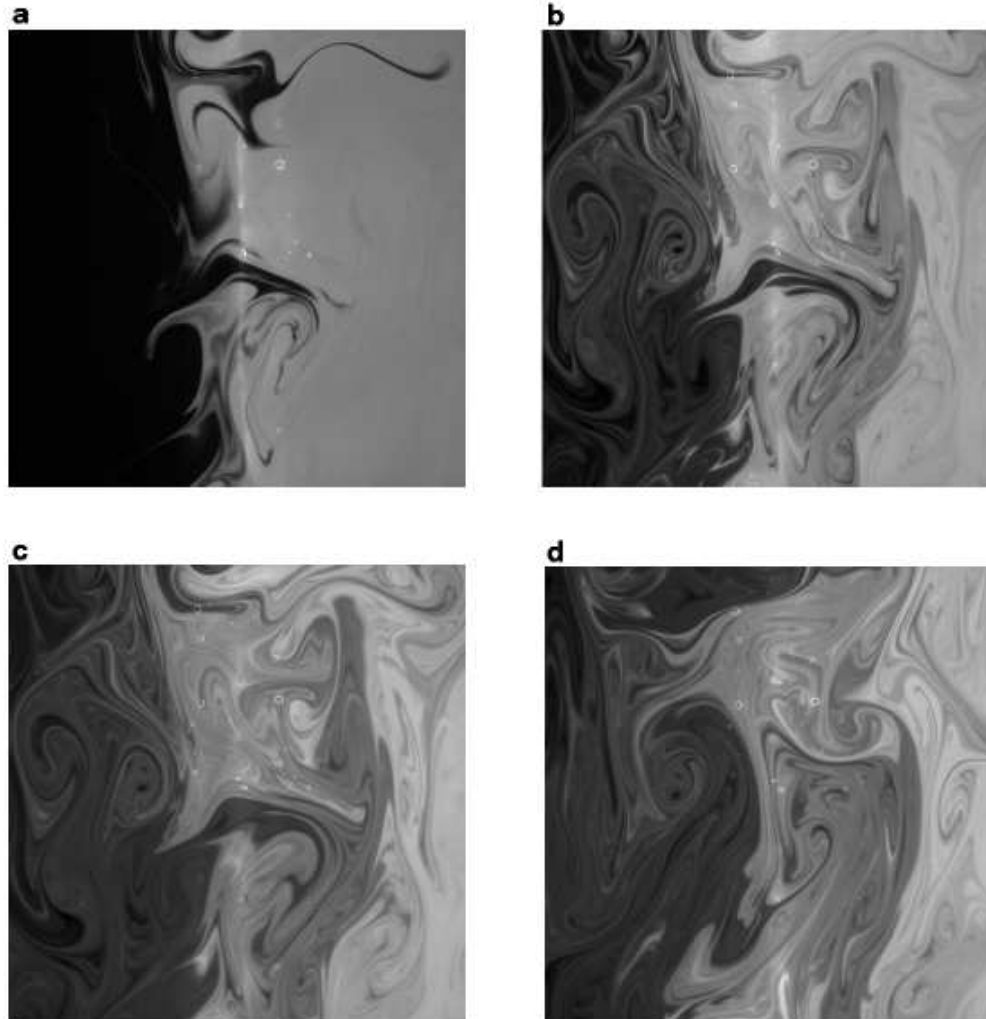
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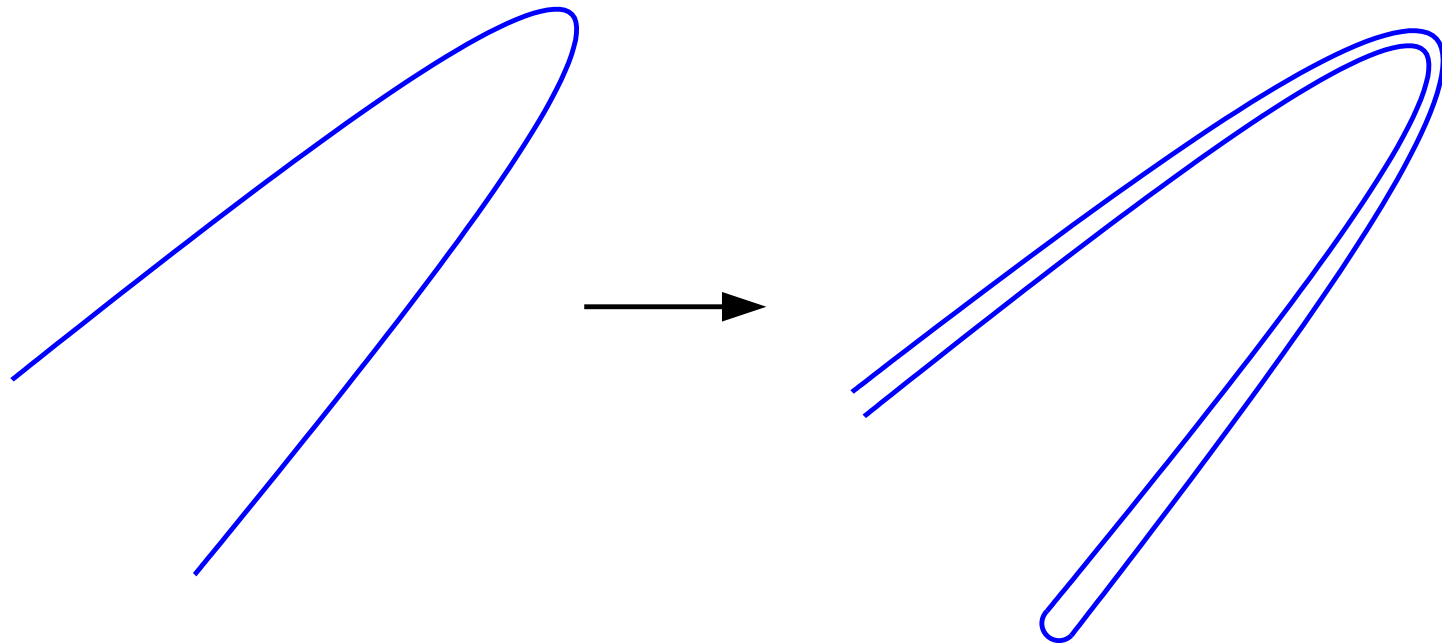
Persistent Pattern

Disordered array ($i = 2, 20, 50, 50.5$)



[Rothstein, Henry, and Gollub, Nature **401**, 770 (1999)]

Evolution of Pattern



- “Striations”
- Smoothed by diffusion
- Eventually settles into “pattern” (eigenfunction)

Local vs Global Regimes of Mixing

Local theory:

- Based on distribution of Lyapunov exponents.

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Average over angles

Statistical model

Statistical model

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Global theory:

- Eigenfunction of advection–diffusion operator.

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Global theory:

- Eigenfunction of advection–diffusion operator.
- [Pierrehumbert, Chaos Sol. Frac. (1994)] Strange eigenmode
- [Fereday et al., Wonhas and Vassilicos, PRE (2002)] Baker's map
- [Sukhatme and Pierrehumbert, PRE (2002)]
- [Fereday and Haynes (2003)] Unified description
- [Thiffeault and Childress (2003)] Modified Cat map

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Global theory:

- Eigenfunction of advection–diffusion operator.

- Map allows analytical results.

The Map

We consider a diffeomorphism of the 2-torus $\mathbb{T}^2 = [0, 1]^2$,

$$\mathcal{M}(\mathbf{x}) = \mathbb{M} \cdot \mathbf{x} + \phi(\mathbf{x}),$$

where

$$\mathbb{M} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}; \quad \phi(\mathbf{x}) = \frac{K}{2\pi} \begin{pmatrix} \sin 2\pi x_1 \\ \sin 2\pi x_1 \end{pmatrix};$$

$\mathbb{M} \cdot \mathbf{x}$ is the **Arnold cat map**.

The map \mathcal{M} is **area-preserving** and **chaotic**.

For $K = 0$ the stretching of fluid elements is **homogeneous in space**.

Advection and Diffusion

Iterate the map and apply the **heat operator** to a scalar field (which we call **temperature** for concreteness) distribution $\theta^{(i-1)}(\mathbf{x})$,

$$\theta^{(i)}(\mathbf{x}) = \mathcal{H}_\epsilon \theta^{(i-1)}(\mathcal{M}^{-1}(\mathbf{x}))$$

where ϵ is the **diffusivity**, with the **heat operator** \mathcal{H}_ϵ . In other words: **advect** instantaneously and then **diffuse** for one unit of time.

After Fourier expanding $\theta^{(i)}(\mathbf{x})$, the effect of advection and diffusion becomes

$$\hat{\theta}^{(i)}(\mathbf{x}) = \sum_{\mathbf{q}} \mathbb{T}_{\mathbf{k}\mathbf{q}} \hat{\theta}_{\mathbf{q}}^{(i-1)},$$

with the **transfer matrix**,

$$\mathbb{T}_{\mathbf{k}\mathbf{q}} = e^{-\epsilon q^2} \delta_{0, Q_2} i^{Q_1} J_{Q_1}((k_1 + k_2) K), \quad \mathbf{Q} := \mathbf{k} \cdot \mathbb{M} - \mathbf{q}.$$

Variance: A measure of mixing

In the absence of diffusion ($\epsilon = 0$), the **variance** $\sigma^{(i)}$

$$\sigma^{(i)} := \int_{\mathbb{T}^2} |\theta^{(i)}(\mathbf{x})|^2 d\mathbf{x} = \sum_{\mathbf{k}} \sigma_{\mathbf{k}}^{(i)}, \quad \sigma_{\mathbf{k}}^{(i)} := |\hat{\theta}_{\mathbf{k}}^{(i)}|^2$$

is **preserved**. (We assume the spatial mean of θ is zero.)

For $\epsilon > 0$ the variance **decays**.

We consider the case $\epsilon \ll 1$, of greatest practical interest.

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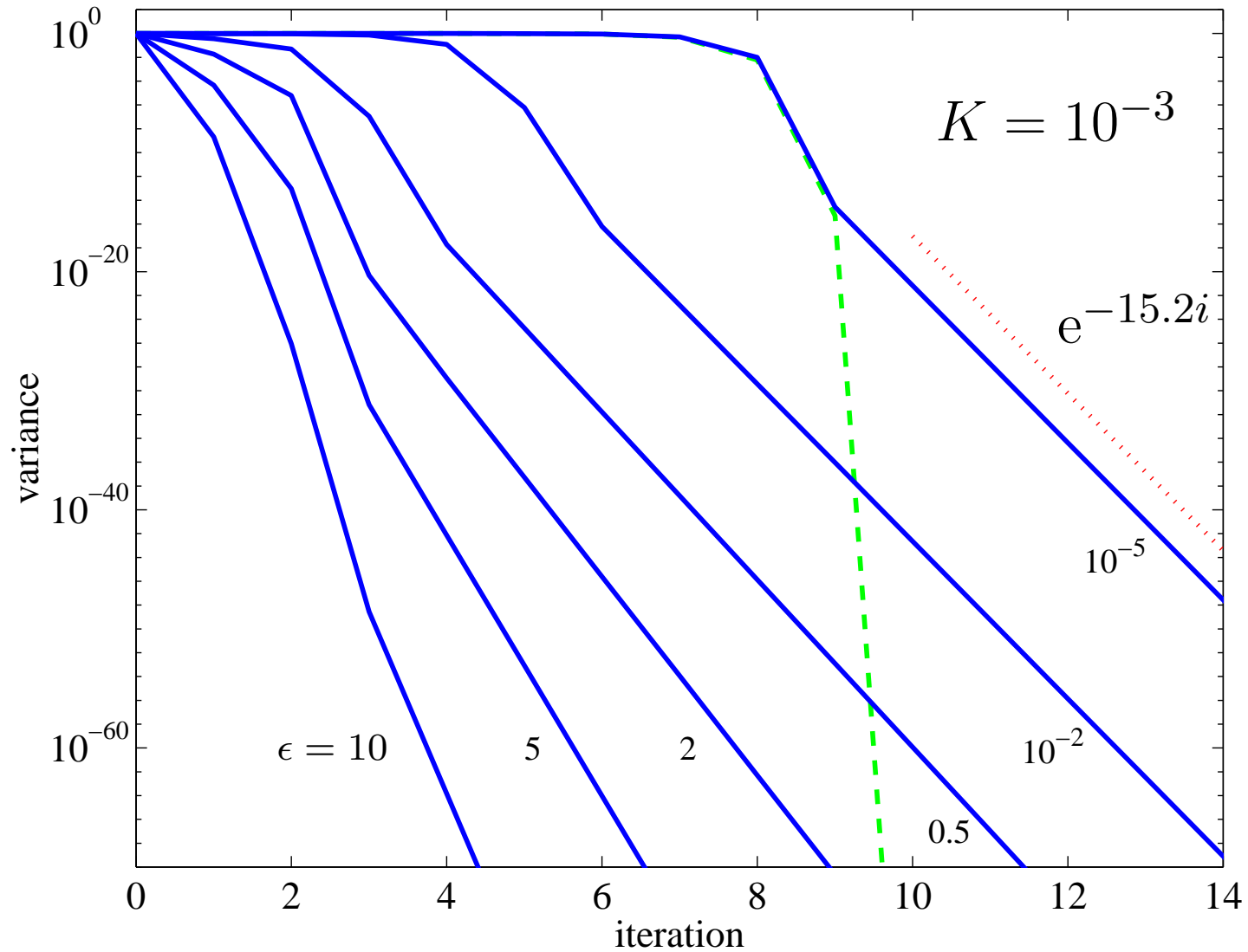
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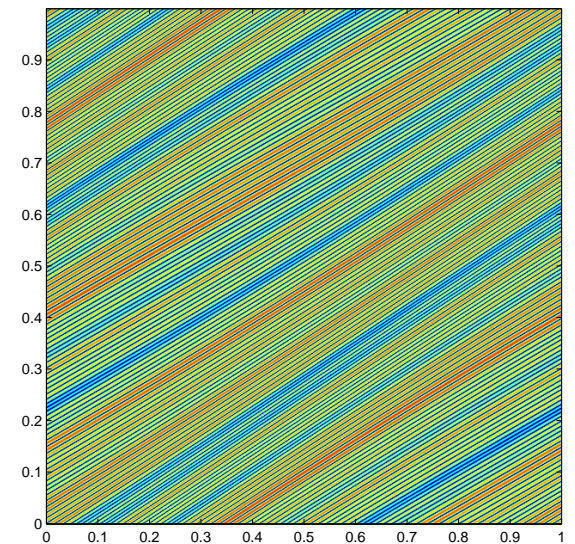
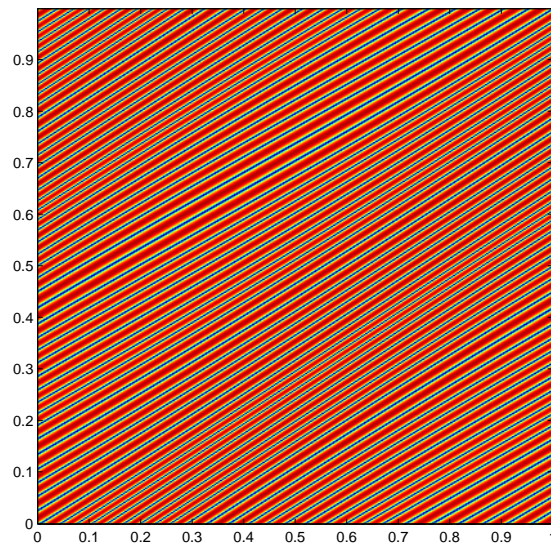
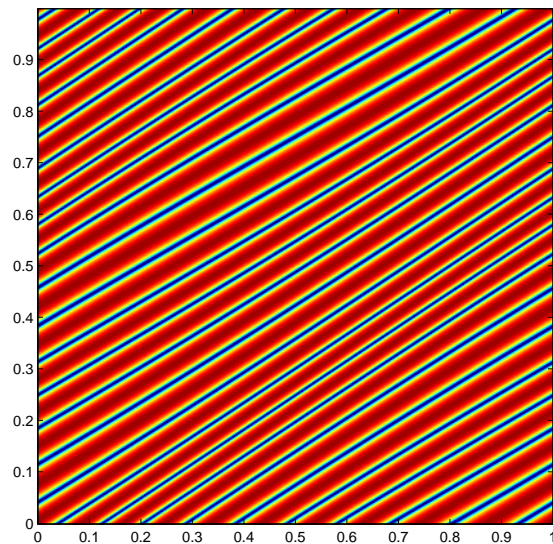
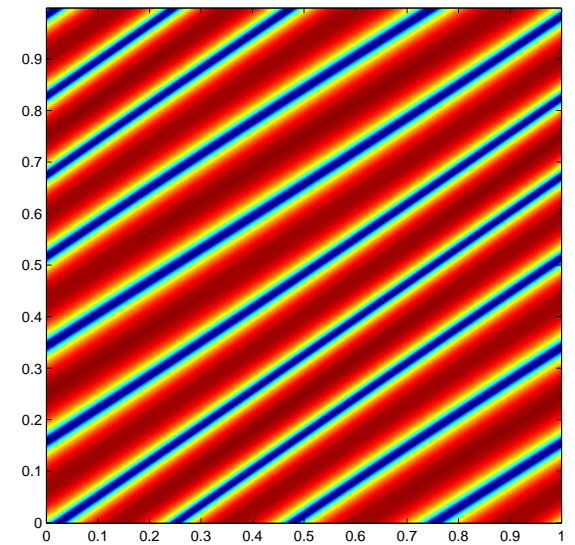
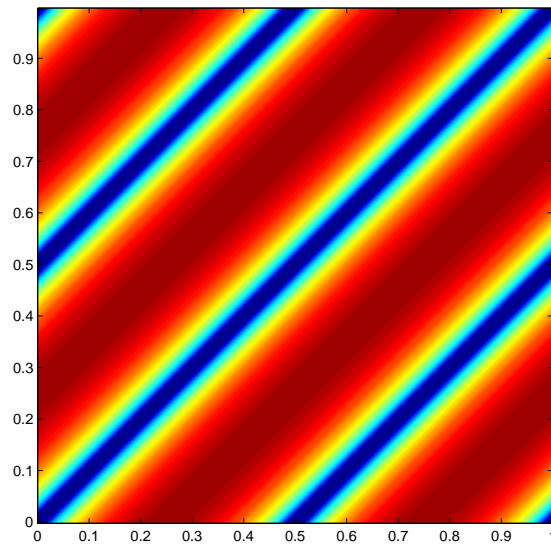
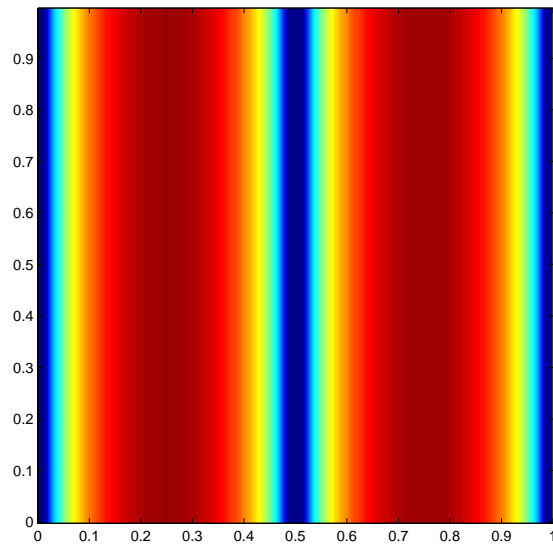
Three phases:

- The variance is initially **constant**;
- It then undergoes a rapid **superexponential** decay;
- $\theta^{(i)}$ settles into an eigenfunction of the A–D operator that sets the **exponential** decay rate.

Decay of Variance

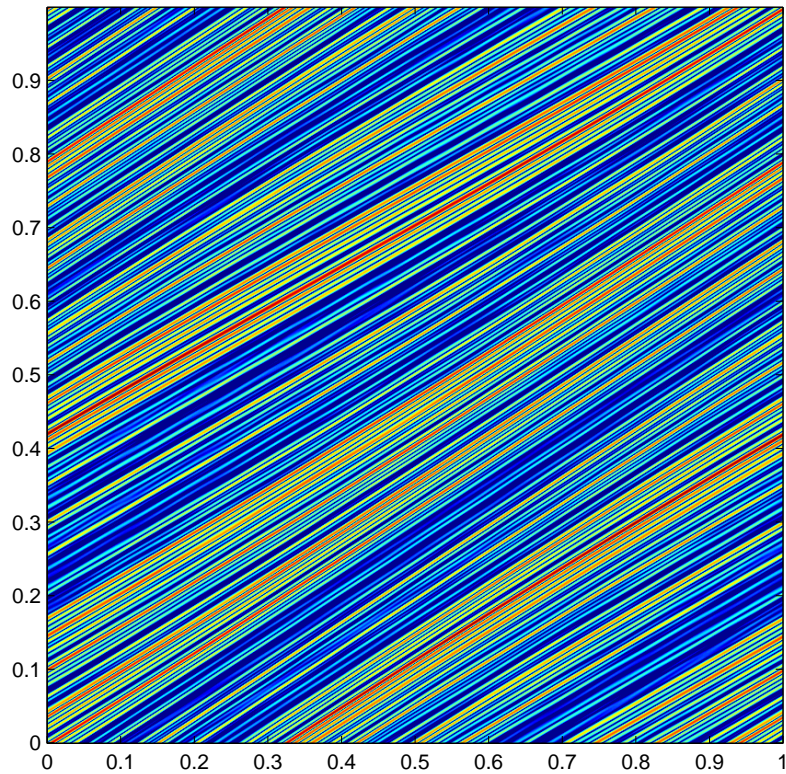


Variance: 5 iterations for $K = 0.3$ and $\epsilon = 10^{-3}$

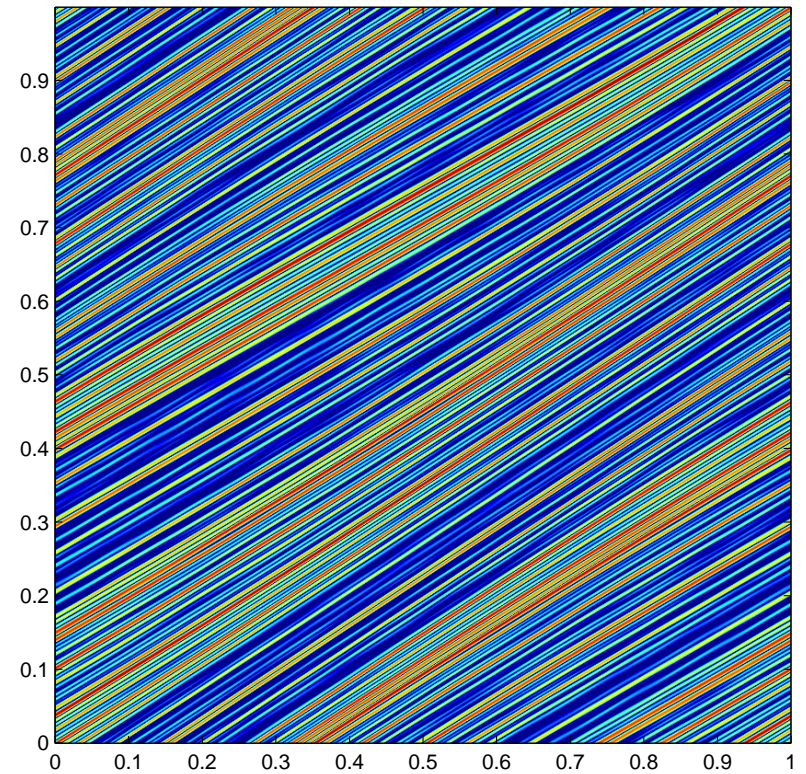


Eigenfunction for $K = 0.3$ and $\epsilon = 10^{-3}$

(Renormalised by decay rate)

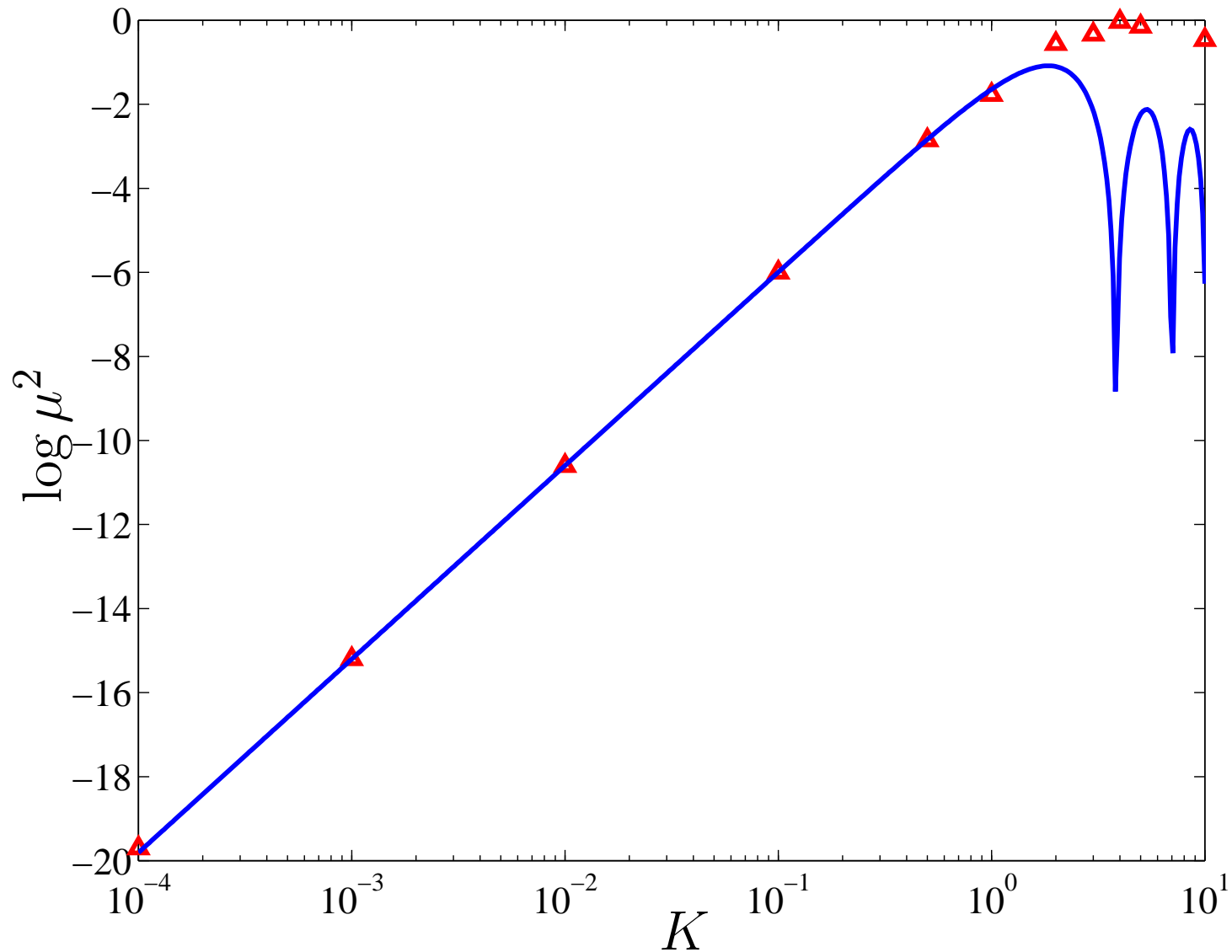


$i = 25$



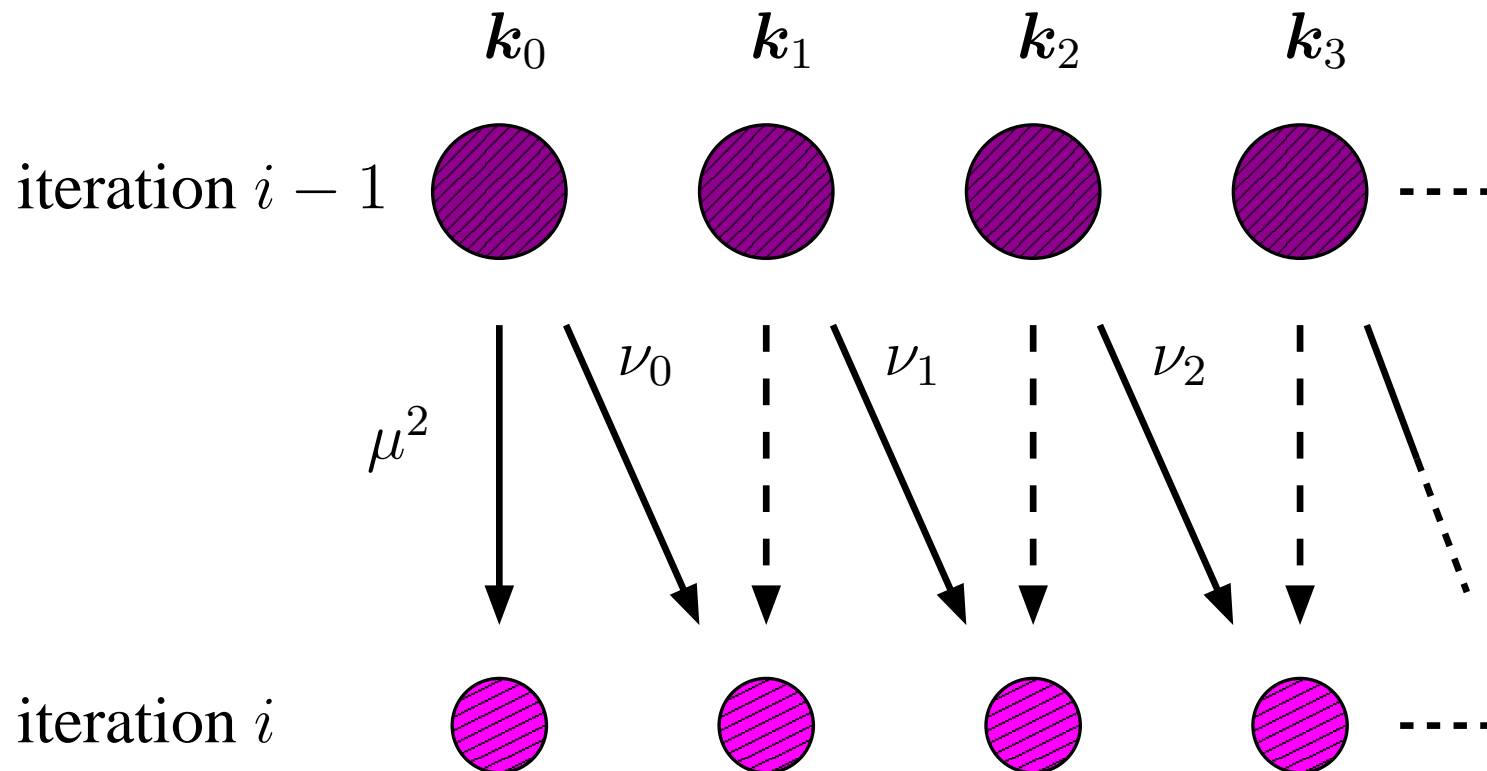
$i = 30$

Decay Rate as $\epsilon \rightarrow 0$

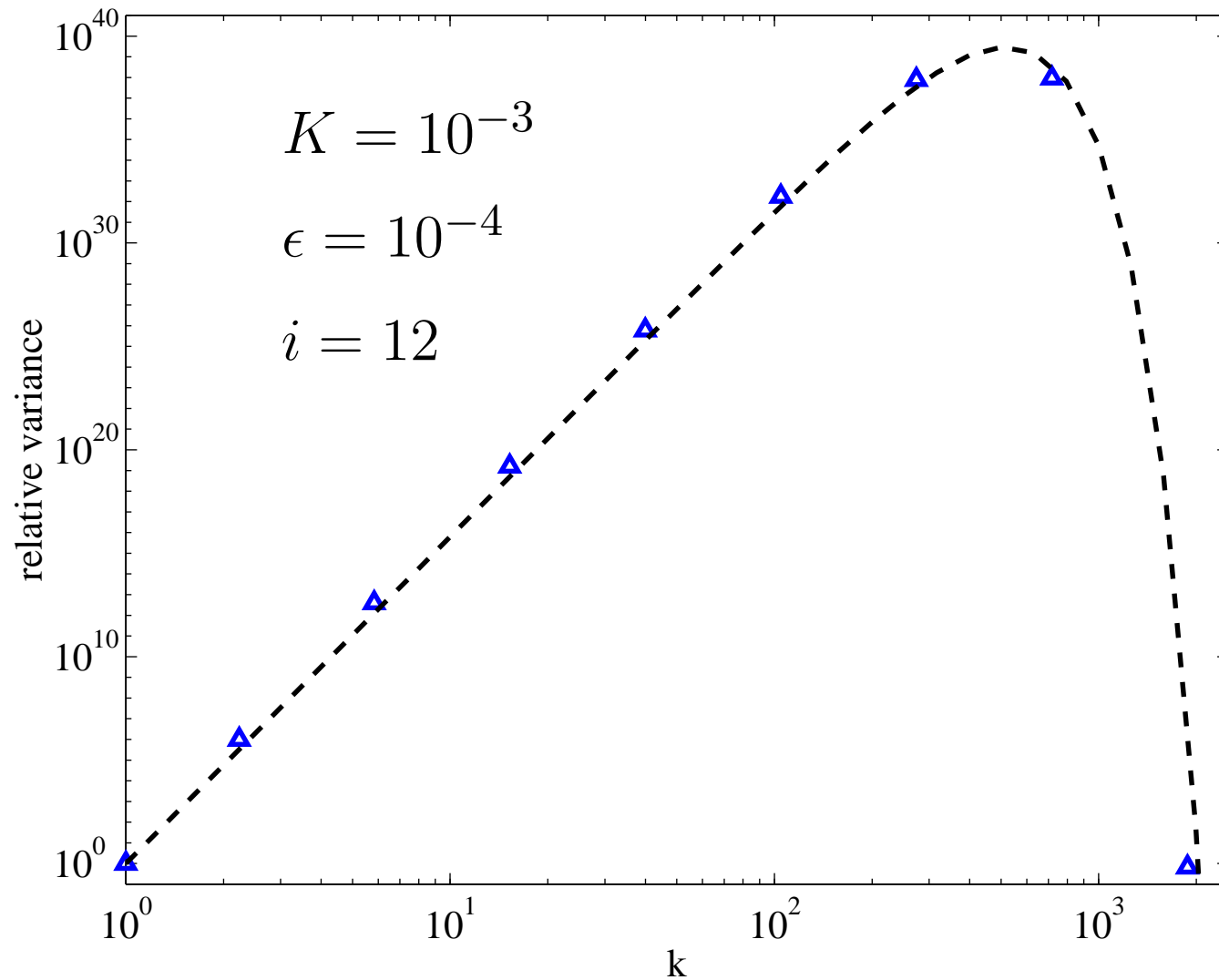


Eigenfunction: One Iteration

The wavenumbers are mapped back to themselves, with their variance decreased by a uniform factor $\mu^2 < 1$ (**vertical arrows**). But at the same time the modes are mapped to next one down the cascade following the **diagonal arrows**.



Spectrum of Variance



Conclusions

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