Hexagonal Fourier-Galerkin Methods for the Two-Dimensional Homogeneous Isotropic Decaying Turbulence

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Abstract. In this paper, we propose two hexagonal Fourier-Galerkin methods for the direct numerical simulation of the two-dimensional homogeneous isotropic decaying turbulence. We first establish the lattice Fourier analysis as a mathematical foundation. Then a universal approximation scheme is devised for our hexagonal Fourier-Galerkin methods for Navier-Stokes equations. Numerical experiments mainly concentrate on the decaying properties and the self-similar spectra of the two-dimensional homogeneous turbulence at various initial Reynolds numbers with an initial flow field governed by a Gaussian-distributed energy spectrum. Numerical results demonstrate that both the hexagonal Fourier-Galerkin methods are as efficient as the classic square Fourier-Galerkin method, while provide more effective statistical physical quantities in general.

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Key words: Fourier-Galerkin methods, hexagonal lattices, homogeneous isotropic turbulence, direct numerical simulation.

1 Introduction

The study of two-dimensional homogeneous isotropic decaying turbulence presents several interests, because of not only its applications to geophysics and astrophysics, but also the basic understanding to hydrodynamic turbulence. Far from being a simplified version of the three-dimensional problem, two-dimensional turbulence presents a rich panorama of new phenomena [3]. There are many remarkable characteristics in 2D turbulence fields, such as coherent structures [1, 13, 19], inverse energy cascade and direct enstrophy cascade [2,14,15]. The inverse energy cascade indicates that the energy is transferred in the inviscid limit from small scales to large scales instead of from large scales...

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to small scales as in the three dimensions. While the enstrophy exhibits a direct cascade process as energy cascade in 3D. The double cascade makes the study of 2D turbulence even more complicated and challenging than that in 3D.

The Navier-Stokes equation for the two-dimensional turbulence may be written in the velocity-vorticity form

$$\frac{\partial \omega}{\partial t} + \nabla \cdot (u \omega) = \nu \Delta \omega,$$

(1.1)

$$\omega = \nabla \times u, \quad \nabla \cdot u = 0,$$

(1.2)

where the kinematic viscosity $\nu$ can be interpreted as the reciprocal of the Reynolds number. Let $\psi$ be the stream function. Then

$$u = (\partial_{x_2} \psi, -\partial_{x_1} \psi), \quad \omega = -\Delta \psi \quad \text{and} \quad \nabla \cdot (u \omega) = \frac{\partial \psi}{\partial x_2} \frac{\partial \omega}{\partial x_1} - \frac{\partial \psi}{\partial x_1} \frac{\partial \omega}{\partial x_2}.$$

Thus, the Navier-Stokes equation for the two-dimensional turbulence can also be written as the stream function-vorticity equation,

$$\frac{\partial \omega}{\partial t} + \frac{\partial \psi}{\partial x_2} \frac{\partial \omega}{\partial x_1} - \frac{\partial \psi}{\partial x_1} \frac{\partial \omega}{\partial x_2} = \nu \Delta \omega,$$

$$-\Delta \psi = \omega.$$

Owing to (1.2), the Jacobian can be reformulated as $\nabla \cdot (u \omega) = \partial_{x_2} \partial_{x_1} (u_2^2 - u_1^2) - (\partial_{x_2}^2 - \partial_{x_1}^2)(u_1 u_2)$, which allows an efficient algorithm for the fast evaluation.

Navier-Stokes equation in two dimensions has the appealing feature to be less demanding on a computational level than the three-dimensional case, allowing to reach relatively high Re numbers in direct numerical simulation (DNS). As a powerful tool, DNS provides some useful theory check and inspires the deeper thoughts of the statistical theory. The spectral method has been becoming very popular in the research of highly accurate numerical simulations since the pioneer work of Orszag and Patterson [21]. DNS of 2D turbulence followed an explosive trend in the early stage with an increasing resolution from $256^2$ to $4096^2$ or even higher [2, 4, 5, 11, 18, 23]. Up to the present, DNS of 2D homogeneous isotropic decaying turbulence are usually been carried out with a millennial resolution using variants of Fourier spectral or pseudospectral methods with the tensorial Fourier basis functions $\{e^{i(k_1 x_1 + k_2 x_2)}\}_{-n/2 \leq k_1, k_2 \leq n/2 - 1}$ subject to periodic boundary conditions [5, 18, 23].

From a general view, the classic Fourier basis functions are just samples of the complex exponential $e^{i\xi \cdot x}$ on the rectangular lattice $\mathbb{Z}^2$ in the frequency space, i.e., with the wave vectors $\xi = k \in \mathbb{Z}^2$. Nevertheless, they actually form a complete orthogonal system on the Voronoi cell $\{x \in \mathbb{R}^2: -\pi \leq x_1, x_2 \leq \pi\}$ of the dual lattice $2\pi \mathbb{Z}^2$ in the physical space, which can represent the solution of (1.1)-(1.2) by a Fourier series with its coefficients to be determined. Inspired by the success of the plane-wave method in quantum physics, one can also choose a proper lattice $L^*$ (e.g., $2\pi A^{-1} \mathbb{Z}^2$ with certain nonsingular matrix $A$).
in the frequency space such that \( \{ e^{i\kappa \cdot x} : \kappa \in L^* \} \) form a complete orthogonal system on the Voronoi cell \( \Omega \) of the dual lattice \( L \) (e.g., \( AZ^2 \)). Then a Fourier spectral method analogue to the well-known plane wave method on \( \Omega \) can be established.

Among various admissible lattices, the hexagonal lattice is undoubtedly one of the most favorable choices. It has been known for decades that isotropically band-limited signals are sampled 13.4% more efficiently by hexagonal lattices than by rectangular lattices [22]. And it was also shown that the processing algorithms for hexagonal systems are similarly 25-50% more efficient than those for rectangular systems with the same frequency resolution [20]. These advantages initiate us to adopt the hexagonal lattice in frequency space and propose the hexagonal Fourier spectral methods which may promote the efficiency of the classic rectangular Fourier spectral methods both theoretically and numerically.

As it can be predicted, the evaluation of the nonlinear term is the main difficulty in the implementation process of our new methods since it dominates the whole time expense. Generally speaking, a computational cost in \( O(n^4) \) arithmetic operators is required in a direct computation. To conquer this difficulty, we follow the idea in a classic Fourier-Galerkin spectral method. At first, a discrete Fourier analysis and the corresponding hexagonal fast Fourier transform (FFT) are introduced, which actually give an efficient routine for the transformation between the function values in the physical space and the Fourier coefficients corresponding to the frequency space. Then a fast algorithm in a computational complexity of \( O(n^2 \log n) \) is devised for the nonlinear Jacobian term through the convolution formula and our FFTs developed.

In all, the primary goal of this paper is to propose the hexagonal Fourier spectral methods and develop their corresponding fast implementation algorithms with a potential promotion in efficiency. In the sequel, we shall concentrate ourselves to prove the effectiveness and efficiency of our new Fourier spectral methods with a series of numerical experiments appeared in the recent physical study on the DNS of the 2D homogeneous isotropic turbulence. We would like to emphasize that all our experiments here are carried out only to recover some interesting results in a classic DNS study by the rectangular Fourier-Galerkin method, without any intention of new mechanism exploration or new phenomena discovery.

The paper is organized as follows. In the next section we present some background materials on planar lattices, Fourier analysis and fast Fourier transform, which set up a mathematical foundation to our further study. Both the dual and the uniform hexagonal Fourier-Galerkin spectral methods are proposed in a unified framework in Section 3 for the DNS of the 2D homogeneous isotropic turbulence. Efficient implementation features, including initial conditions and the fast evaluation of the nonlinear term, are described in details. The performances of our new methods in CPU and GPU environments are also analyzed. Section 4 is then devoted to a numerical comparison between our new methods and the classic Fourier spectral method for the study on the decay of homogeneous isotropic turbulence at both low Reynolds numbers and high Reynolds numbers. The self-similar behaviors of the energy and the enstrophy spectra are examined numerically.
in Section 5 through our Fourier methods proposed. Finally, conclusions and remarks are drawn in Section 6.

2 Mathematical foundation

2.1 Planar lattices and Fourier series

A planar lattice is a discrete subgroup of \( \mathbb{R}^2 \) which represents a regular, periodic array of points in the plane. One can specify each planar lattice \( L \) with a nonsingular matrix \( A \) such that

\[
L = AZ^2 = \{ Ak: k \in \mathbb{Z}^2 \}.
\]

In this case, \( A \) is called a generator matrix of the lattice \( L \). Any plane lattice has a dual lattice (reciprocal lattice) \( L^* \) given by

\[
L^* = \{ 2\pi x \in \mathbb{R}^2: x^t y \in \mathbb{Z}, \forall y \in L \},
\]

where \( x^t \) denotes the transpose of \( x \), and \( x^t y = x \cdot y \) is the usual Euclidean inner product of \( x \) and \( y \). The generator matrix of the dual lattice of \( AZ^2 \) is \( A^* = 2\pi A^{-1} \), i.e., \( (AZ^2)^* = A^*Z^2 \) [16].

A bounded domain \( \Omega \subset \mathbb{R}^2 \) is said to (strictly) title the plane \( \mathbb{R}^2 \) with the lattice \( L \) if

\[
\sum_{\kappa \in L} \chi_\Omega(x + \kappa) = 1, \forall x \in \mathbb{R}^2,
\]

where \( \chi_\Omega \) denotes the characteristic function of \( \Omega \). We write this as \( \Omega + L = \mathbb{R}^2 \), and call such an \( \Omega \) a primitive cell (fundamental domain) of \( L \). There is no unique way to choose a primitive cell. The parallelogram \( \{ Ax: x \in [1,0)^2 \} \) is an obvious primitive cell of \( AZ^2 \); however, such a primitive cell does not necessarily reveal the underlying symmetry of the lattice in which it is embedded. Fortunately, the Wigner-Seitz cell (Voronoi cell) gives a common choice for a primitive cell with the fully symmetry of the given lattice. A Wigner-Seitz cell about a lattice point is the region which is closer to that point than any other lattice points. Hereafter, whenever \( \Omega \) is used as a Wigner-Seitz cell of a lattice \( L \), we fix it as the primitive Wigner-Seitz cell of \( L \) about 0, i.e.,

\[
\Omega + L = \mathbb{R}^2, \quad \text{and} \quad |x| = \min_{y \in L} |x - y| \forall x \in \Omega.
\]

It is worthy to note that the area of a primitive cell \( \Omega \) is uniquely determined by the lattice \( L \) itself. This invariant, independent of the choice of cell, is denote by \( |L| \). For a given lattice \( L = AZ^2 \), \( |\Omega| = |\det(A)| = |L| \).

**Square lattice.** Taking \( S := 2\pi I \) as the generator matrix, one obtains the square lattice \( SZ^2 \) which serves as a simple cubic lattice in the plane. The Wigner-Seitz cell of \( SZ^2 \) is \( \Omega_S := [-\pi,\pi)^2 \), meanwhile the parallelepiped primitive cell \( [0,2\pi)^2 \) may also be conventionally used. The dual lattice of \( SZ^2 \) is exactly the integer lattice \( Z^2 \) with the Wigner-Seitz cell \( \Omega_S^* = [-\frac{1}{2},\frac{1}{2})^2 \).
Hexagonal lattice. Taking
\[ H := \begin{pmatrix} \sqrt{3}r & 0 \\ -r & 2r \end{pmatrix} \]
with \( r > 0 \) as the generator matrix, one obtains the hexagonal lattice \( H\mathbb{Z}^2 \), whose Wigner-Seitz cell is the regular hexagon \( \Omega_H := \{ x \in \mathbb{R}^2 : -\frac{\sqrt{3}}{2} x_1 + \frac{1}{2} x_2, x_2 < r \} \). Explicitly,
\[ H^* = \frac{\pi}{\sqrt{3}r} \begin{pmatrix} 2 & 1 \\ 0 & 3 \end{pmatrix}. \]
Thus the dual lattice \( H^*\mathbb{Z}^2 \) is a dual hexagonal lattice, whose Wigner-Seitz cell is \( \Omega_{H^*} := \{ x \in \mathbb{R}^2 : -\frac{\pi}{\sqrt{3}} x_1 + \frac{1}{\sqrt{3}} x_2, x_1 < \pi \sqrt{3} r \} \). In the current paper, we shall fix \( r = \sqrt{\frac{4}{3} \pi} \) to ensure \( |\Omega_H| = |\Omega_S| = 4\pi^2 \) and \( |\Omega_{H^*}| = |\Omega_{S^*}| = 1 \).

Tiling and Fourier analysis are closely related (cf. [12]) due to the following theorem.

**Theorem 2.1** (Fuglede [12]). Let \( \Omega \) be a bounded domain in \( \mathbb{R}^2 \) and \( L \) be a lattice of \( \mathbb{R}^2 \). Then \( \Omega + L = \mathbb{R}^2 \) if and only if \( \{ \phi_{\kappa}(x) = e^{i\kappa \cdot x} : \kappa \in L^* \} \) is an orthonormal basis,
\[ \langle \phi_j, \phi_\kappa \rangle = \frac{1}{|\Omega|} \int_{\Omega} \overline{\phi_k(x)} \phi_k(x) dx = \delta_{j,\kappa}, \] \( j, \kappa \in L^* \). (2.1)

A point \( x \) is said to be congruent to \( y \in \mathbb{R}^2 \) with respect to the lattice \( L \) if \( x = y + L \). Further a function \( f \) defined on \( \mathbb{R}^2 \) is called periodic with respect to \( L \) if and only if \( f(x) = f(y) \) for any \( x \in y + L \) and \( y \in \mathbb{R}^2 \). If no confusion would arise, we simply call \( f \) periodic.

Let \( \Omega \) be any primitive cell of the lattice \( L \). Assume \( f \) is a periodic function with respect to \( L \), which is square-integrable on \( \Omega \). Then \( f \) has the following Fourier series,
\[ f(x) = \sum_{\kappa \in L^*} \hat{f}_\kappa \phi_\kappa(x), \quad \hat{f}_\kappa = \langle f, \phi_\kappa \rangle = \frac{\hat{f}(\kappa)}{|\Omega|}, \] \( \kappa \in L^* \). (2.2)
where \( \hat{f} \) is the Fourier transform of \( f\chi_{\Omega} \),

\[
\hat{f}(\xi) = \int_{\mathbb{R}^2} f(x) \chi_{\Omega}(x) e^{-i\xi^t x} dx = \int_{\Omega} f(x) e^{-i\xi^t x} dx.
\]

### 2.2 Discrete Fourier transform

Up to present, we have used the lattice \( L \) and its dual lattice \( L^* \) to determine the integral domain \( \Omega \) in physical space and the wavenumber \( \kappa \in L^* \) in the reciprocal (frequency) space, which finally set up a Fourier analysis. To establish a discrete Fourier analysis, we take a series of lattices \( \hat{L}_n \), then choose a proper primitive cell \( \hat{\Omega}_n \) of \( \hat{L}_n \) to make a finite set of wavenumbers \( \kappa \in \hat{\Omega}_n \cap L^* \) in the reciprocal space, and use the dual Lattice \( \hat{L}_n^* \) to set up the quadrature/interpolation nodes \( z \in \Omega \cap \hat{L}_n^* \) in the physical space.

**Theorem 2.2 ([16]).** Let \( \Omega \) be a primitive cell of \( L = AZ^2 \). Assume \( B \) is a non-singular matrix such that all entries of \( B^t A \) are integer. Then for any \( \hat{L}_n = 2\pi n BZ^2 \) with positive integer \( n \),

\[
\frac{1}{N} \sum_{z \in \Omega \cap \hat{L}_n^*} e^{i\kappa^t z} = \begin{cases} 1, & \kappa \in L^* \cap \hat{L}_n, \\ 0, & \kappa \in L^* \setminus \hat{L}_n, \end{cases} \tag{2.3}
\]

where \( N := \text{card}(\Omega \cap \hat{L}_n^*) = |\det(nB^tA)| \).

Further let \( \hat{\Omega}_n \) be the Wigner-Seitz cell of the lattice \( \hat{L}_n \). Denote by \( \hat{Y}_n = \Omega \cap \hat{\Omega}_n^* \) the set of nodes in physical space and by \( \hat{Y}_n^* = \hat{\Omega}_n \cap L^* \) the set of wave numbers in reciprocal space. Then it is easy to verify \( \text{card}(\hat{Y}_n) = \text{card}(Y_n) = N \). We define the following forward and backward discrete Fourier transform (DFT),

\[
\hat{f}_\kappa = \frac{1}{N} \sum_{z \in \hat{Y}_n} f(z) e^{-i\kappa^t z}, \quad \kappa \in \hat{Y}_n, \tag{2.4}
\]

\[
f(z) = \sum_{\kappa \in \hat{Y}_n} \hat{f}_\kappa e^{i\kappa^t z}, \quad z \in \hat{Y}_n. \tag{2.5}
\]

Then the discrete Fourier coefficient \( \hat{f}_\kappa \) provides a spectral approximation to the Fourier coefficient \( \hat{f}_\kappa \) for any \( \kappa \in \hat{Y}_n \),

\[
\hat{f}_\kappa \overset{(2.2)}{=} \frac{1}{N} \sum_{z \in \hat{Y}_n} \sum_{j \in L^*} \hat{f}_j e^{i(j - \kappa)^t z} \overset{(2.3)}{=} \sum_{j \in \kappa + \hat{L}_n} \hat{f}_j, \quad \kappa \in L^*. \tag{2.6}
\]

**DFT on square lattices.** Let \( L = S \mathbb{Z}^2 \) and \( \hat{L}_n = n(S \mathbb{Z}^2)^* = n \mathbb{Z}^2 \). Then \( \Omega = \Omega_S \) and \( \hat{\Omega}_n = n\Omega_S^* = \left[-\frac{n}{2}, \frac{n}{2}\right]^2 \) are the Wigner-Seitz cells of \( L \) and \( \hat{L}_n \), respectively. Now the lattice DFT
in (2.4)-(2.5) is reduced to
\[
\hat{n} = \frac{1}{n^2} \sum_{j \in \Lambda_n^S} f(\frac{2\pi j}{n}) e^{-\frac{2\pi ikj}{n}}, \quad k \in \Lambda_n^S,
\]
\[
f(\frac{2\pi j}{n}) = \sum_{k \in \Lambda_n^S} \hat{n} e^{-\frac{2\pi ikj}{n}}, \quad j \in \Lambda_n^S.
\]
where \(\Lambda_n^S := \mathbb{Z}^2 \cap [-\frac{n}{2}, \frac{n}{2})^2\) = \(\hat{Y}_n = \frac{n}{2\pi}\mathbb{Z}_n\). Owing to the periodicity, this square lattice DFT coincides with the classic DFT up to a permutation, such that it can be efficiently evaluated in \(\mathcal{O}(n^2 \log n)\) arithmetic operations through the classic FFT.

**DFT on dual hexagonal lattices.** Let \(L = H\mathbb{Z}^2\) and \(\tilde{L}_n = nH^*\mathbb{Z}^2\). Take \(\Omega = \Omega_H\) and \(\hat{\Omega}_n = n\Omega_H^*\) = \(\{\xi \in \mathbb{R}^2: -\frac{m}{\sqrt{3}} \leq \sqrt{3} \xi_2 \pm \frac{1}{2} \xi_1 \xi_1 < \frac{\pi m}{\sqrt{3}}\}\) as the Wigner-Seitz cells of the lattice \(L\) and \(\tilde{L}_n\), respectively. The DFT on dual hexagonal lattices is then defined as follows
\[
\hat{f}_{H^*k} = \frac{1}{n^2} \sum_{j \in \Gamma_n} f(\frac{1}{n}Hj) e^{-\frac{2\pi ikj}{n}}, \quad k \in \hat{\Gamma}_n,
\]
\[
f(\frac{1}{n}Hj) = \sum_{k \in \hat{\Gamma}_n} \hat{f}_{H^*k} e^{-\frac{2\pi ikj}{n}}, \quad j \in \Gamma_n,
\]
where \(\Gamma_n := \{k \in \mathbb{Z}^2: -n \leq 2k_1 + 2k_2 + k_1, k_2 - k_1 < n\} = \{k \in \mathbb{Z}^2: \frac{1}{2}Hk \in \Omega_H\}\), and \(\hat{\Gamma}_n := \{k \in \mathbb{Z}^2: -n \leq 2k_1 - 2k_2 + k_1, k_2 + k_1 < n\} = \{k \in \mathbb{Z}^2: H^*k \in n\Omega_H^*\}\). The DFT on dual hexagonal lattices coincides with the classic DFT up to a permutation, thus can be efficiently evaluated through classic FFT just as the DFT on square lattices.

**DFT on uniform hexagonal lattices.** Let \(L = H\mathbb{Z}^2\) and \(\hat{L}_n = \frac{n}{\sqrt{3}}H\mathbb{Z}^2\). Take \(\Omega = \Omega_H\), and \(\hat{\Omega} = \frac{n}{\sqrt{3}}\Omega_H\) as their Wigner-Seitz cells. Now the matrix
\[
N = \left(\frac{n}{2\sqrt{3}} H\right)^* H = \begin{pmatrix} 2n & -n \\ -n & 2n \end{pmatrix}
\]
Figure 3: The set of wavenumbers $n\Omega_H \cap H^*Z^2 = \{H^*k : k \in \tilde{\Gamma}_n\}$ in reciprocal space (left), the index set $\tilde{\Gamma}_n$ (middle) and its congruent index set $\Lambda_n$ (right side). The congruence relation of the index sets provides a storage scheme for the discrete Fourier coefficients.

Figure 4: The set of quadrature/interpolation nodes $\Omega_H \cap n^{-1}H^*Z^2 = \{n^{-1}Hk : k \in \Gamma_n\}$ (left side), index set $\Gamma_n$ (middle) and its congruent index set $\Lambda_n$ (right side). The congruence relation of the index sets provides a storage scheme for the function values in physical space.

has integer entries, and the index set $\Lambda_n^H := \{k \in Z^2 : -n \leq k_1, k_2, k_1 + k_2 \leq n - 1\} = \{k \in Z^2 : H^*k \in \frac{\pi n}{2} \Omega_H\} = \{k \in Z^2 : \frac{r^2}{\pi n} H^*k \in \Omega_H\}$. Then the DFT on uniform hexagonal lattices takes the following form

$$\hat{f}_{H^*k} = \frac{1}{3n^2} \sum_{j \in \Lambda_n^H} f\left(\frac{r^2}{\pi n} H^*j\right) e^{-2\pi ijN^{-1}k}, \quad k \in \Lambda_n^H,$$

$$f\left(\frac{r^2}{\pi n} H^*j\right) = \sum_{k \in \Lambda_n^H} \hat{f}_{H^*k} e^{+2\pi ijN^{-1}k}, \quad j \in \Lambda_n^H,$$

for which the hexagonal FFT is designed by a generalized Cooley-Tukey algorithm based on the periodicity matrix factorization

$$N = nI \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}.$$  

The hexagonal FFT is decomposed into three classic FFTs of size $n \times n$, a series of twiddle factor multiplications and a data permutation, which can be fulfilled eventually in $O(n^2 \log n)$ arithmetic operations [9, 17, 24].
Figure 5: The set of nodes/wavenumbers \( \Omega_H \cap \mathbb{Z}^2 = \{ \frac{n}{2\pi H^*} k : k \in \Lambda_H^* \} / \pi \mathbb{Z} \) in physical/reciprocal space (left), the index set \( \Lambda_H^* \) (center) and its congruent index set \( \Pi_H^* = \mathbb{Z}^2 \cap [0, 2n - 1] \times [0, n - 1] \).

### 3 Hexagonal Fourier-Galerkin methods

In this section, we shall propose a lattice Fourier spectral method and its implementation for the two-dimensional homogeneous isotropic turbulence. This lattice Fourier spectral method serves a canonical framework for both the hexagonal Fourier spectral methods and the classic square Fourier spectral method.

#### 3.1 Fourier-Galerkin approximation schemes

Let \( \Omega \) and \( \hat{\Omega}_n \) be the Wigner-Seitz cell of the lattice \( L = A \mathbb{Z}^2 \) in the physical space and the lattice \( \hat{L}_n = 2\pi n B \mathbb{Z}^2 \) in the reciprocal space, respectively. Assume all entries of \( B^t A \) are integer so that a discrete Fourier analysis can be established. Denote \( \hat{Y}_n = \hat{\Omega}_n \cap L^* \) and \( N = \text{card}(\hat{Y}_n) \).

The semi-discrete Fourier-Galerkin spectral approximation for the direct numerical simulation of 2-dimensional turbulence is to find

\[
\omega(x, t) = \sum_{\kappa \in \hat{Y}_n} \hat{\omega}_\kappa(t) \phi_\kappa(x)
\]

such that

\[
\frac{d\hat{\omega}_\kappa(t)}{dt} + \hat{J}_\kappa(t) = -\nu |\kappa|^2 \hat{\omega}_\kappa(t), \quad \kappa \in \hat{Y}_n,
\]

where \( \hat{J}_\kappa \) is the Fourier coefficient of \( \nabla \cdot (u \omega) \) with respect to the wave-number \( \kappa \),

\[
\hat{J}_\kappa = \left[ \nabla \cdot (u \omega) \right]_\kappa = -\kappa_1 \kappa_2 (u_2^2 - u_1^2)_\kappa + (\kappa_2^2 - \kappa_1^2) (u_1 u_2)_\kappa,
\]

\[
(u_1)_\kappa = \frac{i \kappa_2}{|\kappa|^2} \omega_\kappa, \quad (u_2)_\kappa = -\frac{i \kappa_1}{|\kappa|^2} \omega_\kappa.
\]
which gives Dual hexagonal Fourier-Galerkin method (Method I). Let direct numerical simulation of the two-dimensional homogeneous isotropic turbulence.

\[
\hat{u}_l = \frac{3}{\Delta t} \sum_{m=1}^{\infty} a_m \hat{\omega}_m - \frac{3}{\Delta t} \sum_{m=1}^{\infty} b_m \hat{\omega}_m = -v |\kappa|^2 \hat{\omega}_m, \quad m \geq 3, \quad \kappa \in \hat{Y}_n,
\]

where \( a_0 = \frac{11}{\xi}, a_1 = -3, a_2 = \frac{3}{\xi}, a_3 = -\frac{1}{\xi}, \) and \( b_1 = 3, b_2 = 3, b_3 = 3. \) And the time step \( \Delta t \) is small enough to match the CFL constraint.

Two types of hexagonal Fourier-Galerkin methods are proposed in this paper for the direct numerical simulation of the two-dimensional homogeneous isotropic turbulence.

**Dual hexagonal Fourier-Galerkin method (Method I).** Let \( L = HZ^2 \) and \( \Omega = \Omega_H. \) Further take \( \hat{L}_n = nH^*Z^2 \) and \( \hat{\Omega}_n = n\Omega^*_H. \) Then (3.1) is reduced to the dual hexagonal Fourier-Galerkin method for the \( HZ^2 \)-periodic problem of (1.1)-(1.2) with \( \hat{Y}_n = \{ H^*k : k \in \hat{Y}_n \}. \)

**Uniform hexagonal Fourier-Galerkin method (Method II).** Let \( L = HZ^2 \) and \( \Omega = \Omega_H. \) Meanwhile we take \( \hat{L}_n = \frac{\pi n}{r} HZ^2 \) and \( \hat{\Omega}_n = \frac{\pi n}{r} \Omega_H. \) Then (3.1) is reduced to the uniform hexagonal Fourier-Galerkin method for the \( HZ^2 \)-periodic problem of (1.1)-(1.2) with \( \hat{Y}_n = \{ H^*k : k \in \Lambda^*_n \}. \)

Besides, we shall compare the hexagonal Fourier-Galerkin methods with the classic square Fourier-Galerkin method (Method III). The later one is just enforced by setting \( L = SZ^2, \hat{L}_n = nZ^2 \) and \( \hat{\Omega}_n = -\left( \frac{n}{2}, \frac{n}{2} \right)^2 \) such that \( \hat{Y}_n = \Lambda^*_n, \) which is commonly used for the numerical study on homogeneous turbulence.

### 3.2 Initial conditions

Taking the Fourier transform on (1.2), we have

\[
i\bar{\xi}_1 \hat{u}_2(\xi) - i\bar{\xi}_2 \hat{u}_1(\xi) = \hat{\omega}(\xi), \quad i\bar{\xi}_1 \hat{u}_1(\xi) + i\bar{\xi}_2 \hat{u}_1(\xi) = 0,
\]

which gives

\[
\hat{u}(\xi) = (\hat{u}_1(\xi), \hat{u}_2(\xi))^t = \frac{\hat{\omega}(\xi)}{i|\xi|^2} (-\bar{\xi}_2, \bar{\xi}_1)^t.
\]

Thus, by applying the Parseval’s theorem on the formula of the kinetic energy, one obtains

\[
E = \frac{1}{2 |\Omega|} \int_{\Omega} \|u(x)\|^2 dx = \frac{1}{8\pi^2 |L|} \int_{\mathbb{R}^2} |\hat{u}(\xi)|^2 d\xi = \frac{1}{2 |L|} \int_{\mathbb{R}^2} \frac{|\hat{\omega}(\xi)|^2}{2|\xi|^2} d\xi.
\]

By using the polar coordinates \( \xi = (k \cos \theta, k \sin \theta)^t, \) one further derives

\[
E = \int_0^\infty \left[ \frac{1}{8\pi^2 |L|} \int_0^{2\pi} |\hat{\omega}(k \cos \theta, k \sin \theta)|^2 d\theta \right] dk = \int_0^\infty E(k) dk,
\]
where \( E(k) = \frac{1}{8\pi^2k|L|} \int_0^{2\pi} |\hat{\omega}(k\cos\theta, k\sin\theta)|^2 d\theta \) is the energy spectrum of wave number magnitude \( k \). Suppose \( \hat{\omega} \) is isotropic (axisymmetric), i.e., \( |\hat{\omega}(\xi)| = |\hat{\omega}(|\xi|, 0)| \) for any \( \xi \in \mathbb{R}^2 \). Then

\[
E(k) = \frac{1}{4\pi^2k|L|} |\hat{\omega}(\kappa)|^2 = \frac{|L|}{4\pi^2k} |\hat{\omega}_\kappa|^2, \quad |\kappa| = k.
\]

We specify initial conditions for our flow field by assuming an energy spectrum of wave number magnitude \( k \) of the general form

\[
E(k,t=0) = \frac{(2s+1)^{s+1}}{2^{s+1}\Gamma(s+1)} \bar{u}_0^2 k_c^{-1} \left( \frac{k}{k_c} \right)^{2s+1} \exp \left[ -(s+\frac{1}{2}) \left( \frac{k}{k_c} \right)^2 \right], \quad s \geq 0. \tag{3.3}
\]

The initial vorticity field \( \omega_0 \) is generated in frequency space with random phases and with amplitude corresponding to (3.3), i.e.,

\[
\hat{\omega}_{0,\kappa} = \left( \frac{4\pi|\kappa| E(|\kappa|,0)}{|L|} \right)^{1/2} \exp(2\pi i \zeta),
\]

with \( \zeta \) a different uniform deviate for each \( \kappa \) subject to the requirement of complex conjugate symmetry of the Fourier components.

Given the initial energy spectrum \( E(k,0) \), a numerical simulation is uniquely identified by its Reynolds number \( R(t) \) at \( t=0 \), where the Reynolds number at time \( t \) is defined by

\[
R(t) = \frac{\bar{u}(t)\bar{I}(t)}{\nu}, \quad \bar{u} = \langle u^2 \rangle^{1/2}, \quad \bar{\omega} = \langle \omega^2 \rangle^{1/2}, \quad \bar{I} = \bar{u}/\bar{\omega}, \tag{3.4}
\]

hereafter we use the notations \( \langle \cdot \rangle \) for \( \langle \cdot, \cdot \rangle \). In the case of the initial energy spectrum (3.3), one has

\[
\bar{u}(0) = \bar{u}_0, \quad \bar{\omega}(0) = \sqrt{\frac{2s+2}{2s+1}} \bar{u}_0 k_c, \quad \bar{I}(0) = \sqrt{\frac{2s+1}{2s+2}} k_c^{-1}, \quad R(0) = \sqrt{\frac{2s+1}{2s+2}} \bar{u}_0 k_c. \tag{3.5}
\]

### 3.3 Spectral coefficients of Jacobian

Owing to (3.2), to compute the Fourier coefficients \( \hat{J}_\kappa \) of the Jacobian, it is pivotal to determine the Fourier coefficients of product of two functions \( u,v \in \mathcal{H}_n = \{ \phi_\kappa : \kappa \in \hat{\mathbb{Y}}_n \} \). Owing to the discrete orthogonality (2.3), the Fourier coefficients \( \langle \hat{u} \hat{v} \rangle_\kappa, \kappa \in \hat{\mathbb{Y}}_n := \hat{\Omega}_n \cap L^n \) are given by the following convolution sum,

\[
\hat{\omega}_\kappa = \langle \hat{u} \hat{v} \rangle_\kappa = \sum_{j \in \hat{\mathbb{Y}}_n} \hat{u}_j \hat{v}_{\kappa-j}. \tag{3.6}
\]
As we know, the direct summation implied by (3.6) takes $O(n^4)$ arithmetic operators. This is prohibitively expensive when compared with pseudospectral approximation. However, the aliasing relation (2.6) enables us to use the transform method to evaluate (3.6) in $O(n^2 \log n)$ operations [6]. The key to this method is the de-aliasing technique by using FFTs with the lattice $\mathbf{L}_m = 2\pi m \mathbf{BZ}^2$ rather than $\mathbf{L}_n = 2\pi n \mathbf{BZ}^2$ just as in a Cartesian case [6], where $m \geq 3/2n$. We now give a short description. First extend the Fourier coefficients according to

$$\tilde{u}_\kappa = \begin{cases} \hat{u}_\kappa, & \kappa \in \mathbf{\mathbf{\Lambda}}_n, \\ 0, & \kappa \in \mathbf{\mathbf{\Lambda}}_m \setminus \mathbf{\mathbf{\Lambda}}_n, \end{cases} \quad \tilde{v}_\kappa = \begin{cases} \hat{v}_\kappa, & \kappa \in \mathbf{\mathbf{\Lambda}}_n, \\ 0, & \kappa \in \mathbf{\mathbf{\Lambda}}_m \setminus \mathbf{\mathbf{\Lambda}}_n. \end{cases}$$

Next evaluate $u(z), v(z), z \in Y_m = \Omega \cap \mathbf{\mathbf{\Lambda}}^*_m$ through backward FFTs,

$$u(z) = \sum_{\kappa \in \mathbf{\mathbf{\Lambda}}_m} \tilde{u}_\kappa e^{2\pi i \kappa z}, \quad v(z) = \sum_{\kappa \in \mathbf{\mathbf{\Lambda}}_m} \tilde{v}_\kappa e^{2\pi i \kappa z},$$

and get the function values $w(z) = u(z)v(z), z \in Y_m$. Further by a forward FFT, we obtain the discrete Fourier coefficients

$$\hat{w}_\kappa = \frac{N^2}{m^2 N} \sum_{z \in Y_m} w(z) e^{-2\pi i \kappa z}, \quad \kappa \in \mathbf{\mathbf{\Lambda}}_m.$$ 

Now $\hat{w}_\kappa = \hat{\tilde{w}}_\kappa$ for any $\kappa \in \mathbf{\mathbf{\Lambda}}_n$, and we derive the Fourier coefficients $(uv)_\kappa, \kappa \in \mathbf{\mathbf{\Lambda}}_n$ in three FFTs.

As the Jacobian (3.2) is concerned, one mainly needs two backward FFTs to get the function values of $u_1$ and $u_2$, and two forward FFTs to get the discrete coefficients $(u_1^2 - u_2^2)_\kappa$ and $(u_1 u_2)_\kappa, \kappa \in \mathbf{\mathbf{\Lambda}}_m$, which only amount up to $O(m^2 \log m)$ arithmetic operations.

### 3.4 Performance analysis

Numerical experiments are performed in a heterogeneous system equipped with GPU devices. Each node has a dual Hexacore Intel X5660 CPU (12M Cache, 2.80GHz, 6.40 GT/s Intel QPI) and 32GB Memory. Each node is connected with Nvidia Tesla M2070 GPU (1.15GHz, 448 cores, 6GB 1.5GHz DDR5 Memory). For classic FFTs, the FFTW Library 3.0 is adopted in CPU programs, while the CUFFT Library 4.0 is used in GPU programs. CUFFT has a FFTW compatible data layouts but is optimized only for the input data size $n = 2^a 3^b 5^c 7^d$ in each direction.

We show in this section the performance of the hexagonal Fourier-Galerkin methods for the two-dimensional homogeneous turbulence with a total resolution $N = 64^2, 128^2, 256^2, \ldots, 4096^2$. As a comparison, the performance of the classic square Fourier-Galerkin method is also demonstrated. Recall that $n = \sqrt{N}$ for both the dual hexagonal and classic square Fourier-Galerkin methods (resp. Method I and III), while we use $n \approx$
to match the factorization constraint \( n = 2^{a}3^{b}5^{c}7^{d} \). More explicitly, we set \( n = 36(2^{2}3^{2}),75(3^{1}5^{2}),147(2^{0}3^{1}7^{2}),\cdots,2352(2^{3}3^{1}7^{2}) \) in the uniform hexagonal Fourier-Galerkin method (Method II) for \( N=64,128,256,\cdots,4096 \) such that \( 1 - 3n^2/N \approx 5.08\%, -3.00\%, 1.08\%, \cdots, 1.08\% \), respectively.

The algorithm described in Section 3.3 with \( m = 3/2n \), which is usually referred to as the 3/2-rule for de-aliasing, is adopted for the evaluation of the Fourier coefficients of the nonlinear Jacobian term. To avoid the expensive, time-consuming data transfer between the host (CPU) memory and the device (GPU) memory in the GPU programs, we would like to keep the entire datum residing on the device memory all over the time.

At first, the average elapsed time per time step in the CPU programs for Method I, II and III are reported in Table 1, respectively. Roughly speaking, the elapsed time for the Jacobian calculation takes more than 80\% of the total elapsed time, so that the total time expense is dominated by the cost of FFTs for the computation of the Jacobian (also refer to Fig. 6). Another obvious observation is that Method II becomes more efficient as the resolution increases, and it is about 2 times faster than other methods for \( N \geq 4096^2 \). This is due to our hexagonal FFT algorithms, which decompose the entire FFT into 3 small classic FFTs, and thus make full use of CPU cache to improve the hit rate of memory access. We also find that time cost for Method I is only slightly higher than that of Method III, since Method I uses the same classic FFT as Method III, and differs from Method III only in a less regular data structure (refer to Figs. 2-4), which finally results in a slowdown in the program speed.

Moreover, we present the average elapsed time per time step for the GPU programs in Table 2. Significant acceleration has been demonstrated in comparison to the corresponding CPU programs. For a high resolution \( N = 4096^2 \), a 45-fold speedup is obtained for Method I, and even higher speedup (49x) for Method III. However, the GPU performance of Method II is relatively low owing to the complicated non-uniform data structure and the corresponding conditional branches. On a conditional branch where the
threads diverge in which path to take, the threads taking different paths have to run serially. In spite of the serious performance degradations caused by thread divergence, Method II still attains a speedup about 20 times on the Nvidia Tesla M2070 GPU.

We shall conclude this section with Fig. 7, which indicates the Gflops output decreases

Figure 7: Gflops statistics for three kinds of Fourier-Galerkin spectral methods versus different scale $N$. 

---

Table 1: The elapsed time (ms) for CPU programs.

<table>
<thead>
<tr>
<th>$N$</th>
<th>64$^2$</th>
<th>128$^2$</th>
<th>256$^2$</th>
<th>512$^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Method</td>
<td>Jacobian</td>
<td>Total</td>
<td>Jacobian</td>
<td>Total</td>
</tr>
<tr>
<td>I</td>
<td>1.08</td>
<td>1.13</td>
<td>6.42</td>
<td>6.64</td>
</tr>
<tr>
<td>II</td>
<td>1.38</td>
<td>1.75</td>
<td>6.08</td>
<td>7.34</td>
</tr>
<tr>
<td>III</td>
<td>0.90</td>
<td>0.94</td>
<td>5.34</td>
<td>5.56</td>
</tr>
</tbody>
</table>

Table 2: The elapsed time (ms) for GPU programs.

<table>
<thead>
<tr>
<th>$N$</th>
<th>64$^2$</th>
<th>128$^2$</th>
<th>256$^2$</th>
<th>512$^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Method</td>
<td>Jacobian</td>
<td>Total</td>
<td>Jacobian</td>
<td>Total</td>
</tr>
<tr>
<td>I</td>
<td>0.179</td>
<td>0.226</td>
<td>0.411</td>
<td>0.500</td>
</tr>
<tr>
<td>II</td>
<td>0.456</td>
<td>0.480</td>
<td>0.883</td>
<td>0.910</td>
</tr>
<tr>
<td>III</td>
<td>0.163</td>
<td>0.191</td>
<td>0.340</td>
<td>0.423</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>$N$</th>
<th>1024$^2$</th>
<th>2048$^2$</th>
<th>4096$^2$</th>
<th>8192$^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Method</td>
<td>Jacobian</td>
<td>Total</td>
<td>Jacobian</td>
<td>Total</td>
</tr>
<tr>
<td>I</td>
<td>24.7</td>
<td>30.8</td>
<td>95.7</td>
<td>122.</td>
</tr>
<tr>
<td>II</td>
<td>40.8</td>
<td>40.9</td>
<td>152.</td>
<td>153.</td>
</tr>
<tr>
<td>III</td>
<td>21.5</td>
<td>27.4</td>
<td>83.3</td>
<td>107.</td>
</tr>
</tbody>
</table>
evidently in our CPU programs – especially in those for Method I and Method III, as N increases. Fortunately, however, there is a continuing uptrend in the Gflops output in the GPU programs as the resolution N increases, which partially demonstrates that GPU implementation of our Fourier-Galerkin methods has a better scalability and thus is adequate for the DNS of 2-D homogeneous isotropic decaying turbulence.

4 Decay of homogeneous isotropic turbulence

4.1 Decay at low Reynolds numbers

We first consider the evolution of the flow field at relatively low initial Reynolds numbers. Chasnov postulated the existence of a critical initial Reynolds number above which the Reynolds number of the turbulence increases asymptotically, and below which it decreases to small values eventually attaining the final period of decay [8].

Just as in [8], we define the logarithmic derivative in time of the energy and enstrophy as follows,

\[ p = \frac{d \ln \langle u^2 \rangle}{dt}, \quad q = \frac{d \ln \langle \omega^2 \rangle}{dt}, \]

(4.1)

The advantage of the above definition is obvious — the logarithmic derivatives are just the power-law exponents if the energy and enstrophy decay as power laws in time. Moreover, we shall use the following normalized time

\[ \tau = \int_0^t dt \langle \omega^2 \rangle^{1/2}, \]

(4.2)

which can be considered a measure of the number of eddy turnover times undergone by the flow at time t. The normalized time \( \tau \) best represents the time interval over which one expects significant changes in the power-law exponent.

Numerical results are first reported for the energy and enstrophy decay of the dual hexagonal Fourier-Galerkin method (Method I) and the uniform hexagonal Fourier-Galerkin method (Method II) with initial Reynolds number \( R(0) = 8 \) and spatial resolution \( N = 2048^2 \). The logarithmic derivatives \( p \) and \( q \), are plotted versus \( \tau \) in Fig. 8, in the left and the middle columns for Method I and Method II, respectively. It is obvious that, at large times, \( p \) and \( q \) approach \(-2\) and \(-3\), respectively, and the final period of decay solution [7, 8]

\[ \langle u^2 \rangle \propto B_2 (vt)^{-2}, \quad \langle \omega^2 \rangle \propto B_2 (vt)^{-3}, \]

is approached asymptotically in two hexagonal Fourier spectral methods. To make a deep comparison, we plot the logarithmic derivatives \( p \) and \( q \) of the classic square Fourier-Galerkin method (Method III) in the right column. The plots of two hexagonal Fourier-Galerkin methods agree well with those of the classic Fourier-Galerkin method, so that no significant differences can be observed.
Next, we consider some larger initial Reynolds numbers. The numerical experiments are performed corresponding to $k_c=300$, $u_0=1$ and $R(0)=14, 15.73, 18$ in spatial resolution $N=2048^2$. Just as in the classic square Fourier-Galerkin method, the numerical results presented in Fig. 9 demonstrate the existence of a critical Reynolds number $R_c \approx 15.73$ in the two hexagonal Fourier-Galerkin methods such that for $R(0) < R_c$ the Reynolds number decays monotonically in time and for $R(0) > R_c$ the Reynolds number decreases initially, and then increases asymptotically.

Results for the logarithmic derivatives of the energy and enstrophy with $R(0) = R_c$ versus $\tau$ are shown in Fig. 10 which yield an approximate power-law decay of the energy as $t^{-1}$, and an approximate enstrophy decay as $t^{-2}$. An analytical derivation of these power-law exponents in [8] gives

$$\langle u^2 \rangle = \frac{1}{2} v R_c^2 t^{-1}, \quad \langle \omega^2 \rangle = \frac{1}{4} R_c^2 t^{-2},$$

(4.3)

where $R_c' \approx 12.5$ is a time-independent constant which the Reynolds number $R(t)$ approaches at large times. The energy and enstrophy decay for $R(0) = 15.73$ is compared to the analytical prediction (4.3). The simulation results of each Fourier-Galerkin method and analytical solution are in good agreement at large times.
4.2 Decay at high Reynolds numbers

When \( R(0) > R_c \), the Reynolds number \( R(t) \) increases asymptotically, and some unique similarity states exist since all flows with initial Reynolds numbers greater than \( R_c \) presumably approach infinite Reynolds numbers as \( t \to \infty \).

We present in this subsection the results of direct numerical simulations of two-dimensional turbulence decay using the uniform hexagonal Fourier-Galerkin methods with the spatial resolution \( N = 4096^2 \) and sufficiently small time step \( \Delta t \) such that the small scales of the turbulence are adequately resolved. The initial energy spectrum is chosen with \( \bar{u}_0 = 1 \) and with \( k_c \) ranging from 600, 512, 400, 256, 128, 64, 32 to 16 as the initial Reynolds number increases from 32, 64, 128, 256, 512, 1024, 2048 to 4096. Numerical results are compared with those derived from the classic square Fourier-Galerkin method.

The time evolution of the kinetic energy, enstrophy and palinstrophy, normalized using \( \bar{u}_0, \bar{v}_0 \) and \( \bar{l}_0 \), for the different values of \( R(0) \) are shown in Figs. 12-14. Numerical results of all the hexagonal and square Fourier-Galerkin methods state clearly that the energy and enstrophy decay monotonically in time, while the palinstrophy grows initially and finally decays after reaching its peak. The predictions by different numerical experiments agree well with the theoretical analysis in [8].
The logarithmic derivative of the energy $p$, plotted in Fig. 15, is observed to be an increasing function of the initial Reynolds number, which reveals that the energy decay becomes less steep with increasing initial Reynolds numbers. This observation is also rather obvious from the energy decay itself in Fig. 12.

Further, different qualitative behaviours of $p$ are found in [8] for low and high initial Reynolds numbers. For $R(0) \leq 256$, a rapid initial decay of the energy is observed in Fig. 12, which subsequently becomes less steep as time evolves due to the increasing
Reynolds number of the turbulence. This means $p$ decreases to a minimum and then increases in time, just as shown in Fig. 15. However, the flows with high initial Reynolds number ($R(0) \geq 512$) are nearly inviscid in so much that the energy, Figs. 12 and 15, decays very little over the times simulated. Moreover, the magnitude of $p$ is quite small, and the slow decrease of $p$ in time implies that the energy decay is steepening as time evolves.

Meanwhile, no universal decay exponent of the enstrophy, Fig. 16, at large times for all initial Reynolds numbers greater than $R_c$ is observed. However, as Chasnov discovered, the long-time decay exponent for $R(0) > 1024$ appears to change only slightly and the asymptotic decay law for these large Reynolds numbers behaves approximately as $t^{-0.8}$. Besides, since the energy dissipation is proportional to the enstrophy, the change in the qualitative behaviour of the logarithmic derivative of the energy coincides with the power-law exponent of the enstrophy increasing from less than to greater than negative one.

Finally, we emphasize that both plots of the two hexagonal Fourier-Galerkin methods are almost identical to those of the classic square Fourier-Galerkin method, except for the logarithmic derivative of the enstrophy in Fig. 16, where slight differences are observed among the three types of spectral methods.
5 Self-similar energy/enstrophy spectra

5.1 Self-similar energy spectra

The energy spectrum \( E(k,t) \) of two-dimensional turbulence is found to decay self-similarly, i.e., without change of shape.

A self-similar decay of the entire spectrum occurs for the critical initial Reynolds number \( R(0) = R_c \). The time evolution of the dimensionless energy spectrum \( E(k)/\bar{u}^2\bar{l}_0 \) for \( R(0) = 15.73 \) versus dimensionless wave number \( k\bar{l}_0 \) for \( \tau = 0, 5, 10, \ldots, 35 \) is first plotted in Fig. 17. To validate the self-similar decay of the energy spectrum, we define the self-similar energy spectrum \( \hat{E}(\hat{k}) \) using the quantities in (3.4) [8],

\[
E(k,t) = \bar{u}^2 \hat{E}(\hat{k}), \quad \hat{k} = k\bar{l},
\]

and then plot \( \hat{E}(\hat{k}) \) versus \( \hat{k} \) in Fig. 18. A near-perfect collapse of \( \hat{E}(\hat{k}) \) at the different times is observed in each plot in Fig. 18, which indicates a self-similar decay of the spectrum over all wavenumber. The characteristic spectrum \( E(k,t) = B_2(t)k^3 \) is quite obvious at low wave numbers in the plots of the uniform hexagonal Fourier-Galerkin methods as well as that of the classic square Fourier-Galerkin method. While a careful comparison indicates that the uniform hexagonal Fourier-Galerkin method (Method II) gives us the

Figure 17: Evolution of the normalized energy spectrum at time \( \tau = 0, 5, 10, \ldots, 35 \) with \( R(0) = 15.73 \) with \( N = 4096^2 \).

Figure 18: Rescaling of the energy spectrum of Fig. 17 with \( N = 4096^2 \).
most ideal plot. Moreover, an analytic prediction of the low wavenumber coefficient of the spectrum for $R(0) = R_c$ in [8] gives

$$B_2(t) = \nu^3 R_c^2 t,$$

which indicates substantial nonlinear backscatter of energy from small-to-large scale even for this low Reynolds number turbulence.

We further examine the time evolution of the energy spectra for flows with large initial Reynolds number $R(0) > R_c$. The time evolution of the spectra for $R(0) = 64, 256, \text{ and } 4096$ are shown in Fig. 19, Fig. 21 and Fig. 23, respectively. For a better validation of the self-similar energy spectrum, the self-similar spectra obtained using (5.1) are plotted in Figs. 20, 22 and 24, immediately below the corresponding figure for the spectra evolution. It appears from the reasonable collapse of the spectra at different times that the decay of two-dimensional turbulence at large Reynolds number is also self-similar in the energy containing scales.

In Figs. 20, 22 and 24, we have plotted as a dashed line the expected $k^3$ low wave number behavior for two-dimensional turbulence. Moreover, Batchelor pointed that the energy spectrum in fully developed turbulence will acquire the self-similar $E(k) \propto k^{-3}$ over the so-called “inertial range”, i.e., wavenumbers $k$ extending from the “energy-containing” scale wavenumber $k_0$ to the viscous scale wavenumber $k_v \sim \nu^{-1/2}$. An inertial subrange appears to have developed in Figs. 20 and 22 for $R(0) = 256$ and $R(0) = 2048$, respectively.
Figure 21: The energy spectrum at time $t = 0, 10, 20, \cdots, 120$ for $R(0) = 256$ with $N = 4096^2$.

Figure 22: Rescaling of the energy spectra in Fig. 21 with $N = 4096^2$.

Figure 23: The energy spectrum at time $t = 0, 10, 20, \cdots, 120$ for $R(0) = 4096$ with $N = 4096^2$.

Figure 24: Rescaling of the energy spectra in Fig. 23 with $N = 4096^2$. 
which is slightly steeper than the predicted $k^{-3}$ behavior. In particular, when the initial Reynolds number increases to $R(0)=4096$, this inertial subrange (in Fig. 24) developed in the two hexagonal Fourier-Galerkin methods perfectly matches the predicted $k^{-3}$ inertial subrange behaviour for two-dimensional turbulence.

### 5.2 Self-similar enstrophy spectra

Batchelor [2] predicted that the enstrophy spectrum in fully developed turbulence will acquire the self-similar form $\Omega(k) \propto \chi^{2/3}k^{-1}$ (with $\chi = \nu \langle |\nabla \omega|^2 \rangle$) over a range of wavenumbers $k$ extending from the “energy-containing” scales, say around wavenumber $k_0$, to the viscous scales, say around $k_v \sim \nu^{-1/2}$, i.e., over the range $k_0 \ll k \ll k_v$, the so-called “inertial range”. Batchelor’s theory of two-dimensional turbulence is concluded by assuming that there is a finite, non-zero enstrophy dissipation $\chi$ in the limit of infinite $R$. This theory has been successful in describing certain aspects of numerical simulations at high Reynolds numbers. However, Dritschel et al. [10] found that the enstrophy dissipation in fact vanishes for flows with finite vorticity, and the non-zero enstrophy dissipation assumption is not true. After a careful observation together with a mathematical analysis of vanishing $\chi$, Dritschel finally made a simple modification of Batchelor’s theory by replacing the enstrophy spectrum $\chi^{2/3}k^{-1}$ with $\langle \omega^2 \rangle k^{-1}(\ln R)^{-1}$.

We take the initial condition (3.3) with $s=7/2$, $k_c = \sqrt{32/\pi}$ and $\bar{u}_0 = 1$ such that

$$E(k) = \frac{\sqrt{2\pi^2}}{107520} k^8 \exp\left[-8\pi (k/8)^2\right],$$

which peaks at $k_c \approx 3.2$ and is more than $10^{36}$ times smaller by $k = 32$. Such a spectrum is already peaked at low wavenumbers, no significant inverse energy cascade takes place over the short duration of our simulations.

The viscosity is chosen so that the approximate viscous wavenumber occurs at $3/4$ of the maximum effective wavenumber $n/2$ in the classic square Fourier-Galerkin method, i.e., $\nu = 4\pi / (3n/8)^2$. This choice was found to ensure a finite initial vorticity with $||\omega(0)||_\infty \approx 4\pi$ and adequate dissipation at high wavenumbers to resolve the statistics of $\langle (\Delta \omega)^2 \rangle$ and other fine-scale quantities.

Numerical experiments have been performed with the spatial resolution $N = 4096^2$, and the initial Reynolds number $R(0)=5.55 \times 10^4$ (in view of (3.5)) using the two hexagonal Fourier-Galerkin methods. These resolutions are high enough to examine the nature of dissipation in two-dimensional turbulence. Numerical results are, once again, compared with those of the classic Fourier-Galerkin method.

Since the behaviour of the enstrophy dissipation depends on the palinstrophy evolution, we first concentrate on the palinstrophy spectra, which are shown in Fig. 25. Increasingly, a range close to the classical $k^1$ spectrum develops at low wavenumbers. The spectrum has filled out completely by the palinstrophy peak, and afterward the spectrum at moderate to high wavenumbers decays exponentially while preserving its basic shape. It is meaningful to say that the turbulence is “fully developed” by the time the
Figure 25: Palinstrophy spectra $k^2 \Omega(k,t)$ for $N = 4096^2$ at $t = 0$ and at times when the palinstrophy is half its peak value (and growing), at its peak, and half its peak (and decaying). A prediction of $k^1$ is shown as a dashed line. Left: Method I; middle: Method II; right: Method III.

Figure 26: The scaled enstrophy spectra $\Omega(k) \ln R / \langle \omega^2 \rangle$ at the peak enstrophy dissipation time for $N = 1024^2$ (grey thin), $N = 2048^2$ (thin) and $N = 4096^2$ (thick). A reference slope of $k^{-1}$ is shown as a dashed line.

Figure 27: The scaled enstrophy spectra $\Omega(k) \ln R / \langle \omega^2 \rangle$ at the final time for $N = 1024^2$ (grey thin), $N = 2048^2$ (thin) and $N = 4096^2$ (thick). A reference slope of $k^{-1}$ is shown as a dashed line.

Palinstrophy reaches its maximum. Fig. 25 exhibits no obvious distinction between the palinstrophy evolutions of the three cases. While the spectrum at low wavenumbers is much closer to the reference $k^1$ slope in the plot for Method II.

Further the scaled enstrophy spectra $\Omega(k) \ln R / \langle \omega^2 \rangle$ at the peak enstrophy dissipation time and the final time for $N = 1024^2, 2048^2$ and $4096^2$ have been shown in Fig. 26 and Fig. 27, respectively. The curves collapse together over the common inertial ranges. An interpretation of this phenomena is that dissipation is playing an increasingly benign
role as $R \to \infty$. And the $\langle \omega^2 \rangle (k \ln R)^{-1}$ enstrophy spectrum reflects that a fixed amount of enstrophy must spread itself ever more thinly across a widening inertial range as the Reynolds number grows. Once again, there is only a slight difference between the simulation results of the three kinds of Fourier-Galerkin spectral methods. Meanwhile, Method II provides us a slope of the enstrophy spectra over the “inertial range” closer to $k^{-1}$, which demonstrates the modified self-similarity theory of Dritschel.

6 Conclusion

We have proposed two hexagonal Fourier-Galerkin spectral methods for the direct numerical simulation of the two-dimensional homogeneous isotropy freely-decaying turbulence to obtain an even better sampling/approximation efficiency to the band limited/isotropic functions in comparison to the classic rectangular Fourier spectral method. As the soul of the implementation of the approximation scheme, an efficient algorithm for evaluating the de-aliased Fourier coefficients of the nonlinear term is well designed. By adopting some variants of FFTs, our new Fourier spectral methods preserve the high efficiency of the classic Fourier spectral method; and a higher performance can even be attained in our methods. Numerical experiments show that our hexagonal Fourier-Galerkin spectral methods acquire a slightly better statistical results in general than those of the classic rectangular Fourier spectral method, which are in agreement with the corresponding physical theories. Anyway, hexagonal Fourier-Galerkin spectral methods provide alternative choices for the DNS of the two-dimensional isotropy turbulence indeed.

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