A low-dimensional model (LDM) for turbulent Rayleigh-Bénard convection in a Cartesian cell with square domain, based on the Galerkin projection of the Boussinesq equations onto a finite set of empirical eigenfunctions, is presented. The empirical eigenfunctions are obtained from a joint Proper Orthogonal Decomposition (POD) of the velocity and temperature fields using the Snapshot Method on the basis of a direct numerical simulation (DNS). The resulting LDM is a quadratic inhomogeneous system of coupled ordinary differential equations which we use to describe the long-time temporal evolution of the large-scale mode amplitudes for a Rayleigh number of $10^5$ and a Prandtl number of 0.7. The truncation to a finite number of degrees of freedom, that does not exceed a number of 310 for the present, requires the additional implementation of an eddy viscosity–diffusivity to capture the missing dissipation of the small-scale modes. The magnitude of this additional dissipation mechanism is determined by requiring statistical stationarity and a total dissipation that corresponds with the original DNS data. We compare the performance of two models, a constant so-called Heisenberg viscosity–diffusivity and a mode-dependent or modal one. The latter viscosity–diffusivity model turns out to reproduce the large-scale properties of the turbulent convection qualitatively well, even for a model with only a few hundred POD modes.

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I. INTRODUCTION

For most turbulent flows in nature and technology, it is impossible to resolve all relevant degrees of freedom. Systematic methods to derive models with a reduced number of degrees of freedom from the full set of nonlinear fluid equations are thus necessary. Low-dimensional modeling of transient and turbulent flows using Galerkin projection onto the empirical basis functions which are obtained from a proper orthogonal decomposition (POD) is one such well established method. POD and the development of Galerkin models based on POD modes has been applied to a number of fundamental hydrodynamic flow problems, including simple wall-bounded shear flows, flows over cavities or in the wake of a cylinder. The development of low-dimensional models (LDM) based on POD modes has also been extended in several directions such as to the balanced POD method or to unsteady flow problems for which fast and slow flow modes are separated. Most of these cases have been studied for laminar or transitional flows at lower or moderate Reynolds numbers.

With increasing Reynolds number the flows become turbulent, the number of degrees of freedom grows rapidly and their nonlinear couplings are increasingly relevant. The truncation of the set of nonlinear ordinary differential equations (ODE) which follows from Galerkin projection introduces a number of mode interactions and removes couplings between the degrees of freedom which are necessary for the transfer of kinetic energy from large to small scales. An additional dissipation mechanism has to be implemented in the low-dimensional model to account for the dominant dissipation by the truncated degrees of freedom. The particular way of truncation can then alter the dynamics in the LDM.

Several approaches to this problem have been suggested in the past. Aubry et al. used directly the energy transfer between resolved modes and unresolved modes at smaller scales to formulate a spectral closure in the truncated system of ODEs. Moehlis et al. presented streamwise-invariant truncations for a plane Couette flow, and showed that very low-dimensional models with up to ten degrees of freedom can reproduce transient flow phenomena in low-Reynolds-number shear flows. They also found that the detailed behavior of their LDM depends in a subtle manner on the modes included and that a proper account of the symmetries of the system is crucial. Later Smith et al. included streamwise variations in their model. They also introduced a linear damping term, but only when the particular POD mode expansion coefficient $a^{(n)}$ is significantly anti-correlated with its time derivative, $da^{(n)}/dt$. Cazemier et al. constructed a LDM for driven cavity flows, consisting of the 80 most energetic POD modes computed from 700 snapshots of a direct numerical simulation (DNS). To study the time evolution of the truncated ODE system, a slightly different linear damping term is introduced in their model. This term is calculated from the requirement that the energy of the ODE system is conserved in a statistically stationary sense.

A few attempts to derive LDMs are reported for Rayleigh-Bénard (RB) convection, despite being one of the most
comprehensively studied flows.\textsuperscript{17,18} Studies of RB convection in a finite box, based on the POD procedure, have been mostly done by Sirovich and co-workers.\textsuperscript{19–22} Sirovich and Park\textsuperscript{20,21} discussed the importance of the discrete symmetries describing the velocity–temperature fluctuations field. Deane and Sirovich\textsuperscript{22} made a parametric study of the POD mode spectra for small Rayleigh numbers $Ra \leq 46000$.

Only recently, a snapshot method has been applied to turbulent RB convection in a closed cylindrical cell for Rayleigh numbers up to $Ra = 10^8$ and cell aspect ratios between one half and three.\textsuperscript{24} In this work, emphasis was given to relating the first POD modes to the large-scale flow circulation which is always present in a closed turbulent convection cell.\textsuperscript{17} The disentanglement of the temperature and velocity fields into POD modes allowed the authors to quantify the amount of heat which is transported by the particular POD modes through the convection cell. A change of the large-scale flow from a one-roll to a two-roll pattern, which is observed when the aspect ratio is increased beyond one at a fixed Rayleigh number, was in line with a decrease of transported heat by the primary mode compared to the secondary POD mode.

As a correspondence of the few POD studies of RB convection, only a few works exist with an emphasis on developing a LDM by a Galerkin projection of the Boussinesq equations onto the most energetic POD modes. Tarman\textsuperscript{25} derived a model from POD modes which have been however separately extracted from the velocity and temperature fields. In a second work he proposed an algorithm which incorporates the lost dissipation due to truncation.\textsuperscript{26} Besides the cutoff index based on the energy (mode index $k < k_e$), a second index based on the dissipation ($k_d > k_e$) was considered. The time dependence of the modes with indexes $k_e < k \leq k_d$ was expressed as the quotient of the corresponding nonlinear and dissipation coefficient. No closed forms for the constant coefficients in the ODE system were however obtained in any of these works.

In the present work, we want to extend these studies of RB convection in several directions. First, we construct a LDM for the evolution of the POD mode coefficients $a^{(n)}$ in the case of turbulent Rayleigh-Bénard convection in a Cartesian cell with periodic side walls and isothermal free-slip top and bottom square planes. It is essential to use POD modes of the combined four-vector velocity-temperature field.\textsuperscript{24} In this derivation, it turns out that a cubic term due to the interaction of the velocity with the mean temperature field (denoted as $\mathcal{A}$) becomes linear as a consequence of the orthogonality of the POD modes. The other terms which arise in the Galerkin projection are a linear production term $\mathcal{P}$, a linear dissipation term $\mathcal{D}$, a quadratic nonlinear term $N$, and a constant term $\epsilon(T)$ corresponding to the dissipation due to the mean temperature field. Second, our studies will extend previous works\textsuperscript{21,22,26} in terms of the magnitude of the Rayleigh number of convection. A case with $Ra \sim 10^5$ is considered for which RB convection is turbulent and a DNS data record exists. Third, we are interested in the long-time behavior of the dynamics in the LDM. With a view to more complex convection flows in the future, we are seeking for the least set of POD modes that can reproduce characteristic dynamics of turbulent convection.

A solution which includes the additional dissipation due to the neglected less energetic POD modes has to be considered by an additional eddy viscosity–diffusivity, $\eta \geq 0$. First, we present the so-called Heisenberg model with a constant $\eta$ which exerts the same fraction of dissipation on all POD modes. As will be shown, this closure requires at least a minimum number of degrees of freedom, in particular with respect to the vertical direction, for a qualitatively correct description of the flow. As a consequence, two LDMs with 210 and 310 degrees of freedom, respectively, are chosen. They are taken from a set of 15708 modes (see Sec. III B). As will be seen, this model fails to reproduce the large-scale evolution of convection. For the larger of the two sets of modes, the model relaxes to a statistically stationary state which contains too much energy. Second, we refine this model and include a mode–dependent (or modal) eddy viscosity–diffusivity. The magnitude of both eddy viscosity–diffusivity contributions has to be estimated. In order to do so, we will follow a procedure that has been suggested by Cazemier et al.\textsuperscript{27} The second model yields much more realistic large-scale variations of the most energetic modes, also reproducing with reasonable accuracy the energy spectrum and the turbulence statistics. Therefore, a significant part of the present work discusses the impact of both types of eddy viscosity-diffusivity $\eta$ on the dynamics of the LDM with different number of degrees of freedom and how it compares to DNS.

The outline of the paper is as follows. The equations of motion, the basic idea of POD – in particular for the method of snapshots – is discussed in the next section. The construction of the LDM by Galerkin projection onto POD modes of RB convection follows in Sec. III. In this section, the results of the time integration of the LDM with both eddy viscosity–diffusivity schemes, and the agreement with the DNS are also discussed. We conclude with a summary and give an outlook.
II. METHODS

A. Equations of motion and numerical scheme

Turbulent Rayleigh-Bénard convection is governed by the Boussinesq equations. They are brought into a dimensionless form by rescaling with the domain height $l_z$, the diffusive time scale $t_{\text{diff}} = l_z^2/\kappa$ with $\kappa$ being the thermal diffusivity, the temperature difference $\Delta T = \Theta_{\text{bottom}} - \Theta_{\text{top}} > 0$ and follow to

$$\nabla \cdot \mathbf{u} = 0$$

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} = -\nabla p + Pr\nabla^2 \mathbf{u} + RaPrT \mathbf{e}_z$$

$$\frac{\partial T}{\partial t} + (\mathbf{u} \cdot \nabla) T = w + \nabla^2 T$$

where $\mathbf{u}(x,t)$ is the velocity field, $T(x,t)$ the departure from the linear conduction temperature profile, and $p(x,t)$ is the kinematic pressure. Dimensionless parameters are the Rayleigh number $Ra = g\alpha\Delta T l_z^3/(\nu\kappa)$ and the Prandtl number $Pr = \nu/\kappa$. Besides diffusivity $\kappa$, they contain the kinematic viscosity $\nu$, the gravitational acceleration $g$, and the thermal expansion coefficient $\alpha$. Note that the total temperature field is given in our notation by (see also Ref.20)

$$\Theta(x,t) = \Theta_{\text{bottom}} - \frac{\Delta T}{l_z} z + T(x,t).$$

The vector $\mathbf{e}_z$ is the direction in which buoyancy and gravity work and in which the mean temperature gradient is established. The dimensions of the cell $\Omega$ are $l_x = l_y = 4\pi, l_z = \pi$, where from now on $x, y$ are the horizontal and $z$ the vertical dimensionless coordinates. The aspect ratio is fixed to $l_x/l_z = l_y/l_z = 4$. For convenience, the origin of the coordinate system is in the center of the cell. Therefore, $x \in [-L_x/2, L_x/2], y \in [-L_y/2, L_y/2], z \in [-1/2, 1/2]$ with $L_x = l_x/l_z$ and $L_y = l_y/l_z$. The $x$-, $y$- and $z$-components of the velocity field will be denoted by $u, v$ and $w$, respectively. The boundary conditions are periodic in $x$ and $y$, and free-slip in $z$. This means that at the hot bottom plane at $z/l_z = -1/2$ and the cold top plane at $z/l_z = 1/2$ the following conditions hold:

$$w = T = \frac{\partial u}{\partial z} = \frac{\partial v}{\partial z} = 0.$$  

For the present boundary conditions, the flow can be decomposed in

$$u(x,t) = \langle u(z) \rangle + u'(x,t)$$  

$$v(x,t) = \langle v(z) \rangle + v'(x,t)$$  

$$w(x,t) = \langle w(z) \rangle + w'(x,t)$$  

$$T(x,t) = \langle T(z) \rangle + \theta(x,t)$$

where the mean components, e.g. $\langle u(z) \rangle$ are ensemble averages obtained by averaging over the horizontal $x$-$y$ plane and time, i.e., a sequence of $N_f$ statistically independent snapshots. The ensemble average for the velocity component $u$ is thus given by

$$\langle u(z) \rangle = \frac{1}{N_f} \sum_{n=1}^{N_f} \left( \frac{1}{L_x L_y} \int_{-L_x/2}^{L_x/2} \int_{-L_y/2}^{L_y/2} u(x,y,z,t_n) dy dx \right).$$

Figure II shows the mean vertical profiles of $\Theta$ and $T$ together with the linear thermal conduction profile. Considering the fact that for our problem, due to symmetry considerations, $\langle u \rangle = \langle v \rangle = \langle w \rangle = 0$, we can take the four-vector field

$$\mathbf{v}(x,t) = (u,v,w,\theta)$$

for the POD analysis and LDM derivation. Our analysis is in the statistically stationary regime of convective turbulence. The ensemble average of 8 yields then an expression linking the mean and the fluctuating components of the flow in the form

$$\frac{d^2\langle T \rangle}{dz^2} = \frac{d\langle w\theta \rangle}{dz}$$
FIG. 1: Mean vertical profiles $\langle T(z) \rangle$, $\langle \Theta(z) \rangle$ and the linear conduction profile (see Eq. (11)). Data are obtained from the DNS data record.

Following Sirovich et al.\textsuperscript{19} temperatures and velocities are rescaled with $\hat{\theta}_c = \sqrt{2Nu^3/Ra Pr}$ and $\hat{u}_c = \sqrt{Ra Pr/2Nu}$, respectively. Here $Nu$ is the Nusselt number. The quantity $\hat{\theta}_c$ is obtained by demanding that the turbulent heat flux in the center of the cell must be equal to that due to diffusion at the boundary.

The highly-resolved data record is obtained by a pseudo-spectral DNS which uses fast Fourier transformations.\textsuperscript{27,28} Time stepping is done by a second-order Runge Kutta scheme. For most of the work, we consider a data set with $N' = 320$ full three-dimensional turbulence snapshots which are separated by four convective time units $t_{\text{conv}} = \sqrt{l_z/g\alpha \Delta T}$ from each other. The computational grid consists of $N_x \times N_y \times N_z = 256 \times 256 \times 65$ points in $x$-, $y$-, and $z$-directions, respectively. The spectral resolution is given by $k_{\text{max}}\eta_K = 5.6$. Here, $k_{\text{max}} = \sqrt{2Nz/3}$ and $\eta_K$ the Kolmogorov dissipation length. The Rayleigh number is $Ra = 1.03 \times 10^5$ and the Prandtl number $Pr = 0.7$.

B. Proper Orthogonal Decomposition (POD)

The POD is a model reduction technique that extracts the most energetic modes from a set of realizations or snapshots of the flow. These POD modes are used as a basis for Galerkin projections of the full set of nonlinear equations thus reducing the infinite-dimensional space of solutions to a finite-dimensional system.\textsuperscript{2,29} The two-point correlation tensor or covariance matrix of the four-vector field is defined by

$$K_{mn}(x, x') = \langle v_m(x, t)v_n^*(x', t) \rangle_t$$  \hspace{1cm} (10)

where the asterisk denotes the complex conjugate, $\langle \cdot \rangle_t$ the time average and $m, n = 1, 2, 3, 4$. For Rayleigh-Bénard convection in Cartesian domains with two homogeneous (invariant with respect to translations) directions Eq. (10) takes the form\textsuperscript{30}

$$K_{mn}(x, x') = K_{mn}(x - x', y - y', z)$$  \hspace{1cm} (11)

For a kernel (11), the eigenfunctions have the form

$$\phi^{(p)}_{m; n_x, n_y}(x, y, z) = \frac{\phi^{(p)}_{m; n_x, n_y}(z)}{\sqrt{L_xL_y}} \exp \left( \frac{2\pi in_x x}{L_x} + \frac{2\pi in_y y}{L_y} \right)$$  \hspace{1cm} (12)

where $n_x, n_y$ are integers for the $x$– and $y$– directions, respectively. The superscript $(p)$ denotes a particular POD mode. The determination of $\phi$ follows then from

$$\sum_{n=1}^{4} \int_{-1/2}^{1/2} \kappa_{mn}(n_x, n_y; z, z') \phi^{(p)}_{m; n_x, n_y}(z') dz' = \lambda^{(p)}_{n_x, n_y} \phi^{(p)}_{m; n_x, n_y}(z)$$  \hspace{1cm} (13)

where $\kappa_{mn}$ is the Fourier transform of $K_{mn}$ with respect to the homogeneous directions $x$ and $y$. The kernel $\kappa_{mn}$ is calculated from the numerical data set by first taking the discrete Fourier transform of each realization in the
The associated expansion of the velocity field $v$ due to the reality of the physical space fields a complete set of vector eigenfunctions \{\phi^{(p)}_{m,n_x,n_y}(z)\}_{p=1}^\infty given by Eq. (12) implies that thus the kernel $\kappa_{m,n}$ is Hermitian, non-negative and on physical grounds square integrable, such that the existence of a complete set of vector eigenfunctions \{\phi^{(p)}_{m,n_x,n_y}(z)\}_{p=1}^\infty is assured. Complex conjugation of Eq. (18) and use of Eq. (12) implies that

$$\phi^{(p)}_{m,n_x,n_y}(z) = \phi^{(p)*}_{m,-n_x,-n_y}(z).$$

Due to the reality of the physical space fields $v_m(x, t)$ Eq. (14) implies that

$$F_m^{*}(n_x, n_y; z, t) = F_m(-n_x, -n_y; z, t).$$

The associated expansion of the velocity field $v_m(x, t)$ in terms of the modes is given as

$$v_m(x, t) = \sum_p \sum_{n_x} \sum_{n_y} (a_{n_x,n_y}^{(p)}(t) \exp \left( \frac{2\pi in_x x}{L_x} + \frac{2\pi in_y y}{L_y} \right) \phi^{(p)}_{m,n_x,n_y}(z),$$

and again reality of the four-vector field implies that $a_{n_x,n_y}^{(p)}(t) = a_{n_x,n_y}^{(p)*}(t)$. The index $(p)$ runs over the POD modes. The coefficients are calculated by the scalar product in $L^2(\Omega)$,

$$a_{n_x,n_y}^{(p)}(t) = \langle \Phi^{(p)}_{m,n_x,n_y}(x, y, z), v_m(x, y, z, t) \rangle = \sum_{m=1}^M \int_{L_x/2}^{L_x/2} \int_{L_y/2}^{L_y/2} \int_{-1/2}^{1/2} \int_{-1/2}^{1/2} \Phi^{(p)*}_{m,n_x,n_y}(x, y, z)v_m(x, y, z, t) dx dy dz.$$

Next, the discrete Fourier transform of $v_m(x, t)$ is introduced together with the ansatz of $\Phi$ in Eq. (12) and the orthogonality of the complex exponentials. This gives

$$a_{n_x,n_y}^{(p)}(t) = \sqrt{L_x L_y} \sum_{m=1}^M \int_{-1/2}^{1/2} \int_{-1/2}^{1/2} \phi^{(p)*}_{m,n_x,n_y}(z) F_m(n_x, n_y; z, t) dz.$$

C. The method of snapshots

The snapshot method is one way to obtain the POD modes, particularly when the computational grid size becomes large. The time coordinate $t$ in the equations above has to be substituted now by an index that runs over the sequence of snapshots. It is based on the fact that Eq. (15) is a degenerate kernel. Consequently, an eigenfunction of the kernel $\kappa_{m,n}$ can be represented as

$$\phi^{(p)}_{m,n_x,n_y}(z) = \sum_{l=1}^{N_T} \sum_{\gamma \in G} \gamma \cdot \alpha^{(p)}(n_x, n_y, l) \gamma \cdot F_m(n_x, n_y; z, l),$$

$$= \sum_{k=1}^{N_T} \alpha^{(p)}(n_x, n_y, k) F_m(n_x, n_y; z, k),$$

where $\gamma$ as an element of the symmetry group $G = \{G, ZG\}$ as discussed in Appendix A. The explicit use of symmetries in the problem at hand enlarges the data record with originally $N_T$ snapshots to a total number of $N_T' = 16 N_T$ snapshots thus improving the convergence. For $N_T' = 320$ DNS snapshots we thus end up with $N_T = 5120$ samples that can be used to evaluate the POD modes. Replacing the kernel in Eq. (13) and using Eq. (21) results to

$$\frac{L_x L_y}{N_T} \sum_{m=1}^{N_T} \left( \int_{-1/2}^{1/2} \int_{-1/2}^{1/2} \sum_{j=1}^{4} \sum_{l=1}^{4} F_j^{(p)}(n_x, n_y; z', k) F_j(n_x, n_y; z', m) dz' \right) \alpha^{(p)}(n_x, n_y, m)$$

$$= \lambda^{(p)}_{n_x,n_y} \alpha^{(p)}(n_x, n_y, k),$$

(22)
where \( k, m = 1, 2, ..., N_T \) represent any two snapshots (including all possible symmetries). Then [22] is the matrix problem which yields the eigenvalues \( \lambda \) and eigenfunctions \( \phi \). It is clear that it determines just \( N_T \) of the empirical eigenfunctions for a fixed tuple \((n_x, n_y)\). The eigenvalue \( \lambda_{n_x,n_y}^{(p)} \) of the \( N_T \times N_T \) matrix is the total energy (kinetic energy plus temperature variance) of the \( p \)th POD mode for \((n_x, n_y)\). Recall that in the present case the four-velocity field is expanded into Fourier modes with respect to \( x \) and \( y \) which is characterized by wavenumbers \( n_x \) and \( n_y \), respectively. This results theoretically in an infinite set of POD modes.

### III. RESULTS

#### A. Galerkin Projection of the Boussinesq equations onto the POD modes

Given the full set of nonlinear Boussinesq equations and the POD modes extracted by a snapshot method from the DNS data, we can proceed to derive the LDM. This requires first a Galerkin projection step. Using the dimensionless units introduced in section IIIA, Eqs. (2) and (3) can be rewritten together in a four-vector notation with respect to \( v \),

\[
\frac{\partial v_i}{\partial t} = -\nu_c \sum_{j=1}^{3} v_j \frac{\partial v_i}{\partial x_j} + \frac{Pr}{1 + \delta_{4}(Pr-1)} \sum_{j=1}^{3} \frac{\partial^2 v_i}{\partial x_j^2} + 2\delta_{13} N u^2 v_4 + \delta_{44} \frac{\dot{u}_c}{\theta_c} v_3 + \delta_{44} \left( \sum_{j=1}^{3} \frac{\partial^2 (T)}{\partial x_j^2} - \dot{u}_c \sum_{j=1}^{3} v_j \frac{\partial (T)}{\partial x_j} - s_i \right) + s_i \tag{23}
\]

where \( i = 1, 2, 3, 4 \). Here, \( j = 1, 2, 3 \) correspond to the three spatial coordinates and the term

\[
s_i = 2\delta_{13} N u^2 \langle T \rangle - \frac{1}{u_c} \frac{\partial p}{\partial x_i} \tag{24}
\]

is a source which drops out in the Galerkin projection procedure. This is due to the divergence-free nature of the POD basis functions and the fact that modes \( \Phi_{m,p_x,p_y}^{(p)}(x, y, z) = 0 \), \( \forall p \), respectively. Now one takes the inner product of [23] with modes \( \Phi_{m,p_x,p_y}^{(p)}(x, y, z) \) and inserts the expansion \( v_m(x, t) = \sum_{n_x} \sum_{n_y} \sum_{a_m(x, y, t)} \Phi_{m,n_x,n_y}(x, y, z) \) from equation (18). Due to the orthogonality of the POD modes, the following infinite-dimensional ODE system follows

\[
\hat{a}_{p_x,p_y}^{(p)} = \sum_{n=1}^{\infty} \phi(p_x, p_y, n, p) a_{p_x,p_y}^{(n)} + \sum_{n=1}^{\infty} \epsilon(p_x, p_y, n, p) a_{p_x,p_y}^{(n)} + N(p_x, p_y, p) \sum_{n=1}^{\infty} \phi(p_x, p_y, n, p) a_{p_x,p_y}^{(n)} + s_{p_x,p_y}^{(p)} \tag{25}
\]

The terms on the right hand side of (25) correspond to the production \( \langle \phi \rangle \), dissipation \( \epsilon \), nonlinear transfer \( N \), interaction with the mean flow \( (3) \), and the dissipation due to the mean temperature field \( \langle \epsilon(T) \rangle \), respectively. Closed forms for \( \epsilon \) and \( N \) have been obtained in Refs. [20] and [34], but only for the velocity field. The general equations for the terms on the right hand side of (25) are as follows. The production term is given by

\[
\varphi(p_x, p_y, n, p) = \int_{-1/2}^{1/2} \left( 2 N u^2 \phi_{4,p_x,p_y}^{(n)} \phi_{3,p_x,p_y}^{(p)*} + \frac{\dot{u}_c}{\theta_c} \phi_{3,p_x,p_y}^{(n)} \phi_{4,p_x,p_y}^{(p)*} \right) dz \tag{26}
\]

and the dissipation term by

\[
\epsilon(p_x, p_y, n, p) = -Pr \sum_{j=1}^{3} \left\{ \left[ \left( \frac{2 \pi p_x}{L_x} \right)^2 + \left( \frac{2 \pi p_y}{L_y} \right)^2 \right] \int_{-1/2}^{1/2} \phi_{4,p_x,p_y}^{(n)} \phi_{j,p_x,p_y}^{(p)*} dz + \int_{-1/2}^{1/2} \phi_{4,p_x,p_y}^{(n)} \phi_{j,p_x,p_y}^{(p)*} dz \right\} - \left\{ \left[ \left( \frac{2 \pi p_x}{L_x} \right)^2 + \left( \frac{2 \pi p_y}{L_y} \right)^2 \right] \int_{-1/2}^{1/2} \phi_{4,p_x,p_y}^{(n)} \phi_{4,p_x,p_y}^{(p)*} dz + \int_{-1/2}^{1/2} \phi_{4,p_x,p_y}^{(n)} \phi_{4,p_x,p_y}^{(p)*} dz \right\}. \tag{27}
\]
The nonlinear mode coupling term $N$ is given by

$$N(p_x, p_y, p) = \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} \sum_{m_{xy}=-\infty}^{\infty} \sum_{m_{xy}=-\infty}^{\infty} \hat{B}^{(m,n,p)}(m_x, m_y, p_x, p_y) a^{(m)}_{m_{xy}, m_y} a^{(n)}_{p_x-m_{xy}, p_y-m_y}, \tag{28}$$

where the coefficients are given by

$$\hat{B}^{(m,n,p)}(m_x, m_y, p_x, p_y) = -\frac{\hat{u}_c}{\sqrt{L_x L_y}} \sum_{j=1}^{4} \int_{-1/2}^{1/2} \frac{2\pi im_x}{L_x} \phi^{(n)}_{3j,m_{xy},m_y} + \frac{2\pi im_y}{L_y} \phi^{(m)}_{2j,m_{xy}+m_y} \phi^{(n)}_{3j+m_{xy},m_y} + \phi^{(n)}_{3j,m_{xy}+m_y} \phi^{(m)}_{3j+m_{xy},m_y} \phi^{(n)}_{2m_{xy}+m_y} dz. \tag{29}$$

The last two terms in Eq. \((25)\), $\Im$ and $\epsilon(T)$, are calculated for a quasi-steady flow. The term $\Im$ is related to the ensemble average of temperature, $\mathbf{u} \cdot \nabla \langle T \rangle = w d\langle T \rangle/dz$. However, from the ensemble average of $\hat{w}$ we obtain

$$\frac{d\langle T \rangle}{dz}(z) = \hat{u}_c \langle w\theta(z) \rangle + \frac{d\langle T \rangle}{dz}(z = -1/2) \tag{30}$$

where $\frac{d\langle T \rangle}{dz}(z = -1/2) = -\hat{u}_c \int_{-1/2}^{1/2} \langle w\theta \rangle dz$. In terms of the eigenvectors,

$$\langle w\theta \rangle = \frac{1}{L_x L_y} \sum_{k=1}^{\infty} \sum_{l=1}^{\infty} \sum_{k_y=-\infty}^{\infty} \sum_{k_y=-\infty}^{\infty} \langle a^{(k)}_{k_x,k_y}(t)a^{(l)}_{-k_x,-k_y}(t) \rangle \phi^{(k)}_{3k_x,k_y} \phi^{(l)}_{4-k_x,-k_y} \tag{31}$$

since $\langle w\theta \rangle$ is only a function of $z$. However, $\langle a^{(k)}_{k_x,k_y}(t)a^{(l)}_{-k_x,-k_y}(t) \rangle = \langle a^{(k)}_{k_x,k_y}(t) a^{(l)*}_{k_x,k_y}(t) \rangle = \delta_{kl} \lambda_x^{(k)} \phi^{(k)}_{3k_x,k_y} \phi^{(l)*}_{4-k_x,-k_y} \phi^{(l)*}_{4-k_x,k_y} \tag{32}$

Consequently, the equation for the term $\Im$ is

$$\Im(p_x, p_y, n, p) = -\hat{u}_c^{2} \int_{-1/2}^{1/2} \left( \langle w\theta \rangle - \int_{-1/2}^{1/2} \langle w\theta \rangle dz \right) d\langle T \rangle d\langle T \rangle dz \tag{34}$$

where $\langle w\theta \rangle$ is given by $\langle w\theta \rangle$. For the low-dimensional description of free-shear-flow, Rajac et al.\cite{31} keeps the original time dependence in $\langle w\theta \rangle$, just replacing $\lambda_x^{(k)} = \langle |a^{(k)}_{k_x,k_y}|^2 \rangle$ and considering this as a running-time-average factor, so the resulting $\sum_{n=1}^{\infty} \Im(k_x, k_y, n, k) a^{(n)}_{k_x,k_y}$ term becomes cubic in the ODE system $\hat{w} = \hat{w}$. Finally, we have a dissipation term which is related to the mean temperature profile, $\nabla^2 \langle T \rangle$. The term is nonzero only for the purely thermal and real modes $\phi^{(n)}_{40,0}$. In the present case this corresponds to a maximum energy of $\lambda^{(3)}_{0,0} = 0.1989\%$. The term is given by

$$\epsilon(T)(p_x, p_y, p) = \frac{\hat{u}_c}{\sqrt{L_x L_y}} \sum_{k=1}^{\infty} \sum_{k_y=-\infty}^{\infty} \sum_{k_y=-\infty}^{\infty} \lambda^{(k)}_{k_x,k_y} \int_{-1/2}^{1/2} \phi^{(k)}_{3k_x,k_y} \phi^{(l)*}_{4-k_x,k_y} \phi^{(l)*}_{4-k_x,k_y} dz \tag{35}$$

only if $p_x = p_y = 0$, otherwise $\epsilon(T)(p_x, p_y, p) = 0$. In Eq. \((35)\), we used again $(\cdot)' = d(\cdot)/dz$. This completes the discussion of the different terms that arise due to the Galerkin projection of the Boussinesq equations onto the POD modes.
FIG. 2: (Color online) Reconstruction of the vertical profiles of mean convective heat flux (top) and temperature deviation from the linear profile (bottom). Data from DNS are compared with the two LDMs, M1 & M2, as well as with the complete set of POD modes ($M = 15708$) obtained from the snapshot analysis.

FIG. 3: Estimation of the maximum eddy viscosity and diffusivity following the procedure of Ref. 8. (Top) Ratio $D_i/\epsilon_i$ for the LDM model M2 with 310 POD modes. The dashed line marks the maximum of the ratio. (Bottom) $D_i$ and $\epsilon_i$ are separately shown for the same data.
The integrals along the $z$-axis contained in the coefficients of Eq. (25) are evaluated on the computational grid of the DNS. Since the integrands are discrete functions of the wave and quantum numbers, we include all degrees of freedom with $0 \leq n_x \leq 16$, $-16 \leq n_y \leq 16$, and $1 \leq (n) \leq 28$. This results in a maximum number of POD modes of $M = 15708$. Out of this set of POD modes, we select small subsets of the most energetic POD modes which corresponds to the large-scale structures of the convection dynamics.

The choice of $M = 210$ for our first LDM denoted as M1 follows from the restriction to modes with $|n_x| + |n_y| \leq 4$ and $1 \leq (n) \leq 10$ as indicated in Tab. I. This level of truncation builds on experiences from similar studies by Aubry et al., Moehlis et al., Holmes et al. and Podvivak in wall-bounded shear flows. They reproduced successfully the dynamics close to the walls for a range of values of the so-called Heisenberg parameter $\eta$ (which will be discussed below). Modes with $n_x = 0$ in streamwise and $-5 \leq n_y \leq 5$ in spanwise direction were taken. As explained by Holmes et al., this choice is due to the fact that for the given spanwise domain length, the cross-stream interactions that contribute to the observed bursts of the velocity are well reproduced with at least five nonzero modes. Moreover, when we choose small quantum numbers, $(n) \leq 5$, the present LDM for convection yields solutions which decay monotonically with time. Our model and the resulting degeneracy restrictions for the average field–modes $\lambda_{0,0}^{(n)}$ did not allow us to take $(n) < 8$. A significant improvement of the LDM dynamics is obtained for $(n) \leq 10$. This is due to the fact that the two additional average field–modes with $\lambda_{0,0}^{(9)} = \lambda_{0,0}^{(10)}$ have a degeneracy of 2 and will be incorporated. A second LDM called M2 was introduced with range of horizontal wave numbers $|n_x| + |n_y| \leq 5$ (see Tab. I). The latter LDM will be used for most of the following studies.

Figure 2 represents the Reynolds shear stress $\langle w\theta(z) \rangle$ and the average temperature profile $\langle T(z) \rangle$ as reconstructed from the two LDMs. The calculation of both profiles is done by integration of Eqs. (36) and (37), respectively and explained in section III.A. As can be seen in the figure, the convergence to the vertical DNS profile for $\langle w\theta(z) \rangle$ is very slow. Only the significant enhancement of the degrees of freedom up to $M = 15708$ results in an excellent agreement with the DNS profiles, for the present Rayleigh and Prandtl numbers.

The modes of the LDMs are however not the modes which contribute dominantly to thermal and kinetic energy dissipation. The missing couplings to the small-scale dissipating modes causes numerical stability problems of the LDM. Various methods have been proposed therefore to stabilize the truncated low-dimensional dynamical system as we have discussed in the introduction. Our studies showed that the model introduced by Cazemier et al. worked best. This model introduces a closure based on the mean energy balance as derived from Eq. (25) which can be rewritten as

$$
\dot{a}_{p_x,p_y}^{(p)} = \sum_{n=1}^{M_Q} A(p_x,p_y,n,p) a_{p_x,p_y}^{(n)} + N(p_x,p_y,p) + \epsilon_r(p_x,p_y,p) + \epsilon_T(p_x,p_y,p) + \epsilon_S(p_x,p_y,n,p).
$$

(36)

where $M_Q$ is the highest quantum number in the LDM ($M_Q \leq 28$). The three linear terms in Eq. (25) are summarized to

$$
A(p_x,p_y,n,p) = \varphi(p_x,p_y,n,p) + \epsilon(p_x,p_y,n,p) + \epsilon_T(p_x,p_y,n,p).
$$

(37)

From Eq. (36), one can derive an equation for the total energy by multiplication with $a_{p_x,p_y}^{(n)}$ and summation over all quantum and wavenumbers. An additional linear damping term, $D(p_x,p_y,p)$, is then quantitatively determined from the requirement that the mean total energy of the extended dynamical system is in a statistically stationary state,

| Model | $M$ | $|n_x| + |n_y|$ | $(n)$ | % Energy |
|-------|-----|----------------|------|----------|
| M1    | 210 | $4 \leq (n) \leq 10$ | 76.894 |
| M2    | 310 | $5 \leq (n) \leq 10$ | 82.202 |
| Total | 15708 | $0 \leq n_x \leq 16$ | 100 |
|       |     | $-16 \leq n_y \leq 16$ |      |

TABLE I: Parameters of the two LDMs denoted as M1 and M2. We list the range of the horizontal wavenumbers $n_x$ and $n_y$ and the so-called quantum number $(n)$. In the last two lines of the table, the total number of POD modes that has been calculated is given. It follows from the maximum range of horizontal wave and vertical quantum numbers. The horizontal wavenumber in $x$-direction starts from zero due to symmetry of Eq. (16). The total number of POD modes from the snapshot analysis is thus $M = 17 \times 33 \times 28 = 15708$. The latter LDM will be used for most of the following studies.

B. Truncation to a low-dimensional model

The integrals along the $z$-axis contained in the coefficients of Eq. (25) are evaluated on the computational grid of the DNS. Since the integrands are discrete functions of the wave and quantum numbers, we include all degrees of freedom with $0 \leq n_x \leq 16$, $-16 \leq n_y \leq 16$, and $1 \leq (n) \leq 28$. This results in a maximum number of POD modes of $M = 15708$. Out of this set of POD modes, we select small subsets of the most energetic POD modes which corresponds to the large-scale structures of the convection dynamics.

The choice of $M = 210$ for our first LDM denoted as M1 follows from the restriction to modes with $|n_x| + |n_y| \leq 4$ and $1 \leq (n) \leq 10$ as indicated in Tab. I. This level of truncation builds on experiences from similar studies by Aubry et al., Moehlis et al., Holmes et al. and Podvivak in wall-bounded shear flows. They reproduced successfully the dynamics close to the walls for a range of values of the so-called Heisenberg parameter $\eta$ (which will be discussed below). Modes with $n_x = 0$ in streamwise and $-5 \leq n_y \leq 5$ in spanwise direction were taken. As explained by Holmes et al., this choice is due to the fact that for the given spanwise domain length, the cross-stream interactions that contribute to the observed bursts of the velocity are well reproduced with at least five nonzero modes. Moreover, when we choose small quantum numbers, $(n) \leq 5$, the present LDM for convection yields solutions which decay monotonically with time. Our model and the resulting degeneracy restrictions for the average field–modes $\lambda_{0,0}^{(n)}$ did not allow us to take $(n) < 8$. A significant improvement of the LDM dynamics is obtained for $(n) \leq 10$. This is due to the fact that the two additional average field–modes with $\lambda_{0,0}^{(9)} = \lambda_{0,0}^{(10)}$ have a degeneracy of 2 and will be incorporated. A second LDM called M2 was introduced with range of horizontal wave numbers $|n_x| + |n_y| \leq 5$ (see Tab. I). The latter LDM will be used for most of the following studies.

Figure 2 represents the Reynolds shear stress $\langle w\theta(z) \rangle$ and the average temperature profile $\langle T(z) \rangle$ as reconstructed from the two LDMs. The calculation of both profiles is done by integration of Eqs. (36) and (37), respectively and explained in section III.A. As can be seen in the figure, the convergence to the vertical DNS profile for $\langle w\theta(z) \rangle$ is very slow. Only the significant enhancement of the degrees of freedom up to $M = 15708$ results in an excellent agreement with the DNS profiles, for the present Rayleigh and Prandtl numbers.

The modes of the LDMs are however not the modes which contribute dominantly to thermal and kinetic energy dissipation. The missing couplings to the small-scale dissipating modes causes numerical stability problems of the LDM. Various methods have been proposed therefore to stabilize the truncated low-dimensional dynamical system as we have discussed in the introduction. Our studies showed that the model introduced by Cazemier et al. worked best. This model introduces a closure based on the mean energy balance as derived from Eq. (25) which can be rewritten as

$$
\dot{a}_{p_x,p_y}^{(p)} = \sum_{n=1}^{M_Q} A(p_x,p_y,n,p) a_{p_x,p_y}^{(n)} + N(p_x,p_y,p) + \epsilon_r(p_x,p_y,p) + \epsilon_T(p_x,p_y,p) + \epsilon_S(p_x,p_y,n,p).
$$

(36)

where $M_Q$ is the highest quantum number in the LDM ($M_Q \leq 28$). The three linear terms in Eq. (25) are summarized to

$$
A(p_x,p_y,n,p) = \varphi(p_x,p_y,n,p) + \epsilon(p_x,p_y,n,p) + \epsilon_T(p_x,p_y,n,p).
$$

(37)

From Eq. (36), one can derive an equation for the total energy by multiplication with $a_{p_x,p_y}^{(n)}$ and summation over all quantum and wavenumbers. An additional linear damping term, $D(p_x,p_y,p)$, is then quantitatively determined from the requirement that the mean total energy of the extended dynamical system is in a statistically stationary state,
i.e.,

\[
0 = \sum_{n,m,x,y} \hat{B}(m,n,p)(m_x,m_y,p_x,p_y)\{a^{(m)}_{m_x,m_y}a^{(n)}_{p_x-m_x,p_y-m_y}a^{(p)*}_{p_x,p_y}\} \\
+ \left( A(p_x,p_y,p,p) + D(p_x,p_y,p) \right)\{a^{p}_{p_x,p_y}a^{p*}_{p_x,p_y}\} + \epsilon(T)(p_x,p_y,p)\{a^{(p)*}_{p_x,p_y}\},
\]

(38)

Note, that in this real equation, the last term on the right hand side is nonzero only for the purely thermal modes with \(p_x = p_y = 0\). Also, due to orthogonality, the last two indices of \(A\) have to be equal.

In the so-called Heisenberg dissipation model by Aubry et al. the action of the neglected modes on the ones contained in the LDM is represented in an average sense, namely as a function of the dynamics of these coherent structures. An even simpler approach was chosen by Omurtag and Sirovich. They simply introduced a constant empirical viscosity coefficient for turbulent channel flows, which has a similar effect as the Heisenberg eddy viscosity of Aubry and co-workers. First, we will apply here the constant eddy viscosity–diffusivity, but estimate the magnitude of \(\eta\) by the method of Cazemier et al. as described above. Since \(Pr = 0.7\) and thus \(\nu \approx \kappa\), we also use the same amplitudes of \(\eta\) for all fields. Equations (25) follow to

\[
\hat{a}^{(p)}_{p_x,p_y} = \sum_{n=1}^{M_Q} \varphi(p_x,p_y,n,p)a^{(n)}_{p_x,p_y} + (1 + \eta) \sum_{n=1}^{M_Q} \epsilon(p_x,p_y,n,p)a^{(n)}_{p_x,p_y} + \\
N(p_x,p_y,p) + \sum_{n=1}^{M_Q} \Theta(p_x,p_y,n,p)a^{(n)}_{p_x,p_y} + \epsilon(T)(p_x,p_y,p)
\]

(39)

after truncation and addition of an eddy viscosity and diffusivity term

\[
\eta = \alpha \tilde{\eta} \quad \text{with} \quad \alpha \in \mathbb{R}.
\]

(40)

The constant eddy viscosity-diffusivity \(\tilde{\eta}\) is determined by

\[
\tilde{\eta} = \max \left[ \frac{D(p_x,p_y,p)}{\epsilon(p_x,p_y,n,p)\delta_{np}} \right].
\]

(41)

Figure 3 shows the damping term \(D_i\) for the POD modes of M2 and its ratio with the corresponding dissipation \(\epsilon_i\). In these plots \(i = (p_x,p_y,p)\) represents the POD mode \(i\) of the LDM. The modes are ordered by decreasing energy content (see also the second column of Table III). Note also that in Eq. (38) the dissipation term appears only when \(n = p\). In the top panel of the figure, the modes are grouped in accordance with their degeneracy as given in Tab. III the first two modes are \((0,1,1)\) and \((1,0,1)\) (the other two are obtained from complex conjugation), the second pair \((0,1,2)\) and \((1,0,2)\), etc.

The order of magnitude of the eddy viscosity and diffusivity term is determined by the maximum value of the ratio \(D_i/\epsilon_i\). Figure 3 shows that the Heisenberg eddy viscosity-diffusivity introduces an overwhelming damping for the less energetic modes. This will cause stationary convection solutions for both LDMs. These limitations of the model with constant eddy viscosity–diffusivity make it necessary to extend the LDM to a model–dependent or modal eddy viscosity–diffusivity coefficient, \(\eta(i)\). Such a model matches the decreasing ratio \(D_i/\epsilon_i\).

C. Structure of the POD modes

Before we discuss the time evolution of the LDM with constant and modal eddy viscosity–diffusivity, we want to describe the structure of the POD modes. Table III presents the fifteen most energetic modes \(\phi^{(n)}_{m_x,m_y} \) which are represented by the triplet \((n_x,n_y,n)\). Their decreasing energy content is shown in the fourth and fifth column where the individual eigenvalues \(\lambda_{n_x,n_y}^{(n)}\) and their ratio with respect to \(\lambda_{0,1}^{(1)}\) are listed, respectively. In addition, we show their degeneracy and the resulting share in the total energy content.

The POD analysis has been done for the range of horizontal wavenumbers of \(0 \leq n_x \leq 16\), \(-16 \leq n_y \leq 16\). The order of magnitude of the resulting eigenvalue spectrum is in qualitative agreement with those from Smith et al. and...
| \( k \) | \( i \) | Order | Modes in LDM | \( \lambda^{(i)}_{n_x,n_y} \) | \( \lambda^{(i)}_{n_x,n_y} / \lambda^{(1)}_{0,1} \) | Degeneracy | Energy in% |
|---|---|---|---|---|---|---|---|
| 1 | 1, 2 | (0, 1, 1), (1, 0, 1) | 6.1705 | 1.0000 | 4 | 30.879 |
| 2 | 3, 4 | (0, 1, 2), (1, 0, 2) | 1.6710 | 0.2708 | 4 | 8.362 |
| 3 | 5, 6 | (1, 1, 1), (1, -1, 1) | 0.5999 | 0.0972 | 4 | 3.002 |
| 4 | 7, 8 | (0, 2, 1), (2, 0, 1) | 0.3788 | 0.0614 | 4 | 1.896 |
| 5 | 9, 10 | (0, 0, 1), (0, 0, 2) | 0.3722 | 0.0603 | 2 | 0.932 |
| 6 | 11, 12 | (0, 1, 3), (1, 0, 3) | 0.3583 | 0.0581 | 4 | 1.790 |
| 7 | 13, 14 | (1, 2, 1), (2, 1, 1) | 0.2729 | 0.0442 | 8 | 2.690 |
| 8 | 15, 16 | (1, -2, 1), (2, -1, 1) | 0.2315 | 0.0375 | 4 | 1.158 |
| 9 | 17, 18 | (0, 1, 4), (1, 0, 4) | 0.2274 | 0.0369 | 4 | 1.138 |
| 10 | 19, 20 | (0, 2, 2), (2, 0, 2) | 0.2146 | 0.0348 | 4 | 1.078 |
| 11 | 21, 22 | (0, 3, 1), (3, 0, 1) | 0.2078 | 0.0337 | 4 | 1.039 |
| 12 | 23, 24 | (1, 1, 2), (1, -1, 2) | 0.1975 | 0.0320 | 4 | 0.971 |
| 13 | 25, 26 | (1, 1, 3), (1, -1, 3) | 0.1630 | 0.0264 | 4 | 0.813 |
| 14 | 27, 28 | (2, 2, 1), (2, -2, 1) | 0.1554 | 0.0252 | 1 | 0.194 |
| 15 | 29 | (0, 0, 3) | 0.1531 | 0.0248 | 4 | 0.769 |
| 16 | 30, 31 | (1, 1, 4), (1, -1, 4) | 0.1531 | 0.0248 | 4 | 0.769 |

TABLE II: The first fifteen most energetic POD modes as obtained from the snapshot analysis. The corresponding index in the LDM, the eigenvalues, their ratio with respect to the first eigenvalue, the degeneracy of the modes, and their percentage of the total mean energy content (kinetic energy plus scalar variance) are also given. Ordering is with respect to the eigenvalue \( \lambda^{(n)}_{n_x,n_y} \).

FIG. 4: (Color online) The wall-normal dependence of the components of the four most energetic POD modes which is given by \( \phi_{m,n_x,n_y}^{(i)}(z) \). The index \( m = 1, 2, 3, 4 \). The particular mode is indicated in each panel. Linestyles for different components are the same in all four subfigures.

and Moehlis et al.\textsuperscript{26} for plane Couette flow, as well as those from Deane and Sirovich\textsuperscript{22} for Rayleigh-Bénard convection in a Cartesian cell. The first few modes carry the major share in the total energy and are followed by a tail of slowly decaying and energetically much less significant degrees of freedom. For example, in the plane Couette flow case, the first mode carries about 68% of the total energy, whereas in the RB convection case\textsuperscript{22} about 39% at \( Ra \approx 46000 \). As shown in Ref.\textsuperscript{24}, the number of energetically significant modes in RB convection increases rapidly with increasing Rayleigh number. Thus, at \( Ra = 1.05 \times 10^5 \), our primary POD mode only represents about 31% of the total mean energy. The slow decay of the spectrum results in the POD mode with the triplet (0, 5, 1) which is No. \( i = 101 \) in M2 (index \( i \) corresponds with the second column of Tab.\textsuperscript{11} still carrying 0.28% of the total energy.

Figure 4 displays the wall-normal dependence of the components of the eigenvectors of the four most energetic
POD modes, which are $\phi_{m,0}^{(1)}$, $\phi_{m,1}^{(2)}$, $\phi_{m,1}^{(1)}$, and $\phi_{m,0}^{(1)}$, all of them representing a vertical circulation whose axes, due to their four-fold degeneracy, are along $x$ or $y$, or in the case of $\phi_{m,1}^{(1)}$ along the horizontal diagonals of the periodic box, respectively. We found in our analysis that the convergence of the snapshot method is very slow. This causes for example the slight differences between $\phi_{1,1}^{(1)}$ and $\phi_{2,1}^{(1)}$ as seen in Fig. 4. Figure 5 presents the wall-normal functionality of further POD modes. For example, the POD mode belonging to $\phi_{m,0}^{(1)}$ is a purely mechanical mode with degeneracy 2 and vanishing components $m=3,4$. The two remaining horizontal components ($m=1,2$), which are denoted as $\phi_{m,0}^{(1)}$ and $\phi_{m,0}^{(2)}$, satisfy orthogonality (see panel (a)).

Another interesting mode is the sixth most energetic POD mode belonging to $\phi_{m,0}^{(3)}$. It represents a pumping motion in the $x$ or $y$ directions and is the first of the most energetic POD modes with a non-vanishing vertical vorticity component. The mode is shown in panel (b) of Fig. 5. Furthermore, we also display a purely thermal mode $\phi_{m,0}^{(3)}$ with degeneracy 1 in panel (c) and the mode $\phi_{m,0}^{(3)}$ which shows the growing influence of the viscous and thermal boundary layers close to the walls at higher horizontal wave numbers in panel (d) of the figure.

FIG. 5: (Color online) Wall-normal dependence of further POD modes. The legend in panels (c) and (d) is identical to panel (b).

Figure 6 displays the spatial structure of the velocity field corresponding to the functions $\Phi_{m,0}^{(1)}$ and $\Phi_{m,1}^{(1)}$. They are obtained by summing all individual eigenfunctions as given by Eq. (12) over their degeneracy, i.e., modes that have the same energy content. Figure 7 illustrates the isosurfaces of the temperature fluctuations $\theta(x,t)$ that are captured by the fourth component of the same POD modes. We show $\Phi_{m,0}^{(4)}$ in the top panel and $\Phi_{m,0}^{(4)}$ in the bottom panel of the figure.

One important aspect of the POD mode analysis and LDM setup is the question of how well the largest structures and their dynamics can be represented. Here, we display the reconstruction of a DNS snapshot by the mode set. In Fig. 8, we compare the velocity field as reconstructed with the POD modes $\Phi_{m,n_x,n_y}^{(k)}$ from model M2. The top panel in the figure shows the reconstruction and the bottom panel the full DNS snapshot. The same analysis is repeated for the temperature fluctuations at the same instant. Figure 9 shows the temperature field again reconstructed with M2. Although, not all details are reproduced, both figures indicate that the most important structures of temperature and velocity are captured by M2. Our analysis showed that the temperature fluctuations converged slower than the velocity field. It is known that the temperature field forms so called thermal plumes – fragments of the thermal boundary layer that detach from the cooling and heating plates and move into the bulk (see e.g. Refs. 35, 36 and 37). These fine-scale filamented structures carry the heat across the cell.
FIG. 6: (Color online) Vector plots of the first and third most energetic POD modes $\Phi^{(1)}_{m,0,1}$ (top) and $\Phi^{(1)}_{m,1,1}$ (bottom). The velocity field is therefore projected into three $y$-$z$ planes.

FIG. 7: (Color online) Isosurfaces of the first and third most energetic POD modes $\Phi^{(1)}_{4,0,1}$ (top) and $\Phi^{(1)}_{4,1,1}$ (bottom) which describe the temperature fluctuations. The isosurfaces are taken at the level $0.01 \theta_c$. In addition, we plot contours at the sideplanes. Red color stands for $1.7 \theta_c$ and blue for $-1.7 \theta_c$. 
D. Time evolution of LDM with constant eddy viscosity–diffusivity

In the following, we will discuss the evolution of the LDM. This section is for the case with the constant eddy viscosity–diffusivity, i.e. \( \eta = \alpha \eta \) where \( \eta \) is given by (11), and \( \alpha \in \mathbb{R} \).

As was found in Ref. 8 for the long-term integration of their LDM, this closure can lead to a statistical equilibrium state which accumulates too much energy. One can overcome this behavior in parts following a work by Kalb and Deane, who kept the original gradient of the mean temperature field, \( d\langle T \rangle / dz \), in the evaluation of the last two terms of Eq. (25). Consequently, the equations for \( \Im \) and \( \epsilon(T) \) become

\[
\Im(p_x, p_y, n, p) = -\hat{u}_c \int_{-1/2}^{1/2} \frac{d\langle T \rangle}{dz} \phi^{(p)}_{3, \pm p_x, \pm p_y} \phi^{(p)*}_{4, \pm p_x, \pm p_y} dz . \tag{42}
\]

and,

\[
\epsilon(T)(p_x, p_y, p) = \sqrt{L_x L_y} \int_{-1/2}^{1/2} \frac{d^2\langle T \rangle}{dz^2} \phi^{(p)}_{4, 0, 0} dz . \tag{43}
\]

The long-time evolution of the LDM requires the time integration of the ODE system (39). A fourth-order Runge-Kutta scheme is applied. Our studies found that for \( \alpha \lesssim 0.87 \) the ODE systems for M1 and M2 become unstable which is triggered by the most energetic modes that accumulate energy which cannot be transferred sufficiently fast to small scales. For \( \alpha \geq 0.87 \), we still obtain a regime of the LDM carrying too much energy. With increasing prefactor \( \alpha \), this energy surplus at the first POD modes decreases up to a threshold which enforces the whole dynamical system into a stationary state. For M1 this sets in at \( \alpha \geq 1.0 \) and for M2 at \( \alpha \geq 1.04 \). In Fig. 10 we compare therefore the total energy of both LDMs for different values of \( \alpha \) with the original DNS data. After a relaxation phase both models reach a statistically stationary regime. While model M1 always yields energy time series below that of the original DNS, model M2 comes closer to the evolution of the total energy from the original DNS. Note that the fluctuations of the total energy in both cases are significantly larger than for the original DNS data. This is due to the fact that
only the largest-scale modes are kept in the model and coupled with each other. The exchange of energy among them can cause larger variations. The figure clearly indicates that M1 falls short in representing the long-term dynamics of the convective flow. In model M2 however it is possible to obtain a total energy in the range of the original data. It can be concluded that a further reduction of degrees of freedom below that of M1 is thus not possible.

Figure 11 shows the long-time behavior of $a^{(1)}(0,1)$, $a^{(2)}(1,0)$, $a^{(1)}(1,1)$, and $a^{(1)}(0,2)$ for M2 in the stationary overdamped state. This fixed point of the LDM is connected with a stationary pattern of the velocity and temperature fields. It is now clear how a constant eddy viscosity–diffusivity model produces the observed behavior. The second mode (dashed line) is accumulating too much energy ($|a^{(2)}(1,0)| > |a^{(1)}(0,1)|$), since it is not damped strongly enough. Similar behavior of the LDM has been found and analyzed in detail by Aubry et al.\textsuperscript{4}. They detected a similar fixed point for a specific range of their constant eddy viscosity.

Since the overdamped stationary state is not of interest for our study, the values of $\alpha$ have to be chosen smaller than these limits. Figure 12 shows the time evolution of the modal amplitude for the most energetic POD mode, $a^{(1)}(0,1)(t)$, obtained from the LDMs M1 and M2, respectively at a smaller value of $\alpha$. Data are again compared with DNS time series which are obtained by projection of the snapshots on the particular modes. Clearly, at $\alpha = 0.93$ the agreement of the real part is very good for M2, even for long term evolution. The amplitude of the imaginary part in M2 is too large, the one in M1 is comparable with the DNS. While the amplitudes partly agree, the temporal behavior of the modes differs qualitatively. As observed from Fig. 12, the real and imaginary parts of the expansion coefficients vary in a limited range (in parts periodically) once the initial relaxation to a statistically stationary state is finished. This is in contrast to the projection of the DNS snapshots on the modes. It indicates that the additional dissipation has a strong impact on the dynamics of the large-scale degrees of freedom. The figure unravels the shortcoming of the present straightforward and simplest closure: all modes are affected by the same additional dissipation, the ones that have many couplings within the model as well as those with much less mode interactions. The constant eddy viscosity–diffusivity establishes an additional flux from large resolved to small unresolved scales and seems not to allow a back-scatter which is important and known from other (subgrid-scale) closures.\textsuperscript{38}
FIG. 10: Time series of the total energy for LDM M1 (top panel) and M2 (bottom panel). Different magnitudes of the parameter $\alpha$ in the constant Heisenberg eddy viscosity–diffusivity are applied in both models and given in the legend of the bottom panel. Note that $\alpha = 1.04$ is not taken for M1.

FIG. 11: Time series of the real and imaginary parts of the expansion coefficients $a_{n_x,n_y}^{(\alpha)}(t)$ of the four most energetic POD modes for M2 and $\alpha = 1.04$. 
FIG. 12: Time series for the real (top) and imaginary (bottom) parts of \( a_{0,1}^{(1)} \) of the two LDMs. The parameter \( \alpha \) is set to a value of 0.93.

E. Time evolution of the LDM with modal eddy viscosity–diffusivity

FIG. 13: Modal eddy viscosity–diffusivity following the maximum values of \( D_i/\epsilon_i \). The real prefactor is now slightly larger and given by \( \alpha = 1.146 \) for the solid line. For comparison, we add the constant value of \( \bar{\eta} \) from Fig. 3. Model M3 is used for evaluation of M2 only. The inset shows the degeneracy restrictions that have to be included for the first 41 modes.

A refinement of the constant eddy viscosity–diffusivity model, that overcomes the shortcomings from above, is possible when switching to the modal eddy viscosity–diffusivity. Figure 13 shows the maximum values of the ratio \( D_i/\epsilon_i \) that were already presented in the top panel of Fig. 3 anew for the eddy viscosity–diffusivity of model M2. The sudden decrease in the maxima for \( i \approx 140 \) is a consequence of the truncation. We confirmed this after plotting the modes resulting from a slightly larger model –denoted as M3– that contains 430 modes with \( |n_x| + |n_y| \leq 6 \), \( 1 \leq (n) \leq 10 \) (crossed symbols in the figure). The solid line is a fit which is given by

\[
\eta(i) = \alpha \beta (\gamma)^i. \tag{44}
\]

We kept the variable real prefactor \( \alpha \) in order to compare this case with the former constant eddy viscosity–diffusivity.
Clearly the new function $\beta(\gamma)^i$ is accounting now for the constant $\eta$ in the Heisenberg dissipation model. Data fit then well for $\beta = 2.639$, and $\gamma = 0.99372$, in the range for $i < 140$. It turned out to be necessary to add some more damping through the factor $\alpha > 1$ and to check the results for the ensemble average and the transient total energy. In agreement with Ref. 8, it yields a positive definite dissipative damping term $D_i$ for all modes. The inset shows the degeneracy for the first 41 modes.

Figure 14 (top panel) displays the energy spectrum of model M2 after integration of the ODE system following the modal dependent closure model (solid line). It is compared with the original POD spectrum of the DNS–data (dashed line). Due to the high accuracy at the tail, the first 120 modes are shown only. It is obvious that a further improvement in the accuracy of the M2–spectrum can be attained considering the effect of the $G$–symmetry group (see Appendix A) on the time evolution coefficients, as it was done for the full DNS data. The analytical determination of the effect of these symmetries, as was studied by Smith et al. for turbulent plane Couette flow, is beyond the scope of the present study and must be addressed in the future. Differences appear for the two most energetic modes $(0,1,1), (1,0,1)$ which have slightly smaller energy. Overshoots for modes No. 12 with $(n_x, n_y, n) = (0, 1, 3), 25 with (1, -1, 3), 26 with (1,1,3), and 29 with (0,0,3) are found. Undershoots are detected for modes No. 9 with $(0,0,1), 10 with (0,0,2), 17 with (1,0,4)$ through 20 with $(2,0,2)$, and 30 with $(1,1,4)$.

The bottom panel, which is analogous to the bottom panel of Fig. 10, shows the instantaneous total energy for 500 time units. The agreement between DNS and M2 is now significantly better. Energy remains largely fluctuating and yields an ensemble average equivalent to 91.11% of the amplitude of the DNS, 5.58% smaller than the 96.19% corresponding to constant eddy viscosity–diffusivity at $\alpha = 1.04$ (bottom panel of Fig. 10).

In Fig. 15 we show the turbulence statistics as obtained from a long-time run of model M2 with modal eddy viscosity–diffusivity. Plane and time averaged vertical profiles of $\langle u^2 \rangle, \langle v^2 \rangle, \langle w^2 \rangle$, and $\langle \theta^2 \rangle$, computed following equations analog to (33). Time average is taken over the first 500 time units. In each case, all the main features of the original DNS–profiles are reproduced, and the truncation yields a reasonable accuracy. The profiles are reproduced qualitatively well.

Figure 16 shows the time evolution of the modal amplitude for the most energetic POD mode, $a_{0,1}^1(t)$, obtained from M2 using the modal eddy viscosity–diffusivity. Comparison with the DNS time series shows a reasonable agreement of the real and the imaginary parts until about $t = 90$. Furthermore, the whole temporal evolution of the coefficients is now much closer to those of the DNS. In contrast with Fig. 12, now the real and imaginary components vary in ranges as wide as the extent of variation for the projection of the DNS snapshots on the modes. The relaxation into a quasi-periodic time variation as in the case with constant eddy viscosity–diffusivity as now absent.
FIG. 15: Turbulence statistics $\langle u^2 \rangle$, $\langle v^2 \rangle$, $\langle w^2 \rangle$, and $\langle \theta^2 \rangle$ versus $z$ as computed from the time evolution of model M2 and compared with the DNS. Linestyles are the same in the four subfigures. The profiles are obtained by taking plane-time averages.

FIG. 16: Time series for the real and imaginary parts of $a_{(1,0)}$ of the M2–LDM using the modal–$\eta$ model.

Figure 17 shows the total temperature field, $\Theta(x, t)$ (see Eq. (11)), reconstructed with M2, at the two instants of time $t = 100$ and 400. We clearly identify the mushroom–shaped isosurfaces, which belong to the same type as the ones presented in Fig. 9 for the fluctuations, and were previously observed (see e.g. Refs. 36,37) for RB convection in cylindrical containers for $Ra = 10^5$–$10^9$. Velocity snapshots at the four instants $t = 100, 200, 300$ and 400 are shown in Fig. 18. It is observed that the flow is characterized by strong up- and downward flows, which are in line with the enhanced fluctuations of the total energy in Fig. 14.
FIG. 17: (Color online) Isosurfaces of the total temperature field $\Theta$ at the two instants of time: $t = 100$ (top row) and $t = 400$ (bottom row). Isosurfaces at $\Theta(x,t)/\theta_c = 0.7$ (left column with top view) and $-0.7$ (right column with a view from below). Data are obtained from model M2.

FIG. 18: (Color online) Velocity field snapshots at four instants of time: (a) $t = 100$, (b) $t = 200$, (c) $t = 300$, (d) $t = 400$. Data are obtained from model M2.
IV. SUMMARY AND OUTLOOK

We have studied a low-dimensional model of turbulent Rayleigh-Bénard convection in a Cartesian slab. The POD modes which form the basis of our model have been obtained by a so-called snapshot method from a record of 320 statistically independent realizations of a DNS of convective turbulence for the same geometry. Temperature and velocity field fluctuations have to be considered therefore as a common four-component vector field where the mode selection is done with respect to the total energy in the convective flow, i.e. kinetic energy plus thermal variance. The Navier-Stokes-Boussinesq equations are then projected onto the POD modes. The Galerkin projection is truncated at two different levels and results in the low-dimensional models denoted as M1 and M2.

Our results can be summarized as follows. The LDMs have to be stabilized by an additional eddy viscosity–diffusivity $\eta$ that assures that the generated energy can be dissipated since the small-scale degrees of freedom are missing in the model. This observation is in line with existing works on flows in channels or cavities. First, we introduced a constant $\eta$ whose order of magnitude, $\bar{\eta}$, was determined as the global maximum quotient of the damping term, $D_i$, and the diffusivity $\epsilon_i$. The calculation of term $D_i$ is based on the requirement to have a statistically stationary dynamics in the LDM. We have then studied the dynamics of our LDM as a function of the level of truncation and the magnitude of $\eta$. Similar to the works in simple wall-bounded flow or cavities, we observe a convergence into a stationary regime for amplitudes of $\eta$ that are too large. This regime can be considered as a fixed point which is however not in the focus of the present study. We showed also that the effect of $\eta$ on the dynamics of the largest-scale modes is to force them into a state with small fluctuations after the passage of a longer transient.

Alternatively, we introduced a modal eddy viscosity–diffusivity, $\eta(i)$, by fitting the algebraic power law of the form $\beta(\gamma)^i$ to all the local maxima of the quotient $D_i/\epsilon_i$, and considering the restrictions due to the degeneracy of the most energetic modes. For the model M2, the long time integration of the ODE system yields solutions with remarkable accuracy in the value of the ensemble average of the total energy (5.6% below the value for constant $\eta$). Also, the energy spectrum of the POD, is very well reproduced, especially in the tail for modes with higher indexes. The vertical profiles of plane-time averaged fluctuations agree qualitatively with those from DNS. Characteristic coherent structures of convection, such as thermal plumes, are reproduced. We can thus conclude that the second approach with the modal eddy viscosity–diffusivity can model the long-term dynamics of turbulent convection qualitatively well. We wish to stress here again that this was in the focus of the present work, namely how far we can advance with a least set of POD modes.

The question is interesting and important in view to more complex situations, e.g. the problem of mixed convection in indoor ventilation systems. Can the same POD framework (with a mathematical foundation) be carried over to more complex convection flows? A big advantage of the present turbulent convection case in the Cartesian box with periodic side walls is that we have 16 symmetries that significantly enhance the data base. In view of applications in more complex geometries, this indicates that a similar approach might be much more complicated and could enhance the limitations that showed up already for the present turbulent flow. Nevertheless, since the questions, for example with the long-term behavior of the large-scale circulation in turbulent convection or indoor ventilation, are important, we believe that it is still interesting to further follow this route of LDM development based on the POD framework. Some of these efforts will be hopefully presented in the near future.

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Appendix A: Symmetry considerations

The numerical simulation of Rayleigh-Bénard convection for the case of a square section, $L_x = L_y$, generates a maximal amount of symmetry. These discrete symmetries form a group $G$ of eight elements:

$$G = \{I, R, R^2, R^3, F, FR, FR^2, FR^3\}$$

(A1)

whose generators are the rotation by $90^\circ$

$$R(x, y, z, u, v, w, \theta) = (-y, x, z, -v, u, w, \theta)$$

(A2)
and the reflection in $x$

$$\mathbf{F}(x, y, z, u, v, w, \theta) = (-x, y, z, -u, v, w, \theta) \tag{A3}$$

Furthermore, another symmetry group $\{I, Z\}$ acts on the vertical direction, where $Z$ is the reflection in $z$

$$\mathbf{Z}(x, y, z, u, v, w, \theta) = (x, y, -z, u, v, -w, -\theta) \tag{A4}$$

resulting in a symmetry group of sixteen elements

$$\mathcal{G} = \{\mathbf{G}, \mathbf{ZG}\} \tag{A5}$$

when combined. Since each element of the symmetry group generates a possible flow, the ensemble is enlarged by a factor of sixteen, thus increasing the accuracy of any statistical evaluation of the flow.

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