

UDNS or LES, That Is the Question

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Abstract

In the framework of the spectral element method, a comparison is carried out on turbulent firstand second-order statistics generated by large eddy simulation (LES), under-resolved (UDNS) and fully resolved direct numerical simulation (DNS). The LES is based on classical models like the dynamic Smagorinsky approach or the approximate deconvolution method. Two test problems are solved: the lid-driven cubical cavity and the differentially heated cavity. With the DNS data as benchmark solutions, it is shown that the numerical results produced by the UDNS calculation are of the same accuracy, even in some cases of better quality, as the LES computations. The conclusion advocates the use of UDNS and calls for improvement of the available algorithms.

Keywords

Spectral Element Method, Under-Resolved Direct Numerical Simulation, Large Eddy Simulation, Lid-Driven Cavity, Differentially Heated Cavity

1. Introduction

The numerical simulation of turbulent flows still remains a major challenge, especially at high values of the Reynolds number. While direct numerical simulation (DNS) is feasible at the expense of large computational resources for moderate Reynolds numbers of the order $O(10^5)$, developed turbulence in the range $O(10^6 - 10^7)$ is presently still out of reach and needs exascale computers. Over the last decades, large eddy simulation (LES) (cf. e.g. [1]-[3]) has become the efficient tool to tackle those flows, even in industrial applications.

In a LES, the dynamics of the gross structures of the flow is computed by integrating the filtered Navier-Stokes (NS) equations, while the fine structures of the flow that cannot be resolved by the computational grid are modeled. To obtain the LES equations, a low-pass filter built through a convolution operator is applied to the NS equations. In the context of high-order methods like the spectral element method (SEM) [4], the LES implementation favors a complete disconnection between the LES model and the filtering procedure. For example, the dynamic Smagorinsky model [5] [6] may be chosen and the modal [7] or nodal [8] filters represent one of the basic features of the numerical procedure. This approach was successfully carried out by Blackburn and Schmidt [9] and Bosshard *et al.* [10].

A fundamental issue of LES consists in checking the convergence of the model used with respect to some reference benchmark like experimental results or DNS data. Here the two DNS test cases are the lid-driven cavity (LDC) problem [11] [12] and the differentially heated cavity (DHC) [13]. Both problems were solved by a Chebyshev spectral method with discretizations resolving all spatial scales till the Kolmogorov scale. LES computations for the LDC are reported in [14] [15] while for the DHC they are detailed in [10]. The LES computations were compared with the DNS results; they showed excellent agreement for first-order statistics and very good concordance for the second-order statistics. Furthermore, in [10], a grid convergence study is performed showing that when the number of LES grid points increases the LES results get closer and closer to the DNS results.

In this paper we want to examine another viewpoint of comparison between UDNS and LES. Namely, we will examine the question: Can a UDNS with a coarse grid yield comparable results with the LES calculations? Phrased another way, do we need an LES model if we can achieve through a UDNS the same results?

The paper is organized as follows. Section 2 presents the two test cases: the LDC and DHC problems. In Section 3 the filtered equations are given with the various LES models used in this study. The spectral element method is briefly summarized in Section 4 with the space and time discretizations and the associated filters. Section 5 treats the LDC results, while the DHC is the subject of Section 6. Conclusions are drawn in Section 7.

2. Test Cases Description

We will describe the geometrical features of the test problems and the associated mathematical models.

2.1. The lid-Driven Cavity

The first numerical test treats the lid-driven cubical cavity as shown in **Figure 1**. No-slip walls are imposed. The top wall is driven by a regularized velocity profile, tangential to the surface. This profile is expressed by a high-order polynomial that avoids the presence of discontinuities along the lid edges and the upper corners, namely

$$u_1(x, y = h, z) = U\left(\left(1 - \left(\frac{x}{h}\right)^{18}\right) \left(1 - \left(\frac{z}{h}\right)^{18}\right)\right)^2,$$
(1)

where U is the maximum wall velocity. Figure 1 shows the set-up of the lid-driven cavity $[-h,h]^3$ with a side length of 2h and the origin of the coordinate system located at the cavity center. The lid-driven cavity flow involves multiple counter-rotating recirculation flow regions and a rich variety of flow physics. The simulations are run at a Reynolds number of 12,000 that corresponds to a locally-turbulent regime [12] [14].

The mathematical model is given by the Navier-Stokes equations for a viscous Newtonian incompressible fluid

$$\frac{\partial \boldsymbol{u}}{\partial t} + \boldsymbol{u} \cdot \nabla \boldsymbol{u} = -\nabla p + \frac{1}{\operatorname{Re}} \Delta \boldsymbol{u} + \boldsymbol{f}, \ \forall (\boldsymbol{x}, t) \in \Omega \times \boldsymbol{I},$$
(2)

$$\nabla \cdot \boldsymbol{u} = 0, \forall (\boldsymbol{x}, t) \in \Omega \times I, \tag{3}$$

where u is the velocity field, p the reduced pressure (normalized by the constant fluid density), f is the body force per unit mass and Re is the Reynolds number

$$\operatorname{Re} = \frac{UL}{v},\tag{4}$$



expressed in terms of the characteristic length L = 2h, the characteristic velocity U, and the constant kinematic viscosity ν . The symbol Ω denotes the computational domain. The evolution of the system is studied in the time interval $I = [t_0, T]$. The governing Equations (2)-(3) are supplemented with appropriate no-slip boundary conditions for the fluid velocity u. As the problem is time-dependent, a given divergence-free velocity field is required as initial condition in the internal fluid domain.

2.2. The Differentially Heated Cavity

In this case the geometry is a cube where all walls are fixed. The flow is generated by a temperature difference between the hot left side wall and the cold right side wall. All other walls are insulated. Figure 2 displays the problem set-up and the geometry. The cavity is given in non-dimensional coordinates by $[0,1]^3$ with the origin of the coordinate system located in the lower left corner of the cavity.

With the Boussinesq approximation the fluid is considered as incompressible. The mathematical model includes the advection-diffusion temperature equation. Therefore the buoyancy term in the momentum equations incorporates the influence of the temperature field. The non-dimensional Boussinesq equations read

$$\frac{\partial \boldsymbol{u}}{\partial t} + \boldsymbol{u} \cdot \nabla \boldsymbol{u} = -\nabla p + \frac{Pr}{\sqrt{Ra}} \nabla^2 \boldsymbol{u} - PrT \frac{\boldsymbol{g}}{|\boldsymbol{g}|},\tag{5}$$

$$\nabla \cdot \boldsymbol{u} = 0, \tag{6}$$

$$\frac{\partial T}{\partial t} + \boldsymbol{u} \cdot \nabla T = \frac{1}{\sqrt{Ra}} \nabla^2 T, \tag{7}$$

where T is the temperature, the air Prandtl number Pr = 0.71 and the Rayleigh number is defined as

$$Ra = \frac{g\beta\Delta TL^2}{\nu\alpha},\tag{8}$$

with g the gravity, β the fluid thermal expansion coefficient, ΔT the temperature difference between the two lateral walls, L the height of the cavity and α the thermal diffusivity, respectively.

For high Rayleigh numbers of $O(Ra = 10^9)$, the flow shows a main convective rotating core with two recirculating pockets located in the corner regions and hook-like structures. These structures present high curvature time-averaged streamlines located between the recirculation pocket and the primary counterflow outside the vertical boundary layer. A full description of these phenomena is given in [10] [13] [16].

3. The Filtered Boussinesq Equations and LES Models

3.1. Filtered Boussinesq Equations

In this section we present the LES Boussinesq equations to get acquainted with the filtering procedure and the LES modeling. For the LDC case, the LES NS equations are an isothermal subset of the Boussinesq equations as the model does no longer contain the buoyancy term in the momentum equations and the temperature equation.

Filtering the Navier-Stokes and temperature equations, we obtain the filtered Boussinesq equations

$$\frac{\partial \overline{u}_i}{\partial t} + \frac{\partial}{\partial x_j} \left(\overline{u}_i \overline{u}_j \right) = -\frac{\partial \overline{p}}{\partial x_i} + \frac{Pr}{\sqrt{Ra}} \frac{\partial^2 \overline{u}_i}{\partial x_j \partial x_j} - \frac{\partial \overline{\tau}_{ij}}{\partial x_j} - Pr\overline{T} \frac{g_i}{|\mathbf{g}|}, \ i = 1, 2, 3, \tag{9}$$



Figure 2. The differentially heated cavity.

$$\frac{\partial \overline{u_i}}{\partial x_i} = 0, \tag{10}$$

$$\frac{\partial \overline{T}}{\partial t} + \frac{\partial}{\partial x_j} \left(\overline{u}_j \overline{T} \right) = \frac{1}{\sqrt{Ra}} \frac{\partial^2 \overline{T}}{\partial x_j \partial x_j} - \frac{\partial h_j}{\partial x_j},\tag{11}$$

with the subgrid scale (SGS) tensor τ_{ii}

$$\tau_{ij} = \overline{u_i u_j} - \overline{u_i} \overline{u_j}, \tag{12}$$

and the subgrid scale heat flux:

$$h_j = \overline{u_j T} - \overline{u}_j \overline{T}.$$
(13)

The filtered variables denoted by an overbar are computed as a convolution with a filter kernel G. If we filter the variable ϕ we have

$$\overline{\phi}(\mathbf{x},t) \coloneqq G \star \phi \int_{\Omega} \phi(\mathbf{x}',t) G(\mathbf{x}-\mathbf{x}',t) \,\mathrm{d}\mathbf{x}.$$
(14)

The filters will be presented in section 4.2.

3.2. Subgrid-Scale Field and Reynolds Decomposition

By the filtering operation (14), any variable is decomposed into a resolved part ϕ and the subgrid-scale field ϕ'

$$\phi = \overline{\phi} + \phi'. \tag{15}$$

The decomposition of a statistically stationary flow field ϕ in its ensemble average $\langle \phi \rangle$ and a fluctuating part $\phi - \langle \phi \rangle$ is known as the Reynolds decomposition.

In a LES, in general only the filtered flow solution $\overline{\phi}$ is known but not the subgrid-scale field ϕ' . We define the notation ϕ^{\diamond} for the fluctuating part of a filtered field

$$\phi^{\diamond} = \overline{\phi} - \left\langle \overline{\phi} \right\rangle. \tag{16}$$

In the following, we assume that the turbulent flow has reached a statistically steady state, and the Reynolds average is computed as a time average

$$\left\langle \overline{\phi} \right\rangle = \lim_{T \to \infty} \frac{1}{T} \int_0^T \overline{\phi}\left(t\right) \mathrm{d}t. \tag{17}$$

3.3. LES Models

The additional variables τ_{ii} and h_i in Equations (12)-(13) need to be modeled.

3.3.1. Dynamic Smagorinsky Model

The subgrid scale tensor uses the dynamic Smagorinsky model [5] [6]

$$\tau_{ij} - \frac{1}{3}\tau_{kk}\delta_{ij} = -2\nu_{sgs}\overline{S}_{ij}, \quad \overline{S}_{ij} = \frac{1}{2}\left(\frac{\partial\overline{u}_i}{\partial x_j} + \frac{\partial\overline{u}_j}{\partial x_i}\right),\tag{18}$$

where the SGS viscosity is computed by the relation

$$\boldsymbol{\nu}_{sgs} = \boldsymbol{C}_d \overline{\Delta}^2 \left(2 \overline{\boldsymbol{S}}_{ij} \overline{\boldsymbol{S}}_{ij} \right)^{1/2}. \tag{19}$$

Here the quantity $\overline{\Delta}$ is related to the mesh size. The reader is referred to [10] for a detailed description of the filter width in the SEM framework.

The subgrid heat flux is modelled by a subgrid diffusivity

$$h_j = -\kappa_{sgs} \frac{\partial T}{\partial x_j},\tag{20}$$

where the SGS diffusivity is evaluated as

$$\kappa_{sgs} = \frac{V_{sgs}}{Pr_{sgs}},\tag{21}$$

and is based on a Reynolds analogy assumption.

The Smagorinsky constant C_d in Equation (19) and the subgrid Prandtl number Pr_{sgs} in (21) will be calculated with the help of the dynamic procedure that uses a coarser test filter denoted by $\tilde{\Delta}$. Usually the ratio of the length-scale $\tilde{\Delta}$ associated with the test-filter and the grid length-scale $\bar{\Delta}$ is chosen as two (cfr. [10]). The process allows for the adaption of these parameters to the characteristics of the local flow. The assumption behind this approach rests on a scale-similarity hypothesis which considers that the behavior of the smallest resolved scales is similar to the modeled subgrid scales. The application of the test filter to the filtered Boussinesq equations produces twice filtered equations. The Germano identities allow the evaluation of the difference between the residual stress tensor and the residual heat flux resulting from the double-filtered quantities, namely T_{ii} , H, and the filtered subgrid scale variables, *i.e.* $\tilde{\tau}_{ii}$, \tilde{h}

$$L_{ij} := \left(\widetilde{u_i u_j} - \widetilde{u_i u_j}\right) = \left(\underbrace{\widetilde{u_i u_j} - \widetilde{u_i u_j}}_{T_{ij}}\right) - \left(\underbrace{\widetilde{u_i u_j} - u_i u_j}_{\widetilde{\tau_{ij}}}\right)$$
$$c := \left(\widetilde{uT} - \widetilde{uT}\right) = \left(\underbrace{\widetilde{uT} - \widetilde{uT}}_{H}\right) - \left(\underbrace{\widetilde{uT} - uT}_{\widetilde{h}}\right).$$

We note that L_{ij} and c are composed of explicit terms that are readily computed. The subgrid heat flux models for once and twice filtered temperature equation are defined by the relationships

$$h = -\frac{v_{sgs}\left(\overline{\Delta}\right)}{Pr_{sgs}}\nabla\overline{T}, \quad H = -\frac{v_{sgs}\left(\overline{\Delta}\right)}{Pr_{sgs}}\nabla\overline{T}.$$

3.3.2. Approximate Deconvolution Method

The approximate deconvolution method (ADM) defilters the filtered fields. Following the lead of Stolz *et al.* [17] [18], we suppose that the inverse of *G* in Equation (14) exists and can be evaluated through a finite series in \overline{u} . Therefore the velocity field resulting from the ADM is written as

$$\boldsymbol{u}^{*} = \boldsymbol{G}_{a}^{-1} \boldsymbol{\overline{u}}, \tag{22}$$

where G_a^{-1} is an approximation of the exact inverse G^{-1} . An explicit series approximation is due to van Cittert

$$G_a^{-1} = \sum_{n=0}^{N} (I - G)^n, \qquad (23)$$

where *I* is the identity operator.

Typical deconvolutions to third or fifth order are given by

$$\boldsymbol{u}^{*} = 4\left(\overline{\boldsymbol{u}}^{(1)} + \overline{\boldsymbol{u}}^{(3)}\right) - 6\overline{\boldsymbol{u}}^{(2)} - \overline{\boldsymbol{u}}^{(4)},$$
$$\boldsymbol{u}^{*} = 5\left(\overline{\boldsymbol{u}}^{(1)} + \overline{\boldsymbol{u}}^{(5)}\right) - 11\left(\overline{\boldsymbol{u}}^{(2)} + \overline{\boldsymbol{u}}^{(4)}\right) + 14\overline{\boldsymbol{u}}^{(3)} - \overline{\boldsymbol{u}}^{(6)},$$
(24)

with the notation $\overline{u}^{(k)}$ indicating that the velocity field is filtered k times.

The subgrid scale tensor is computed as

$$\tau_{ij} = \overline{u_i^* u_j^*} - \overline{u_i^*} \overline{u_j^*}.$$
(25)

3.3.3. ADM-DMS Model

Because the approximate deconvolution method does not take account of interactions from the computational grid unresolved scales, it needs to be supplemented with a dissipative term. Stolz *et al.* used an empirical relaxation term to stabilize the computation. Another possibility is to combine the ADM model with a subgrid-viscosity model. The subgrid scale tensor is modeled by the relation

$$\tau_{ij} = \overline{u_i^* u_j^*} - \overline{u_i^* u_j^*} - 2\nu_{sgs} \overline{S_{ij}}.$$
(26)

The ADM-DMS was introduced by Bouffanais in his thesis [19] and in Habisreutinger *et al.* [15]. The mixed scale (MS) model computes a weighted geometric average of the models based on the large scales and those based on the energy at the cutoff, denoted by q_c , as proposed by Loc and Sagaut [3]. In [15] the ADM was coupled with the full MS model leading to the equation

$$\tau_{ij} = \overline{u_i^* u_j^*} - \overline{u_i^*} \overline{u_j^*} - 2 \left(C_d \overline{\Delta}^{1+\gamma} \left| \overline{S_{ij}} \right|^{\gamma} \left(\overline{q_c} \right)^{\frac{1-\gamma}{2}} \right) \overline{S_{ij}},$$
(27)

$$\overline{q_c} = \frac{1}{2} \left(\overline{\boldsymbol{u}} - \widetilde{\boldsymbol{u}} \right) \left(\overline{\boldsymbol{u}} - \widetilde{\boldsymbol{u}} \right), \tag{28}$$

$$\gamma = 0.5. \tag{29}$$

The ADM-DMS method with $\gamma = 1$ was applied to the DHC in [20].

4. Spectral Element Method

4.1. Space and Time Discretizations

The spectral element method (SEM) decomposes the computational domain into $E_x \times E_y \times E_z$ macro-elements where E_i denotes the number of elements in the *i*-th direction. The space discretization of the velocity u and temperature T fields is performed using Lagrange-Legendre approximation polynomials defined on a Gauss-Lobatto-Legendre (GLL) grid, built as a tensor product of one-dimensional grid points that are the roots of the relation

$$(1-\xi^2)L'_N(\xi) = 0,$$
 (30)

where $L_N(\xi)$ is the Legendre polynomial of degree N and $-1 \le \xi \le +1$ is the reference or parent element.

In order to avoid spurious pressure modes, the pressure p is staggered and approximated on a Gauss-Legendre (GL) grid based on points that are the roots of the equation

$$L_N(\xi) = 0. \tag{31}$$

From the functional point of view velocity and pressure are in $IP_N - IP_{N-2}$ spaces of polynomial degree N and N-2, respectively.

The discrete equations are designed using the weak formulation of the Galerkin method. The continuous integrals of the weak formulation are approximated by Gauss-Legendre numerical quadratures. With the notations borrowed from the monograph [4], the semi-discrete problem corresponding to the Boussinesq equations (9)-(11) generates a set of non-linear algebraic-differential equations

$$M \frac{\mathrm{d}\overline{u}}{\mathrm{d}t} + \frac{Pr}{\sqrt{Ra}} K \underline{\overline{u}} + C \left(\underline{\overline{u}} \right) \underline{\overline{u}} - D^{\mathrm{T}} \underline{\overline{p}} = M \underline{\overline{f}}, \qquad (32)$$

$$-D\overline{u} = 0, \tag{33}$$

$$\boldsymbol{M} \, \frac{\mathrm{d}\overline{\underline{T}}}{\mathrm{d}t} + \frac{1}{\sqrt{Ra}} \boldsymbol{K}\overline{\underline{T}} + \boldsymbol{C}\left(\overline{\underline{u}}\right)\overline{\underline{T}} = \boldsymbol{M}\,\overline{\underline{g}},\tag{34}$$

where M is the diagonal mass matrix, K the stiffness matrix, D^{T} the discrete gradient, $C(\underline{\overline{u}})$ the nonlinear convection operator and D the discrete divergence (transpose of D^{T}). The source term \underline{f} includes the subgrid tensor and the buoyancy term, while \overline{g} contains the subgrid heat flux.

The time integration scheme rests upon an implicit treatment of the transient Stokes operator and the linear diffusive terms in order to avoid the stringent stability restrictions. This is performed by an Euler backward scheme of order two (Euler2). The non-linear terms are treated explicitly by extrapolation in time (EXT2). The global scheme Euler2/EXT2 has no splitting error and is globally of second order accuracy. A real advantage of SEM comes from the minimal numerical dissipation and dispersion. However the explicit treatment of the advection term imposes a *CFL* condition on the time step

$$CFL = \max \max_{u,v,w} \left(\frac{|u|\Delta t}{\Delta x}, \frac{|v|\Delta t}{\Delta y}, \frac{|w|\Delta t}{\Delta z} \right) \le \text{const.}$$

The spectral element method is implemented in the toolbox Speculoos [21] that is available as an open source software in [22].

4.2. Filtering

Two types of filters may be used in the SEM methodology.

4.2.1. Nodal Filtering

The nodal filter due to Fischer and Mullen [8] acts in physical space. Introducing $h_{N,j}$, $j = 0, \dots, N$ the set of Lagrange-Legendre interpolant polynomials of degree N based on the GLL grid nodes $\xi_{N,k}$, $k = 0, \dots, N$, the rectangular interpolation matrix operator I_N^M of size $(M+1) \times (N+1)$ is such that

$$\left(I_{N}^{M}\right)_{ji} = h_{N,j}\left(\xi_{M,i}\right). \tag{35}$$

Therefore, the matrix operator of order M

$$\Pi_M = I_M^N I_N^M, \tag{36}$$

interpolates on the GLL grid of degree M a function defined on the GLL grid of degree N and transfers the data back to the original grid. This process eliminates the highest modes of the polynomial representation. A one-dimensional representation of the filter is given by the relation

$$\overline{u} = \left[\alpha \Pi_M + (1 - \alpha) I_N^N \right] u, \tag{37}$$

with $0 \le \alpha \le 1$. In all calculations reported in the sequel, $\alpha = 0.05$.

4.2.2. Modal Filter

Here the variable is filtered in the spectral Legendre space that is built on the hierarchical basis (cf. [7])

$$\phi_{0} = \frac{1-\xi}{2}, \qquad \phi_{1} = \frac{1+\xi}{2},$$
$$\phi_{j} = L_{j}(\xi) - L_{j-2}(\xi), \quad 2 \le j \le N.$$

In the spectral space a low-pass filter is easily implemented and allows to prune the high-wave number spurious modes. A fully detailed description is yielded in [10].

For the ADM-DMS model, the modal filter was used on a polynomial space of degree 8. Figure 3 shows the transfer function that was used by Habisreutinger *et al.* [15]. When the filter is applied to a flow field, every mode determined from the polynomial basis is multiplied by the corresponding value of the transfer function. Because the filter transfer function is nowhere zero and due to the finite dimensional polynomial space, it is possible to compute an exact filter inverse. This exact filter inverse and the approximate deconvolutions of third and fifth order are also shown in Figure 3. We note that the approximate deconvolution of fifth order is very close to the exact filter inverse.

5. Under-Resolved DNS of the Lid-Driven Cavity Problem

The ADM-DMS model was first applied to the lid-driven cavity problem as reported in [15]. LES simulations of the same problem with the dynamic Smagorinsky model and a mixed model proposed by Zang *et al.* [23] can be found in reference [14]. Here, we perform an under-resolved DNS (UDNS) of the lid-driven cavity problem. The spectral element resolution is the same as in the LES computations of Bouffanais, but no model is used. A detailed description of this flow problem is given in [12] [14]. **Table 1** presents the values of the numerical parameters for the DNS and UDNS calculations.

For a quantitative comparison between the UDNS and the DNS, we plot first-order statistics in the mid-plane z/h = 0, namely $\langle \overline{u_1} \rangle$ on the vertical centerline x/h = 0 and $\langle \overline{u_2} \rangle$ on the horizontal centerline y/h = 0 in **Figure 4**. These are the same statistics given in the references [14] [15] [19]. The mean fields along the centerlines of the UDNS shown in **Figure 4** virtually match the DNS solution.



Figure 3. Transfer function of filter (red, solid line), approximate deconvolution of fifth order (blue, dashed line), approximate deconvolution of third order (cyan dashed line) and exact filter inverse (green, dash-dotted line).

|--|

	Re	Elements	Polynomial degree	Time step	Ave. time
		$E_x \times E_y \times E_z$	$N_x = N_y = N_z$	Δt	
DNS	12000	1	128	2.5×10 ⁻³	1000
UDNS	12000	8×8×8	8	10 ⁻³	319



Figure 4. In the mid-plane z/h = 0, DNS (red, solid line) and UDNS (blue, dashed line). (a) $\langle \overline{u}_1 \rangle$ on the line x/h = 0, (b) $\langle \overline{u}_2 \rangle$ on the line y/h = 0.

Now we plot second-order statistics in the mid-plane z/h=0 on the vertical centerline x/h=0 and on the horizontal centerline y/h=0 in Figure 5. In the Figures 5 (a)-(f) three components of the Reynolds stress are shown. The UDNS produces correct second order statistics with only minor differences compared to the DNS. Especially in figures (a), (b) and (e), the UDNS overpredicts the values of the peak and in (d) the value of the peak is underpredicted. Note that for the case (e), the statistical values are of small amplitude $O(10^{-4})$.

The comments made about the transfer function in section 4.2.2 raise the question, if it would be a good idea to use the exact filter inverse to model the subgrid-scale tensor (12). Such a model could be readily implemented. But since no information is lost, the filtering can be considered as a change of variables and according to Domaradzki *et al.* [24] [25] is fundamentally equivalent to an under-resolved DNS. In view of the new UDNS results presented here, in our opinion it remains an open question if ADM is beneficial with the modal filter and how exactly an ADM-based model should be implemented in this context.

6. Under-Resolved DNS of the Differentially Heated Cavity

In order to better understand the efficiency and performance of the LES models, we carried out two underresolved direct numerical simulations (UDNS) of the DHC. These simulations used the same computational parameters as the LES simulations published in [10], but with no model for the subgrid-scale variables. The computational parameters of the UDNS and the DNS are shown in Table 2.

We evaluate the number of grid points for the case UDNS1000 in one direction as the product of the number of elements in one direction times the polynomial degree (The number of grid points in a direction within one element is equal the polynomial degree plus one, but values at the element boundaries need to be counted only once). Therefore, the total number of grid points for the case UDNS1000 is 512,000. For the DNS, the total number of grid points is 4,826,809 or a factor 9.4 larger. The two cases, UDNS512 and UDNS1000, correspond to the two LES computations (cf. [10]) denoted previously as LES512 and LES1000 with 512 elements and 1000 elements, respectively. The polynomial degree is 8 in both cases. With respect to the finest spectral element mesh, the UDNS1000 represents only 11% of the DNS grid points. Due to the coarse under-resolution, the UDNS512 simulation becomes unstable as spurious energy is built up in the highest Legendre modes. Surprisingly, the case UDNS1000 is not only stable, but the flow statistics are also rather accurate and even more accurate than the LES simulations presented in [10].

In Figure 6 we show the decomposition of the computational domain in spectral elements. The first element is chosen to cover the thin boundary layers at the walls and the interior elements follow a geometric progression of ratio 1.3. In spanwise direction, the flow is close to be two-dimensional with predominant three-dimensional effects close to the spanwise end walls. The sizes of the interior elements are chosen to be uniform except near



Figure 5. In the mid-plane z/h = 0, DNS (red, solid line) and UDNS (blue, dashed line). (a) $\langle u_1^{\diamond} u_1^{\diamond} \rangle^{1/2}$ on the line y/h = 0, (b) $\langle u_1^{\diamond} u_1^{\diamond} \rangle^{1/2}$ on the line x/h = 0, ... figure continued on page 12.



Figure 6. Spectral element decomposition with $10 \times 10 \times 10$ elements. (a) View in any plane normal to y, (b) view in any plane normal to x or z.

	Ra	Elements	Polynomial degree	Time step	Sampl. freq.	Ave. time
		$E_x \times E_y \times E_z$	$N_x = N_y = N_z$	Δt		
DNS	10 ⁹	1	169	5×10^{-3}	20	470
UDNS512	10 ⁹	$8 \times 8 \times 8$	8	2×10^{-3}	10	unstable
UDNS1000	10 ⁹	10×10×10	8	2×10^{-3}	10	200

Table 2. Numerical parameters of DNS and UDNS simulations for differentially heated cavity.

the walls (cfr. **Figure 6(b)**). A detailed discussion of the issue on how to choose the spectral element discretization was given in [10].

In Figure 7(a) and Figure 7(b), we compare in the mid-plane y = 0.5 the profile of the mean values of the velocity component $\langle u_3 \rangle$ and the temperature $\langle T \rangle$ along the line z = 0.1, for DNS in red, UDNS in blue and LES with the dynamic Smagorinsky model with subgrid Prandtl number in green. These profiles cross the turbulent recirculation pocket.

The comparison for the velocity component $\langle u_1 \rangle$ in **Figure 8** is done in the mid-plane along the line z = 0.01. This is selected because for higher z-values the absolute value of this velocity component becomes very small. These are the same flow statistics that were presented for the LES simulations in [4]. The agreement with DNS data is very good and the maximal pointwise error is 0.0088 along these lines.

In Figure 9 we show the normal stresses $\langle u_1^{\circ}u_1^{\circ} \rangle$, $\langle u_2^{\circ}u_2^{\circ} \rangle$, $\langle u_3^{\circ}u_3^{\circ} \rangle$ and the temperature variance $\langle T^{\circ}T^{\circ} \rangle$ in the mid-plane y = 0.5 along the line z = 0.1. The normal stresses $\langle u_1^{\circ}u_1^{\circ} \rangle$, $\langle u_3^{\circ}u_3^{\circ} \rangle$ and the temperature variance $\langle T^{\circ}T^{\circ} \rangle$ are accurately predicted with pointwise maximal errors of 9.4×10^{-5} , 2.5×10^{-4} and 1.6×10^{-4} , respectively. For the normal stress $\langle u_2^{\circ}u_2^{\circ} \rangle$, the relative difference between DNS and UDNS1000 is higher, but this is only because this normal stress is about one order of magnitude smaller. It is also interesting to note that the accuracy close to the wall is very high, which is a sign that the sudden increase of the spectral element size has no negative impact on the accuracy. Compared to the LES simulations with the dynamic Smagorinsky model and subgrid Prandtl number, we conclude that a UDNS with at least 1000 elements provides the same accuracy.

7. Conclusions

The UDNS was done with a Legendre spectral element code presented in [21], while the DNS results were produced with a Chebyshev spectral method described in [12] [13].

Although it is unclear how the accuracy of two different numerical methods can be compared, there is no doubt that the resolution of UDNS1000 is coarser than the one in the DNS. It is not possible to perform a



Figure 7. Average $\langle u_3 \rangle$ and $\langle T \rangle$ profiles in the mid-plane y = 0.5 along the line z = 0.1.



Figure 8. Average $\langle \overline{u_1} \rangle$ profile in the mid-plane y = 0.5 along the line z = 0.01. DNS in red, UDNS in blue.



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Figure 9. Comparison of DNS (solid, red line) with UDNS (dashed, blue line) in mid-plane y = 0.5 along the line z = 0.1: (a) Normal stress $\langle u_1^{\circ} u_1^{\circ} \rangle$, (b) normal stress $\langle u_2^{\circ} u_2^{\circ} \rangle$, (c) normal stress $\langle u_3^{\circ} u_3^{\circ} \rangle$, (d) temperature variance $\langle T^{\circ} T^{\circ} \rangle$.

simulation with significantly less grid points using the Chebyshev spectral program as the computation blows up after a while due to the under-resolution. In our opinion, the high accuracy of the SEM UDNS is linked to the fact that a higher order method with domain decomposition is used. It seems that the Chebyshev spectral method is more sensitive to the under-resolution than the Legendre spectral element method, especially as the grid points scale like N^{-2} near the walls and as N^{-1} in the center of the grid. Furthermore, the Chebyshev collocation method is a global method. Therefore any under-resolution in a part of the domain is affecting the flow solution everywhere. In the spectral element computation, we use a very weak filter of Fischer and Mullen ($\alpha = 0.05$) [8], introduced in Section 4.2.1. This weak filtering procedure is implemented to stabilize the spectral element computations to get rid of the spurious energy that builds up in the highest wavenumber modes. One possible explanation for the accurate results of the UDNS is that on a coarse grid and for an under-resolved flow, this stabilizing procedure acts as an implicit LES.

The improvement of the available algorithms should involve more stable time schemes in order to avoid the stringent CFL condition of the explicit treatment of the non-linear terms. That could be achieved by explicit schemes with a larger stability region or by resorting to implicit schemes.

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