CaNS 2.0 Theory Manual

This document introduces the numerics in CaNS 2.0, which solves the incompressible NS equations using a projection method on staggered grids. LES capabilities have been fully supported in the extended version.

1. Staggered grid

We show the calculation of convective H and viscous L terms on a staggered grid,

$$\frac{u_{i+1/2,j}^{n+1} - u_{i+1/2,j}^n}{\Delta t} = H_{i+1/2,j} + L_{i+1/2,j}$$
(1)

The following scheme can be obtained using finite difference/volume scheme. We take a 2D problem as an example, the two terms in the *u*-momentum equation,

$$H_{i+1/2,j} = -\frac{\left(uu\right)_{i+1/2,j} - \left(uu\right)_{i-1/2,j}}{\Delta x} - \frac{\left(uv\right)_{i+1/2,j+1/2} - \left(uv\right)_{i+1/2,j-1/2}}{\Delta y} - \frac{p_{i+1,j} - p_{i,j}}{\Delta x}$$
(2)

$$L_{i+1/2,j} = \nu \left(\frac{u_{i+3/2,j} - 2u_{i+1/2,j} + u_{i-1/2,j}}{\Delta x^2} + \frac{u_{i+1/2,j+1} - 2u_{i+1/2,j} + u_{i+1/2,j-1}}{\Delta y^2} \right)$$
(3)

The two terms in the *v*-momentum direction,

$$H_{i,j+1/2} = -\frac{(uv)_{i+1/2,j+1/2} - (uv)_{i-1/2,j+1/2}}{\Delta x} - \frac{(vv)_{i,j+1} - (vv)_{i,j}}{\Delta y} - \frac{p_{i,j+1} - p_{i,j}}{\Delta y}$$
(4)

$$L_{i,j+1/2} = \nu \left(\frac{v_{i+1,j+1/2} - 2v_{i,j+1/2} + v_{i-1,j+1/2}}{\Delta x^2} + \frac{v_{i,j+3/2} - 2v_{i,j+1/2} + v_{i,j-1/2}}{\Delta y^2} \right)$$
 (5)

In the projection method, we also need the divergence of velocity

$$\partial_i u_i = \frac{u_{i+1/2,j} - u_{i-1/2,j}}{\Delta x} + \frac{v_{i,j+1/2} - v_{i,j-1/2}}{\Delta y}$$
 (6)

Figure 1 shows the stencil points for solving u, v and p. The same stencil points are used even if the viscous cross-terms are included. The stencil points for calculating eddy viscosity are considered separatedly.

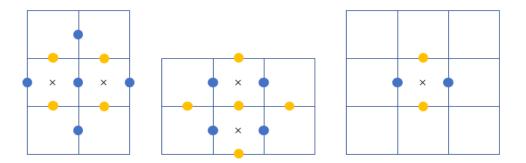


Figure 1 Stencil points for the equations of u (a), v (b) and p (c,d)

2. Poisson equation solver

The elliptic Poisson equation is

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} = R \tag{7}$$

It can be solved via iterative methods, such as Jacobi, Gauss-Seidel (with successive overrelaxation). ADI (AF) can also be used, but it can only obtain an approximate solution of Eq. (7). For 3D problems with two periodic (homogeneous) directions, namely x and y, and one non-homogeneous direction, z, it requires two consecutive FFTs in the homogeneous directions. We denote $\Delta x = L_x/N_x$ and $\Delta y = L_y/N_y$, where N_x and N_y are the numbers of cells in the x and y directions,

$$\frac{\phi_{i-1,j,k} - 2\phi_{i,j,k} + \phi_{i+1,j,k}}{\Delta x^2} + \frac{\phi_{i,j-1,k} - 2\phi_{i,j,k} + \phi_{i,j+1,k}}{\Delta y^2} + \frac{\phi_{i,j+1,k}}{\Delta z^2} + \frac{\phi_{i,j,k-1} - 2\phi_{i,j,k} + \phi_{i,j,k+1}}{\Delta z^2} = R_{i,j,k}$$
(8)

Fourier series in the x direction,

$$\phi_{i,j,k} = \frac{1}{N_x} \sum_{k_x=0}^{N_x-1} \hat{\phi}_{k_x,j,k} e^{\frac{i2\pi k_x i}{N_x}}$$
(9)

Fourier series in the y direction,

$$\hat{\phi}_{k_x,j,k} = \frac{1}{N_y} \sum_{k_y=0}^{N_y-1} \hat{\hat{\phi}}_{k_x,k_y,k} e^{\frac{i2\pi k_y j}{N_y}}$$
(10)

Hence,

$$\phi_{i,j,k} = \frac{1}{N_x N_y} \sum_{k_y=0}^{N_x - 1} \sum_{k_y=0}^{N_y - 1} \hat{\phi}_{k_x,k_y,k} e^{i\left(\frac{2\pi k_x i}{N_x} + \frac{2\pi k_y j}{N_y}\right)}$$
(11)

A modified wavenumber is defined as

$$k_x'^2 = \frac{2\left(1 - \cos\left(\frac{2\pi k_x}{N_x}\right)\right)}{\Delta x^2} = \frac{4}{\Delta x^2} \sin^2\left(\frac{\pi k_x}{N_x}\right) = -\frac{\lambda_x}{\Delta x^2}$$
(12)

$$k_y'^2 = \frac{2\left(1 - \cos\left(\frac{2\pi k_y}{N_y}\right)\right)}{\Delta y^2} = \frac{4}{\Delta y^2} \sin^2\left(\frac{\pi k_y}{N_y}\right) = -\frac{\lambda_y}{\Delta y^2}$$
(13)

The eigenvalue λ_x has no units, but the wavenumber has unit $1/m^2$. Hence,

$$\left(\frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2} + \delta_z^2\right) \hat{\hat{\phi}}_{k_x, k_y, k} = \hat{\hat{R}}_{k_x, k_y, k} \tag{14}$$

Given (k_x, k_y) , Eq. (14) is a tridiagonal system of equations; there are in total $N_x \times N_y$ systems of equations. For square ducts, the same form can be obtained by performing inverse DCT-II in the y direction after DFT in the x direction. The modified wavenumber is defined as

$$k_x'^2 = \frac{2\left(1 - \cos\left(\frac{2\pi k_x}{N_x}\right)\right)}{\Delta x^2} = \frac{4}{\Delta x^2} \sin^2\left(\frac{\pi k_x}{N_x}\right) = -\frac{\lambda_x}{\Delta x^2}$$
(15)

$$k_y'^2 = \frac{2\left(1 - \cos\left(\frac{\pi k_y}{N_y}\right)\right)}{\Delta y^2} = \frac{4}{\Delta y^2} \sin^2\left(\frac{\pi k_y}{2N_y}\right) = -\frac{\lambda_y}{\Delta y^2}$$
(16)

The wavenumber k_x is indexed as 0, 1, ..., N-1. The kind of transform is completely determined by the boundary conditions of ϕ .

3. Boundary condition

The Poisson equation on nonuniform grids is

$$\frac{\phi_{i-1,j,k} - 2\phi_{i,j,k} + \phi_{i+1,j,k}}{\Delta x^2} + \frac{\phi_{i,j-1,k} - 2\phi_{i,j,k} + \phi_{i,j+1,k}}{\Delta y^2} + \frac{1}{\Delta z_{f,k}} \left(\frac{\phi_{i,j,k+1} - \phi_{i,j,k}}{\Delta z_{c,k}} - \frac{\phi_{i,j,k} - \phi_{i,j,k-1}}{\Delta z_{c,k-1}} \right) = R_{i,j,k}$$
(17)

$$R_{i,j,k} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} \tag{18}$$

For boundary cells, the calculation of $R_{i,j,k}$ in Eq. (8) has to consider the boundary conditions in the implicit directions. We assume Dirichlet boundary conditions, i.e., $\phi_{i,j,1/2} = c$ and $\phi_{i,j,N,+1/2} = c$. For k = 1,

$$\frac{\phi_{i-1,j,1} - 2\phi_{i,j,1} + \phi_{i+1,j,1}}{\Delta x^2} + \frac{\phi_{i,j-1,1} - 2\phi_{i,j,1} + \phi_{i,j+1,1}}{\Delta y^2} - \frac{1}{\Delta z_{f,1}} \left(\frac{1}{\Delta z_{c,1}} + \frac{2}{\Delta z_{c,0}}\right) \phi_{i,j,1} + \frac{1}{\Delta z_{f,1}} \frac{1}{\Delta z_{c,1}} \phi_{i,j,2} = R_{i,j,1}$$
(19)

$$R_{i,j,1} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} - \frac{2c}{\Delta z_{f,1} \Delta z_{c,0}}$$
(20)

For $k = N_z$,

$$\frac{\phi_{i-1,j,N_z} - 2\phi_{i,j,N_z} + \phi_{i+1,j,N_z}}{\Delta x^2} + \frac{\phi_{i,j-1,N_z} - 2\phi_{i,j,N_z} + \phi_{i,j+1,N_z}}{\Delta y^2} + \frac{1}{\Delta z_{f,N_z}} \frac{1}{\Delta z_{c,N_z-1}} \phi_{i,j,N_z-1} - \frac{1}{\Delta z_{f,N_z}} \left(\frac{2}{\Delta z_{c,N_z-1}} + \frac{1}{\Delta z_{c,N_z-1}} \phi_{i,j,N_z} - \frac{1}{\Delta z_{c,N_z-1}} + \frac{1}{\Delta z_{c,N_z-1}} \phi_{i,j,N_z} - \frac{1}{\Delta z_{c,N_z-1}} \right) \phi_{i,j,N_z} = R_{i,j,N_z}$$

$$R_{i,j,N_z} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} - \frac{1}{\Delta z_{f,N_z}} \frac{2c}{\Delta z_{c,N_z}}$$
(22)

If homogenous boundary conditions, the added term is zero at the right-hand side. In CaNS, both x and y directions must be homogeneous boundary conditions when the two directions are treated implicitly. Consequently, only the z direction is essentially considered in the calculation of the right-hand side at the boundary. We perform transforms in the x and y directions. For internal cell,

$$\frac{1}{\Delta z_{f,k}} \left(\frac{\hat{\phi}_{i,j,k+1} - \hat{\phi}_{i,j,k}}{\Delta z_{c,k}} - \frac{\hat{\phi}_{i,j,k} - \hat{\phi}_{i,j,k-1}}{\Delta z_{c,k-1}} \right) + \left(\frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2} \right) \hat{\phi}_{i,j,k} = \hat{R}_{i,j,k}$$
(23)

$$a\hat{\hat{\phi}}_{i,i,k-1} + b\hat{\hat{\phi}}_{i,i,k} + c\hat{\hat{\phi}}_{i,i,k+1} = \hat{\hat{R}}_{i,i,k}$$
 (24)

$$a = \frac{1}{\Delta z_{f,k}} \frac{1}{\Delta z_{c,k-1}} \tag{25}$$

$$b = -\left(a+c\right) + \left(\frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2}\right) = -\frac{1}{\Delta z_{f,k}} \left(\frac{1}{\Delta z_{c,k-1}} + \frac{1}{\Delta z_{c,k}}\right) + \frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2}$$
(26)

$$c = \frac{1}{\Delta z_{f,k}} \frac{1}{\Delta z_{c,k}} \tag{27}$$

For k = 1,

$$\left[-\frac{1}{\Delta z_{f,l}} \left(\frac{2}{\Delta z_{c,0}} + \frac{1}{\Delta z_{c,l}} \right) + \frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2} \right] \hat{\phi}_{i,j,l} + \frac{1}{\Delta z_{f,l}} \frac{1}{\Delta z_{c,l}} \hat{\phi}_{i,j,2} = \hat{R}_{i,j,l}$$
(28)

$$a_1 = 0 \tag{29}$$

$$b_{1} = b - a = -\frac{1}{\Delta z_{f,1}} \left(\frac{2}{\Delta z_{c,0}} + \frac{1}{\Delta z_{c,1}} \right) + \frac{\lambda_{x}}{\Delta x^{2}} + \frac{\lambda_{y}}{\Delta y^{2}}$$
(30)

$$c_1 = \frac{1}{\Delta z_{f,1}} \frac{1}{\Delta z_{c,1}} \tag{31}$$

For $k = N_z$,

$$\frac{1}{\Delta z_{f,N_z}} \frac{1}{\Delta z_{c,N_z-1}} \hat{\phi}_{i,j,N_z-1} + \left(\frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2} - \frac{1}{\Delta z_{f,N_z}} \left(\frac{2}{\Delta z_{c,N_z}} + \frac{1}{\Delta z_{c,N_z-1}} \right) \right) \hat{\phi}_{i,j,N_z} = R_{i,j,N_z}$$
(32)

$$a_{1} = \frac{1}{\Delta z_{f,N_{z}}} \frac{1}{\Delta z_{c,N_{z}-1}}$$
 (33)

$$b_{1} = b - c = -\frac{1}{\Delta z_{f,N_{c}}} \left(\frac{1}{\Delta z_{c,N_{c}-1}} + \frac{2}{\Delta z_{c,N_{c}}} \right) + \frac{\lambda_{x}}{\Delta x^{2}} + \frac{\lambda_{y}}{\Delta y^{2}}$$
(34)

$$c_1 = 0 \tag{35}$$

We highlight that all the extra treatments of the boundary conditions are due to the implicit z direction. In the following, we list the calculation of b and $R_{i,j,k}$ for various boundary conditions. If Dirichlet boundary conditions,

$$R_{i,j,1} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} - \frac{2c}{\Delta z_{f,1} \Delta z_{c,0}}$$
(36)

$$R_{i,j,N_z} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} - \frac{2c}{\Delta z_{f,N_z} \Delta z_{c,N_z}}$$
(37)

$$\begin{bmatrix} b-a & c & & \\ a & b & c & \\ & a & b & c \\ & & a & b-c \end{bmatrix} \begin{bmatrix} \hat{\phi}_{i,j,1} \\ \hat{\phi}_{i,j,2} \\ \hat{\phi}_{i,j,3} \\ \hat{\phi}_{i,j,4} \end{bmatrix} = \begin{vmatrix} \hat{R}_{i,j,1} \\ \hat{R}_{i,j,2} \\ \hat{R}_{i,j,3} \\ \hat{R}_{i,j,4} \end{vmatrix}$$
(38)

If Neumann boundary conditions,

$$R_{i,j,1} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} + \frac{c \Delta z_{c,0}}{\Delta z_{f,1} \Delta z_{c,0}}$$
(39)

$$R_{i,j,N_z} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} - \frac{c \Delta z_{c,N_z}}{\Delta z_{f,N_z} \Delta z_{c,N_z}}$$
(40)

$$\begin{bmatrix} b+a & c & & \\ a & b & c & \\ & a & b & c \\ & & a & b+c \end{bmatrix} \begin{bmatrix} \hat{\phi}_{i,j,1} \\ \hat{\phi}_{i,j,2} \\ \hat{\phi}_{i,j,3} \\ \hat{\phi}_{i,j,4} \end{bmatrix} = \begin{bmatrix} \hat{R}_{i,j,1} \\ \hat{R}_{i,j,2} \\ \hat{R}_{i,j,3} \\ \hat{R}_{i,j,4} \end{bmatrix}$$
(41)

If periodic boundary conditions,

$$R_{i,j,1} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} \tag{42}$$

$$R_{i,j,N_z} = \frac{\partial_i \hat{u}_i^k}{\gamma^k \Delta t} \tag{43}$$

$$\begin{bmatrix} b & c & a \\ a & b & c \\ a & b & c \\ c & a & b \end{bmatrix} \begin{bmatrix} \hat{\phi}_{i,j,1} \\ \hat{\phi}_{i,j,2} \\ \hat{\phi}_{i,j,3} \\ \hat{\phi}_{i,j,4} \end{bmatrix} = \begin{bmatrix} \hat{R}_{i,j,1} \\ \hat{R}_{i,j,2} \\ \hat{R}_{i,j,3} \\ \hat{R}_{i,j,4} \end{bmatrix}$$
(44)

Eq. (44) is efficiently solved using Thomas algorithm for periodic tridiagonal systems. Note that patched boundaries are not relevant here, since the tridiagonal system contains all cells along the z direction after transposing pencils to z direction. Hence, we should obtain the same results from using different numbers of threads.

4. Low-storage Runge-Kutta scheme

In CaNS, the low-storage Runge-Kutta scheme is combined with the projection method (Chorin 1968; Harlow & Welch 1965; Kim & Moin 1985)

$$u_{i}^{*} = u_{i}^{k} + \Delta t \left(\gamma_{k+1} \left(H_{i}^{k} + \nu L_{jj} u_{i}^{k} \right) + \rho_{k+1} \left(H_{i}^{k-1} + \nu L_{jj} u_{i}^{k-1} \right) - \alpha_{k+1} \partial_{i} p^{k-1/2} \right)$$

$$(45)$$

$$L_{ij}\phi = \frac{\partial_i u_i^*}{\alpha_{k+1}\Delta t} \tag{46}$$

$$u_i^{k+1} = u_i^* - \alpha_{k+1} \Delta t \partial_i \phi \tag{47}$$

$$p^{k+1/2} = p^{k-1/2} + \phi (48)$$

where (k+1)=1,2,3, $u_i^0=u_i^n$ and $u_i^3=u_i^{n+1}$. The variable u_i^{-1} is not needed due to $\rho_1=0$. The operator H_i denotes nonlinear terms, including convective and modelled stress term,

$$H_{i} = \frac{\partial u_{j} u_{i}}{\partial x_{j}} + \frac{\partial \tau_{ij}}{\partial x_{j}}$$

$$\tag{49}$$

The operator L_{ij} denotes second-order derivative,

$$L_{ij} = \frac{\partial^2}{\partial x_i \partial x_j} \tag{50}$$

The coefficients

$$\gamma_{1} = \frac{8}{15}, \gamma_{2} = \frac{5}{12}, \gamma_{3} = \frac{3}{4}$$

$$\rho_{1} = 0, \rho_{2} = -\frac{17}{60}, \rho_{3} = -\frac{5}{12}$$

$$\alpha_{1} = \frac{8}{15}, \alpha_{2} = \frac{2}{15}, \alpha_{3} = \frac{1}{3}$$
(51)

The stability condition is

$$\Delta t \le \min \left[\frac{1.65}{4 \left((v_T + v) \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} + \frac{1}{\Delta z^2} \right) \right), \left(\frac{u_x}{\Delta x} + \frac{u_y}{\Delta y} + \frac{u_z}{\Delta z} \right) \right]$$
 (52)

Its implementation is

$$\Delta t_{exp} \le \min \left[\frac{1.65}{4 \left(v_T \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} + \frac{1}{\Delta z^2} \right) + v \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} + \frac{1}{\Delta z^2} \right) \right), \left(\frac{u_x}{\Delta x} + \frac{u_y}{\Delta y} + \frac{u_z}{\Delta z} \right) \right]$$
(53)

$$\Delta t_{imp1d} \le \min \left[\frac{1.65}{4 \left(v_T \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} + \frac{1}{\Delta z^2} \right) + v \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} \right) \right), \frac{\sqrt{3}}{\left(\frac{u_x}{\Delta x} + \frac{u_y}{\Delta y} + \frac{u_z}{\Delta z} \right)} \right]$$
(54)

$$\Delta t_{imp} \le \min \left[\frac{1.65}{4 \left(v_T \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} + \frac{1}{\Delta z^2} \right) \right), \left(\frac{u_x}{\Delta x} + \frac{u_y}{\Delta y} + \frac{u_z}{\Delta z} \right)} \right]$$
 (55)

The eddy viscosity part is typically less than the laminar viscous part, so it can be neglected. As the Reynolds number increases, the inviscid part dominates the stability condition. If the viscous term is treated implicitly in all the three directions,

$$u_{i}^{*} = u_{i}^{k} + \Delta t \left(\gamma_{k+1} H_{i}^{k} + \rho_{k+1} H_{i}^{k-1} + \alpha_{k+1} \frac{v}{2} L_{jj} \left(u_{i}^{*} + u_{i}^{k} \right) - \alpha_{k+1} \partial_{i} p^{k-1/2} \right)$$
(56)

Let $\Delta u_i = u_i^* - u_i^k$,

$$\left(1 - \alpha_{k+1} \frac{\nu \Delta t}{2} L_{jj}\right) \Delta u_i = \Delta t \left(\gamma_{k+1} H_i^k + \rho_{k+1} H_i^{k-1} + \alpha_{k+1} \left(\nu L_{jj} u_i^k - \partial_i p^{k-1/2}\right)\right)$$
(57)

If L_{jj} is discretized via second-order central differencing, the left-hand side of Eq. (57) is a heptadiagonal matrix. The modified Helmholtz equation can be solved via approximate factorization (AF) or transform method. If the viscous term is treated implicitly only in the z direction,

$$u_{i}^{*} = u_{i}^{k} + \Delta t \left(\gamma_{k+1} \left(H_{i}^{k} + \nu \left(L_{11} + L_{22} \right) u_{i}^{k} \right) + \rho_{k+1} \left(H_{i}^{k-1} + \nu \left(L_{11} + L_{22} \right) u_{i}^{k-1} \right) + \alpha_{k+1} \frac{\nu}{2} L_{33} \left(u_{i}^{*} + u_{i}^{k} \right) - \alpha_{k+1} \partial_{i} p^{k-1/2} \right)$$

$$(58)$$

Let $\Delta u_i = u_i^* - u_i^k$,

$$\left(1 - \alpha_{k+1} \frac{v\Delta t}{2} L_{33}\right) \Delta u_{i} = \Delta t \left(\gamma_{k+1} \left(H_{i}^{k} + v \left(L_{11} + L_{22}\right) u_{i}^{k}\right) + \rho_{k+1} \left(H_{i}^{k-1} + v \left(L_{11} + L_{22}\right) u_{i}^{k-1}\right) + \alpha_{k+1} \left(v L_{33} u_{i}^{k} - \partial_{i} p^{k-1/2}\right)\right) \tag{59}$$

Hence, the right-hand sides of Eq. (45), (57) and (59) are

$$R_i^{exp} = \Delta t \left(\gamma_{k+1} \left(H_i^k + \nu L_{ij} u_i^k \right) + \rho_{k+1} \left(H_i^{k-1} + \nu L_{ij} u_i^{k-1} \right) - \alpha_{k+1} \partial_i p^{k-1/2} \right)$$
(60)

$$R_{i}^{imp} = \Delta t \left(\gamma_{k+1} H_{i}^{k} + \rho_{k+1} H_{i}^{k-1} + \alpha_{k+1} \left(\nu L_{ii} u_{i}^{k} - \partial_{i} p^{k-1/2} \right) \right)$$
 (61)

$$R_{i}^{imp1d} = \Delta t \Big(\gamma_{k+1} \Big(H_{i}^{k} + \nu (L_{11} + L_{22}) u_{i}^{k} \Big) + \rho_{k+1} \Big(H_{i}^{k-1} + \nu (L_{11} + L_{22}) u_{i}^{k-1} \Big) + \alpha_{k+1} \Big(\nu L_{33} u_{i}^{k} - \partial_{i} p^{k-1/2} \Big) \Big)$$

$$(62)$$

In computer codes, the explicitly treated terms are lumped together, and the implicit terms are lumped together. Eq. (57) can be written as

$$\left(1 - \alpha_{k+1} \frac{\nu \Delta t}{2} L_{jj}\right) u_i^* = \Delta t \left(\gamma_{k+1} H_i^k + \rho_{k+1} H_i^{k-1} + \alpha_{k+1} \left(\nu L_{jj} u_i^k - \partial_i p^{k-1/2}\right)\right) + \left(1 - \alpha_{k+1} \frac{\nu \Delta t}{2} L_{jj}\right) u_i^k \tag{63}$$

In CaNS, it is split into

$$u_i^{**} = u_i^k + \Delta t \left(\gamma_{k+1} H_i^k + \rho_{k+1} H_i^{k-1} + \alpha_{k+1} \left(\nu L_{jj} u_i^k - \partial_i p^{k-1/2} \right) \right)$$
 (64)

$$u_{i}^{*} - \alpha_{k+1} \frac{v\Delta t}{2} L_{jj} \left(u_{i}^{*} \right) = u_{i}^{**} - \alpha_{k+1} \frac{v\Delta t}{2} L_{jj} u_{i}^{k}$$
 (65)

$$p^{k+1/2} = p^{k-1/2} + \phi - \alpha_{k+1} \frac{\nu \Delta t}{2} L_{jj} \phi$$
 (66)

Eq. (59) can be written as

$$\left(1 - \alpha_{k+1} \frac{v\Delta t}{2} L_{33}\right) u_{i}^{*} = \Delta t \left(\gamma_{k+1} \left(H_{i}^{k} + v \left(L_{11} + L_{22}\right) u_{i}^{k}\right) + \rho_{k+1} \left(H_{i}^{k-1} + v \left(L_{11} + L_{22}\right) u_{i}^{k-1}\right) + \alpha_{k+1} \left(v L_{33} u_{i}^{k} - \partial_{i} p^{k-1/2}\right)\right) + \left(1 - \alpha_{k+1} \frac{v\Delta t}{2} L_{33}\right) u_{i}^{k} \tag{67}$$

In CaNS, it is split into

$$u_{i}^{**} = u_{i}^{k} + \Delta t \left(\gamma_{k+1} \left(H_{i}^{k} + \nu \left(L_{11} + L_{22} \right) u_{i}^{k} \right) + \rho_{k+1} \left(H_{i}^{k-1} + \nu \left(L_{11} + L_{22} \right) u_{i}^{k-1} \right) + \alpha_{k+1} \left(\nu L_{33} u_{i}^{k} - \partial_{i} p^{k-1/2} \right) \right)$$

$$(68)$$

$$\left(1 - \alpha_{k+1} \frac{\nu \Delta t}{2} L_{33}\right) u_i^* = u_i^{**} - \alpha_{k+1} \frac{\nu \Delta t}{2} L_{33} u_i^k$$
(69)

$$p^{k+1/2} = p^{k-1/2} + \phi - \alpha_{k+1} \frac{v\Delta t}{2} L_{33} \phi$$
 (70)

The variable u_i^{**} is a better approximation of the final velocity than the sum of the terms in the right-hand-side of Eq. (65). Then, the Poisson equation can be solved to obtain ϕ for the interior cells, and the implementation of the boundary conditions assigns values to the ghost cells. These values are used to correct the velocity field. Note that the correction essentially does not correct the ghost cell wall-parallel velocities, although the wall-normal velocity at a wall can be corrected. If there is a homogeneous pressure boundary condition, the wall-normal velocity at the wall is not corrected either,

When implicit 1D/3D are used, the added term to the right-hand side at the boundary is determined by the boundary condition. The left-hand side terms are due to the viscous term $L_{ij}(u_i^*)$, so the wall-normal velocity gradient (Neumann boundary condition) should be used to compute the extra terms added to the left-hand side, rather than no-slip boundary condition. Hence, the only correct approach for coupling a wall model with an implicit scheme is to modify the boundary condition from no-slip to Neumann boundary condition. Hence, the correct implementation of wall model is to completely change the boundary condition for both the left-and right-hand sides, rather than to just modify the right-hand side.

5. Transform method

Equation 错误!未找到引用源。 with 3D implicit viscous diffusion can be solved via transform method,

$$\left(1 - \beta \left(\frac{\lambda_x}{\Delta x^2} + \frac{\lambda_y}{\Delta y^2} + \delta_z^2\right)\right) \Delta \hat{\hat{u}}_{k_x, k_y, k} = \hat{\hat{R}}_{k_x, k_y, k} \tag{71}$$

The transform needs both staggered and non-staggered forms. For example, u and v use staggered and non-staggered transforms in the y direction, respectively. Table 1 lists the limitations of the transform method, and Table 2 lists the actual limitations of CaNS.

Table 1 Boundary conditions

	Explicit RK	Hybrid RK/CN	Hybrid RK/CN (1D)
Mesh-x	Uniform	Uniform	Uniform
Mesh-y	Uniform	Uniform	Uniform
Mesh-z	Non-uniform	Non-uniform	Non-uniform
cbcvel-x	PP, DD, NN, DN	PP, D0D0, N0N0, D0N0	PP, DD, NN, DN
cbcvel-y	PP, DD, NN, DN	PP, D0D0, N0N0, D0N0	PP, DD, NN, DN
cbcvel-z	PP, DD, NN, DN	PP, DD, NN, DN	PP, DD, NN, DN

Table 2 Boundary conditions in CaNS

	Explicit RK	Hybrid RK/CN	Hybrid RK/CN (1D)
Mesh-x	Uniform	Uniform	Uniform
Mesh-y	Uniform	Uniform	Uniform
Mesh-z	Non-uniform	Non-uniform	Non-uniform
bcvel-x	PP, DD, NN, DN	PP, D0D0	PP, DD, NN, DN
bcvel-y	PP, DD, NN, DN	PP, D0D0	PP, DD, NN, DN
bcvel-z	PP, DD, NN, DN	PP, DD, NN, DN	PP, DD, NN, DN

6. LES

The filtered NS equations obtained using a uniform filter,

$$\frac{\partial \overline{u}_i}{\partial x_i} = 0 \tag{72}$$

$$\frac{\partial \overline{u}_i}{\partial t} + \frac{\partial \overline{u_i u_j}}{\partial x_j} = -\frac{1}{\rho} \frac{\partial \overline{p}}{\partial x_i} + v \frac{\partial^2 \overline{u}_i}{\partial x_j \partial x_j}$$
(73)

It is written as

$$\frac{\partial \overline{u}_{i}}{\partial t} + \overline{u}_{j} \frac{\partial \overline{u}_{i}}{\partial x_{j}} = -\frac{1}{\rho} \frac{\partial \overline{p}}{\partial x_{i}} + \nu \frac{\partial^{2} \overline{u}_{i}}{\partial x_{j} \partial x_{j}} - \frac{\partial \tau_{ij}}{\partial x_{j}}$$

$$(74)$$

$$\tau_{ij} = \overline{u_i u_j} - \overline{u}_i \overline{u}_j \tag{75}$$

We need a subgrid model to compute τ_{ij} , since $\overline{u_iu_j}$ is unknown. In practical simulations, the filtering is implicitly achieved in the finite-volume discretization process. It can be understood as follows. We have a DNS turbulent flow field, and we initialize FVM by average flow quantities over each cell. Then, we can do the first time step (infinitesimal), using accurate subgrid stress and numerical flux. The quantities are cell-average values, equivalent to box-filtered quantities. In practical simulations, we must have spatial/temporal discretization errors and SGS errors. If central scheme is applied, we only have numerical dispersion errors and SGS dissipation errors. The subgrid stress is modelled as

$$\tau_{ij} - \frac{1}{3} \tau_{kk} \delta_{ij} = 2 \nu_t \overline{S}_{ij} \tag{76}$$

The isotropic part is ignored in CaNS. Consequently, the viscous term becomes

$$\frac{\partial \tau_{xx}}{\partial x} + \frac{\partial \tau_{yx}}{\partial y} + \frac{\partial \tau_{zx}}{\partial z} = v \left(\frac{\partial^2 \overline{u}}{\partial x^2} + \frac{\partial^2 \overline{u}}{\partial y^2} + \frac{\partial^2 \overline{u}}{\partial z^2} \right) + \frac{\partial}{\partial x} \left(v_t \left(\frac{\partial \overline{u}}{\partial x} + \frac{\partial \overline{u}}{\partial x} \right) \right) + \frac{\partial}{\partial y} \left(v_t \left(\frac{\partial \overline{u}}{\partial y} + \frac{\partial \overline{v}}{\partial x} \right) \right) + \frac{\partial}{\partial z} \left(v_t \left(\frac{\partial \overline{u}}{\partial z} + \frac{\partial \overline{w}}{\partial x} \right) \right)$$

$$(77)$$

$$\frac{\partial \tau_{xy}}{\partial x} + \frac{\partial \tau_{yy}}{\partial y} + \frac{\partial \tau_{zy}}{\partial z} = v \left(\frac{\partial^2 \overline{v}}{\partial x^2} + \frac{\partial^2 \overline{v}}{\partial y^2} + \frac{\partial^2 \overline{v}}{\partial z^2} \right) + \frac{\partial}{\partial x} \left(v_t \left(\frac{\partial \overline{v}}{\partial x} + \frac{\partial \overline{u}}{\partial y} \right) \right) + \frac{\partial}{\partial y} \left(v_t \left(\frac{\partial \overline{v}}{\partial y} + \frac{\partial \overline{v}}{\partial y} \right) \right) + \frac{\partial}{\partial z} \left(v_t \left(\frac{\partial \overline{v}}{\partial z} + \frac{\partial \overline{w}}{\partial y} \right) \right)$$

$$(78)$$

$$\frac{\partial \tau_{xz}}{\partial x} + \frac{\partial \tau_{yz}}{\partial y} + \frac{\partial \tau_{zz}}{\partial z} = v \left(\frac{\partial^2 \overline{w}}{\partial x^2} + \frac{\partial^2 \overline{w}}{\partial y^2} + \frac{\partial^2 \overline{w}}{\partial z^2} \right) + \frac{\partial}{\partial x} \left(v_t \left(\frac{\partial \overline{w}}{\partial x} + \frac{\partial \overline{u}}{\partial z} \right) \right) + \frac{\partial}{\partial y} \left(v_t \left(\frac{\partial \overline{w}}{\partial y} + \frac{\partial \overline{v}}{\partial z} \right) \right) + \frac{\partial}{\partial z} \left(v_t \left(\frac{\partial \overline{w}}{\partial z} + \frac{\partial \overline{w}}{\partial z} \right) \right)$$

$$(79)$$

The resolved viscous terms can be explicitly/implicitly treated, while the modelled terms are always explicitly treated. Next, we discuss the Smagorinsky models. The static Smagorinsky model reads as

$$v_{t} = \left(C_{s} \Delta D(y)\right)^{2} \overline{S} \tag{80}$$

where $\overline{S} = \sqrt{2\overline{S}_{ij}\overline{S}_{ij}}$, and van Driest damping function,

$$D(y) = 1 - \exp\left(-\frac{y^+}{25}\right) \tag{81}$$

The channel test case demonstrates that the performance of the model is sensitive to grid resolutions, grid anisotropy and model constant. The dynamic Smagorinsky model is (Pope 2020)

$$V_t = c_s \Delta^2 \overline{S} \tag{82}$$

$$c_{s} = C_{s}^{2} = \frac{M_{ij}L_{ij}}{M_{ii}M_{ii}}$$
 (83)

$$L_{ij} = \overline{u}_i \overline{u}_j - \tilde{\overline{u}}_i \tilde{\overline{u}}_j \tag{84}$$

$$M_{ij} = 2\Delta^2 \overline{S} \overline{S}_{ij} - 2(\alpha \Delta)^2 \tilde{S} \tilde{\overline{S}}_{ij}$$
 (85)

where the bar denotes the first filter, and the tilde denotes the second filter. In the following, we first discuss its derivation, then the implementation details. We do a second filtering operation on Eq. (74),

$$\frac{\partial \tilde{u}_{i}}{\partial t} + \tilde{u}_{j} \frac{\partial \tilde{u}_{i}}{\partial x_{j}} = -\frac{1}{\rho} \frac{\partial \tilde{p}}{\partial x_{i}} + v \frac{\partial^{2} \tilde{u}}{\partial x_{j} \partial x_{j}} - \frac{\partial T_{ij}}{\partial x_{j}}$$
(86)

$$T_{ij} = \overline{u_i u_j} - \tilde{u}_i \tilde{u}_j = \tilde{\tau}_{ij} + \overline{u}_i \overline{u}_j - \tilde{u}_i \tilde{u}_j$$
(87)

Consequently,

$$L_{ij} = T_{ij} - \tilde{\tau}_{ij} = \overline{u}_i \overline{u}_j - \tilde{\overline{u}}_i \tilde{\overline{u}}_j$$
 (88)

The derivation of Germano's identity assumes uniform grids. It is used to determine the coefficient of SGS models. In the following, we assume $c_s(x,t)$ is constant within a filter width

and that it is the same for the two filtering operations. In practice, this is satisfied when averaging is applied in the homogeneous (or temporal) directions. The Smagorinsky model is

$$\tau_{ij} - \frac{1}{3}\tau_{kk}\delta_{ij} = -2c_s\overline{\Delta}^2 \overline{S}\overline{S}_{ij} \tag{89}$$

If the two filtering operations are low-pass spectral filters, we have

$$T_{ij} = \overline{u_i u_j} - \tilde{\overline{u}_i} \tilde{\overline{u}_j} = u_i u_j - \tilde{u}_i \tilde{u}_j$$

$$\tag{90}$$

$$T_{ij} - \frac{1}{3} T_{kk} \delta_{ij} = -2c_s \tilde{\Delta}^2 \tilde{S} \tilde{S}_{ij} = -2c_s \tilde{\Delta}^2 \tilde{S} \tilde{S}_{ij}$$

$$\tag{91}$$

Substituting Eq. (89) and (91) into Eq. (88),

$$L_{ij} = -2c_s \tilde{\Delta}^2 \tilde{S} \tilde{S}_{ij} + \frac{1}{3} T_{kk} \delta_{ij} + 2c_s \tilde{\Delta}^2 \tilde{S} \tilde{S}_{ij} - \frac{1}{3} \tilde{\tau}_{kk} \delta_{ij}$$

$$\tag{92}$$

It is further written that

$$L_{ij} - \frac{1}{3} \delta_{ij} \left(T_{kk} - \tilde{\tau}_{kk} \right) = 2c_s \overline{\Delta}^2 \overline{S} \overline{S}_{ij} - 2c_s \overline{\tilde{\Delta}}^2 \overline{\tilde{S}} \overline{\tilde{S}}_{ij}$$
 (93)

Since

$$L_{kk} = T_{kk} - \tilde{\tau}_{kk} \tag{94}$$

Eq. (93) becomes

$$L_{ij} - \frac{1}{3} L_{kk} \delta_{ij} = -2c_s \tilde{\overline{\Delta}}^2 \tilde{\overline{S}} \tilde{\overline{S}}_{ij} + 2c_s \bar{\Delta}^2 \bar{\overline{S}} \bar{\overline{S}}_{ij}$$
 (95)

Define

$$M_{ij} = 2\bar{\Delta}^2 \bar{S} \bar{S}_{ij} - 2\tilde{\bar{\Delta}}^2 \tilde{S} \tilde{\bar{S}}_{ij}$$
 (96)a

and squared error

$$E = \left(L_{ij} - \frac{1}{3}L_{kk}\delta_{ij} - c_s M_{ij}\right)^2 \tag{97}$$

We have

$$\frac{\partial E}{\partial c_s} = -2M_{ij} \left(L_{ij} - \frac{1}{3} L_{kk} \delta_{ij} - c_s M_{ij} \right) = -2 \left(L_{ij} M_{ij} - c_s M_{ij} M_{ij} \right)$$

$$(98)$$

The squared error is minimized when

$$c_s = \frac{L_{ij}M_{ij}}{M_{ij}M_{ij}} \tag{99}$$

Eq. (99) is proposed by Lilly in 1992 (Pope 2020). The dynamic procedure uses the following assumptions:

- 1) Uniform Δx and Δy ;
- 2) Model coefficient is uniform at least within a filter width;
- 3) Model coefficient is the same for the two filtering operations;
- 4) The solution is obtained by minimizing the mean-square error.

The dynamic procedure is a method for computing the model coefficient, rather than obtaining an optimal or theoretical coefficient. This is particularly relevant when considering errors arising from numerical discretization and the static form of the subgrid-scale (SGS) model. In CaNS, the default filter is a 3D filter, because it is consistent with the initial implicit 3D box filter and provides more accurate results than a 2D filter in the channel test case. The choice of filter is discussed by (Balaras 1995; Cabot & Moin 1999). In the following, the bar denoting the first filter is left out for brevity. In CaNS, the implementation of the dynamic model is

$$V_T = c_s S \tag{100}$$

$$c_{s} = \frac{M_{ij}L_{ij}}{M_{ij}M_{ij}} = C_{s}^{2}\Delta^{2}$$
 (101)

$$L_{ij} = u_i u_j - \tilde{u}_i \tilde{u}_j \tag{102}$$

$$M_{ij} = 2\left(SS_{ij} - \alpha^2 \tilde{S} S_{ij}\right) \tag{103}$$

where

$$\tilde{S}_{ij} = \frac{1}{2} \left(\frac{\partial \tilde{u}_i}{\partial x_j} + \frac{\partial \tilde{u}_j}{\partial x_i} \right)$$
 (104)

In CaNS, the grid in the z direction is stretched, so the derivative and filtering operations are not commutable,

$$\tilde{S}_{ij} \neq \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$$
 (105)

though we assume uniform cell size <u>locally</u> within a test filter size. The calculation of \tilde{S}_{ij} requires the wall boundary conditions for \tilde{u}_i . It can be shown that

$$\tilde{u}_{j} = \frac{1}{2} \left(\tilde{u}_{j-1/2} + \tilde{u}_{j+1/2} \right) = \frac{1}{4} \left(u_{j-1} + 2u_{j} + u_{j+1} \right)$$
(106)

where

$$u_{j} = \frac{1}{2} \left(u_{j-1/2} + u_{j+1/2} \right) \tag{107}$$

We compute the two terms in Eq. (102) by averaging velocities from cell faces to cell centers, and then performing the second filtering operation using these cell-centered velocities, as implemented in Bae's code.

Regarding the wall model, the nondimensionalized log law reads as

$$\frac{U}{u_{\tau}} = \frac{1}{\kappa} \ln \left(\frac{yu_{\tau}}{v} \right) + B \tag{108}$$

We use Newton-Raphson iteration to solve for u_{τ} . We construct

$$f\left(u_{\tau}\right) = \frac{U}{u_{\tau}} - \frac{1}{\kappa} \ln\left(\frac{yu_{\tau}}{v}\right) - B \tag{109}$$

So

$$f'(u_{\tau}) = -\frac{1}{u_{\tau}} \left(\frac{U}{u_{\tau}} + \frac{1}{\kappa} \right) \tag{110}$$

The iteration formula is

$$u_{\tau,k+1} = u_{\tau,k} - \frac{f\left(u_{\tau}\right)}{f'\left(u_{\tau}\right)} \tag{111}$$

The wall stress yielded from the wall model is represented by modifying the values of wall-parallel velocities at the ghost cells. No-penetration boundary condition is set on the wall-normal velocity, which ensures zero convective flux on the wall.

7. Verification

Verification is to demonstrate that a wall model has been correctly implemented into CaNS. It is done via comparison to other codes or analytical solutions (if existing). Validation is to demonstrate whether a model is good or not, in terms of representing/modeling the true physics. It is done via comparing to experiments, or analytical solutions (if existing). Here, verification is needed to show if the implementation of wall model is correct.

Table 3 Selected verification cases

	Re _τ	Time advancement
Laminar channel2D		Explicit
Laminar channel2D		Implicit 1D
Laminar channel3D		Explicit
Laminar channel3D		Implicit 1D
Turbulent channel2D	180-10 ¹⁰	Explicit
Turbulent channel2D	180-10 ¹⁰	Implicit 1D
Turbulentchannel3D	180-10 ¹⁰	Explicit
Turbulent channel3D	180-10 ¹⁰	Implicit 1D

Table 4 Selected validation cases

	Re	Time advancement
Square duct		Explicit

Appendix A

The eigenvalues are given by $\lambda_k = -4\sin^2(\theta_k)$. Table 5 and Table 6 list the transforms and inverse transforms for different combinations of boundary conditions. N is the number of input points, 0,1,...,N-1 instead of the number of cells. Zero-value points are always left out. The

current index of input points is consistent with Scipy and wiki. The factors in the inverse transforms are consistent with Costa's paper and Scipy.

Table 5 Transforms on a staggered grid (pressure)

BC	θ_k	Transform	Inverse transform
PP		DFT	IDFT
NN	$\frac{\pi k}{2N}$	DCT-II	$\frac{1}{2N}$ DCT-III
DD	$\frac{\pi(k+1)}{2N}$	DST-II	$\frac{1}{2N}$ DST-III
ND	$\frac{\pi(2k+1)}{4N}$	DCT-IV	$\frac{1}{2N}$ DCT-IV
DN		DST-IV	$\frac{1}{2N}$ DST-IV

Table 6 Transforms on a non-staggered grid (velocity)

BC	$ heta_q$	Transform	Inverse transform
PP		DFT	IDFT
NN	$\frac{\pi k}{2(N-1)}$	DCT-I	$\frac{1}{2(N-1)}DCT-I$
DD	$\frac{\pi(k+1)}{2(N+1)}$	DST-I	$\frac{1}{2(N+1)} DST-I$
ND	$\frac{\pi(2k+1)}{4N}$	DCT-III	$\frac{1}{2N}$ DCT-II
DN		DST-III	$\frac{1}{2N}$ DST-II

We list the transforms and inverse transforms for the above 8 types. The formulas below have been checked using Python.

1) staggered NN

$$X_{k} = 2\sum_{n=0}^{N-1} x_{n} \cos\left[\frac{\pi}{N}k\left(n + \frac{1}{2}\right)\right], k = 0, ..., N-1$$
(112)

$$x_{n} = \frac{1}{2N} \left\{ X_{0} + 2 \sum_{k=1}^{N-1} X_{k} \cos \left[\frac{\pi}{N} \left(n + \frac{1}{2} \right) k \right] \right\}, n = 0, ..., N - 1$$
 (113)

2) staggered DD

$$X_{k} = 2\sum_{n=0}^{N-1} x_{n} \sin \left[\frac{\pi}{N} (k+1) \left(n + \frac{1}{2} \right) \right], k = 0, ..., N-1$$
 (114)

$$x_{n} = \frac{1}{2N} \left\{ (-1)^{n} X_{N-1} + 2 \sum_{k=0}^{N-2} X_{k} \sin \left[\frac{\pi}{N} \left(n + \frac{1}{2} \right) (k+1) \right] \right\}, n = 0, ..., N-1$$
 (115)

3) staggered ND

$$X_{k} = 2\sum_{n=0}^{N-1} x_{n} \cos\left[\frac{\pi}{N}\left(k + \frac{1}{2}\right)\left(n + \frac{1}{2}\right)\right], k = 0, ..., N - 1$$
(116)

$$x_{n} = \frac{1}{2N} \left\{ 2 \sum_{k=0}^{N-1} X_{k} \cos \left[\frac{\pi}{N} \left(n + \frac{1}{2} \right) \left(k + \frac{1}{2} \right) \right] \right\}, n = 0, ..., N - 1$$
 (117)

4) staggered DN

$$X_{k} = 2\sum_{n=0}^{N-1} x_{n} \sin \left[\frac{\pi}{N} \left(k + \frac{1}{2} \right) \left(n + \frac{1}{2} \right) \right], k = 0, ..., N - 1$$
 (118)

$$x_{n} = \frac{1}{2N} \left\{ 2 \sum_{k=0}^{N-1} X_{k} \sin \left[\frac{\pi}{N} \left(n + \frac{1}{2} \right) \left(k + \frac{1}{2} \right) \right] \right\}, n = 0, ..., N - 1$$
 (119)

5) Non-staggered NN

$$X_{k} = x_{0} + (-1)^{k} x_{N-1} + 2 \sum_{n=1}^{N-2} x_{n} \cos\left(\frac{\pi}{N-1} kn\right), k=0,...,N-1$$
 (120)

$$x_{n} = \frac{1}{2(N-1)} \left\{ \left(X_{0} + \left(-1\right)^{n} X_{N-1} \right) + 2 \sum_{k=1}^{N-2} X_{k} \cos \left(\frac{\pi}{N-1} nk \right) \right\}, n = 0, ..., N-1$$
 (121)

6) Non-staggered DD

$$X_{k} = 2\sum_{n=0}^{N-1} x_{n} \sin \left[\frac{\pi}{N+1} (k+1)(n+1) \right], k = 0, ..., N-1$$
 (122)

$$x_{n} = \frac{1}{2(N+1)} \left\{ 2 \sum_{k=0}^{N-1} X_{k} \sin \left[\frac{\pi}{N+1} (n+1)(k+1) \right] \right\}, n = 0, \dots, N-1$$
 (123)

7) Non-staggered ND

$$X_{k} = x_{0} + 2\sum_{n=1}^{N-1} x_{n} \cos\left[\frac{\pi}{N}\left(k + \frac{1}{2}\right)n\right], k = 0, ..., N - 1$$
(124)

$$x_{n} = \frac{1}{2N} \left\{ 2 \sum_{k=0}^{N-1} X_{k} \cos \left[\frac{\pi}{N} n \left(k + \frac{1}{2} \right) \right] \right\}, n = 0, ..., N - 1$$
 (125)

8) Non-staggered DN

$$X_{k} = (-1)^{k} x_{N-1} + 2 \sum_{n=0}^{N-2} x_{n} \sin \left[\frac{\pi}{N} \left(k + \frac{1}{2} \right) (n+1) \right], k = 0, \dots, N-1$$
 (126)

$$x_n = \frac{1}{2N} \left\{ 2 \sum_{k=0}^{N-1} X_k \sin \left[\frac{\pi}{N} (n+1) \left(k + \frac{1}{2} \right) \right] \right\}, n = 0, ..., N - 1$$
 (127)

Appendix B

In LES, eddy viscosity is introduced. The hybrid RK-CN scheme Eq. (56) becomes

$$\hat{u}_{i}^{k} = u_{i}^{k-1} + \Delta t \left(\gamma^{k} H_{i}^{k-1} + \rho^{k} H_{i}^{k-2} + \frac{\alpha^{k}}{2 \operatorname{Re}} L \left(\hat{u}_{i}^{k} + u_{i}^{k-1} \right) - \alpha^{k} \partial_{i} p^{k-1} \right)$$
(128)

where operator L denotes the viscous term,

$$L(u_i) = \frac{\partial}{\partial x_j} \left(2(v + v_T) S_{ij} \right) = \frac{\partial}{\partial x_j} \left((v + v_T) \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right) \right)$$
(129)

Re-express it using $\Delta u_i^k = \hat{u}_i^k - u_i^{k-1}$,

$$\Delta u_i^k - \frac{\Delta t \alpha^k}{2 \operatorname{Re}} L(\Delta u_i^k) = \Delta t \left(\gamma^k H_i^{k-1} + \rho^k H_i^{k-2} + \frac{\alpha^k}{\operatorname{Re}} L(u_i^{k-1}) - \alpha^k \partial_i p^{k-1} \right)$$
(130)

Let
$$\beta^k = \frac{\Delta t \alpha^k}{2 \operatorname{Re}}$$
,

$$(1 - \beta^k L)(\Delta u_i^k) = \Delta t (\gamma^k H_i^{k-1} + \rho^k H_i^{k-2} - \alpha^k \partial_i p^{k-1}) + 2\beta^k L(u_i^{k-1})$$
(131)

Expand it,

$$(1 - \beta^k (L_1 + L_2 + L_3)) \Delta u_i^k = \Delta t (\gamma^k H_i^{k-1} + \rho^k H_i^{k-2} - \alpha^k \partial_i p^{k-1}) + 2\beta^k L(u_i^{k-1})$$
 (132)

$$L_{1}\left(\Delta u_{i}^{k}\right) = \frac{\partial}{\partial x_{1}}\left(\left(\nu + \nu_{T}\right)\left(\frac{\partial \Delta u_{i}^{k}}{\partial x_{1}} + \frac{\partial \Delta u_{1}^{k}}{\partial x_{i}}\right)\right) \tag{133}$$

$$L_{2}\left(\Delta u_{i}^{k}\right) = \frac{\partial}{\partial x_{2}} \left(\left(\nu + \nu_{T}\right) \left(\frac{\partial \Delta u_{i}^{k}}{\partial x_{2}} + \frac{\partial \Delta u_{2}^{k}}{\partial x_{i}}\right)\right) \tag{134}$$

$$L_{3}\left(\Delta u_{i}^{k}\right) = \frac{\partial}{\partial x_{3}}\left(\left(\nu + \nu_{T}\right)\left(\frac{\partial \Delta u_{i}^{k}}{\partial x_{3}} + \frac{\partial \Delta u_{3}^{k}}{\partial x_{i}}\right)\right) \tag{135}$$

Eq. (131) is no longer a modified Helmholtz equation, so transform method cannot be applied. If we simplify the viscous term as done by Bae's code, Eq. (132) becomes

$$(1 - \beta^k (L_{11} + L_{22} + L_{33})) \Delta u_i^k = \Delta t (\gamma^k H_i^{k-1} + \rho^k H_i^{k-2} - \alpha^k \partial_i p^{k-1}) + 2\beta^k L_{ij} (u_i^{k-1})$$
 (136)

$$L_{ij}\left(\Delta u_i^k\right) = \frac{\partial}{\partial x_i} \left(\left(v + v_T\right) \frac{\partial \Delta u_i^k}{\partial x_i}\right) \tag{137}$$

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