SUNSET code: Scalable Unstructured Node-SET code for high fidelity fluid simulations in complex geometries

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This document contains a brief description and guide to the SUNSET code. Numerical methods in this code follow [1] (also available at arXiv:2102.02019).

I. OVERVIEW

The Scalable Unstructured Node-SET code (SUNSET code) is designed for high fidelity simulations of fluid flows on non-trivial geometries. The spatial discretisation is by a combination of LABFM [1, 2] and finite differences. Order of accuracy of spatial discretisation can be specified (between 4^{th} and 12^{th}). Time integration via explicit 3^{rd} order Runge-Kutta scheme. Boundary conditions with the characteristic based formulation, with non-periodic boundaries dropping to 4^{th} order at walls, inflows and outflows. Resolution is non-uniform in first two dimensions. Scalable via OpenMP and MPI. Thermodynamics can vary from isothermal to thermal, with temperature dependent transport properties. Solves for an arbitrary number of miscible chemical species.

II. GOVERNING EQUATIONS

The code solves the compressible Navier-Stokes equations in the form

$$\frac{\partial \ln \rho}{\partial t} + \boldsymbol{u} \cdot \nabla \ln \rho = -\nabla \cdot \boldsymbol{u} \tag{1}$$

$$\frac{\partial \boldsymbol{u}}{\partial t} + \boldsymbol{u} \cdot \nabla \boldsymbol{u} = -\frac{1}{\rho} \nabla p + \frac{1}{\rho} \nabla \cdot \boldsymbol{\tau} + \boldsymbol{f}$$
(2)

$$\frac{\partial \rho E}{\partial t} + \nabla \cdot (\boldsymbol{u} \rho E) = -\nabla \cdot (p\boldsymbol{u}) - \nabla \cdot \boldsymbol{q} + \nabla \cdot (\boldsymbol{\tau} \boldsymbol{u}) + \boldsymbol{u} \cdot \boldsymbol{f}$$
(3)

and

$$\frac{\partial Y_{\alpha}}{\partial t} + \boldsymbol{u} \cdot \nabla Y_{\alpha} = \frac{\dot{\omega}_{\alpha}}{\rho} - \frac{1}{\rho} \nabla \cdot (\rho \boldsymbol{V}_{\alpha} Y_{\alpha}) \qquad \forall \alpha \in [1, N_S]$$
(4)

where ρ is the density, \boldsymbol{u} the velocity, p the pressure, E the total energy, $\boldsymbol{\tau}$ the viscous stress, \boldsymbol{f} a body force, \boldsymbol{q} the heat flux, Y_{α} the concentration of species $\alpha \in [1, N_S]$, and \boldsymbol{V}_{α} is the molecular diffusion velocity of species α . $\dot{\omega}_{\alpha}$ is the production rate of species α .

The viscous stress is defined as

$$\tau = \mu \left(\nabla \boldsymbol{u} + \nabla \boldsymbol{u}^T - \frac{2}{3} \boldsymbol{I} \nabla \cdot \boldsymbol{u} \right), \tag{5}$$

where μ is the dynamic viscosity and I is the identity tensor. The energy is

$$E = \sum_{\alpha} h_{\alpha} Y_{\alpha} - \frac{p}{\rho} + \frac{1}{2} \boldsymbol{u} \cdot \boldsymbol{u}$$
 (6)

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where the sum over α is over all $\alpha \in [1, N_S]$, and h_{α} is the enthalpy of species α , given by

$$h_{\alpha} = \int_{T_{ref}}^{T} c_{p,\alpha} dT + h_{\alpha,ref} \tag{7}$$

with $h_{\alpha,ref}$ the enthalpy of species α at the reference temperature T_{ref} , and $c_{p,\alpha}$ the specific heat capacity at constant pressure of species α , which in general may be a function of temperature T. The molecular diffusion term in (4) is modelled as a Fickian diffusion process with

$$-\rho V_{\alpha} Y_{\alpha} = \rho D_{\alpha} \nabla Y_{\alpha} - Y_{\alpha} \sum_{\beta \in [1, N_S]} \rho D_{\beta} \nabla Y_{\beta}, \tag{8}$$

in which D_{α} is the molecular diffusivity of species α , and the final term is a correction to ensure that

$$\sum_{\alpha} Y_{\alpha} = 1. \tag{9}$$

The heat flux vector \mathbf{q} is then given by

$$\mathbf{q} = -\lambda \nabla T + \sum_{\alpha} \rho \mathbf{V}_{\alpha} Y_{\alpha} h_{\alpha}, \tag{10}$$

where λ is the thermal conductivity of the mixture. The system is closed with an ideal gas equation of state

$$p = \rho R_{mix} T = \rho R_0 T \sum_{\alpha} \frac{Y_{\alpha}}{W_{\alpha}}$$
 (11)

where R_{mix} is the gas constant for the mixture, R_0 is the universal gas constant, and W_{α} is the molar mass of species α

A. Evaluation of temperature and heat capacity

The temperature dependence of the specific heat capacity of species α is given by a polynomial of order J

$$c_{p,\alpha} = \sum_{j=1}^{J} a_{\alpha,j} T^{j-1},$$
 (12)

which is valid over a specified temperature range $T \in [T_{low}, T_{high}]$, and in which the coefficients $a_{\alpha,j}$ are taken from CHEMKIN tables. Integration of this yields an expression for the enthalpy

$$h_{\alpha} = \sum_{j=1}^{J} \frac{a_{\alpha,j}}{j} T^{j} + a_{\alpha,J+1},$$
 (13)

whilst the temperature derivative of $c_{p,\alpha}$ is

$$\frac{dc_{p,\alpha}}{dT} = \sum_{j=1}^{J} (j-1) a_{\alpha,j} T^{j-2}.$$
 (14)

Noting that the specific heat capacity of species α is R_0/W_α , where R_0 is the universal gas constant and W_α is the molar mass of species α , the energy can be written as

$$E = \sum_{\alpha} Y_{\alpha} \left(h_{\alpha} - R_0 \frac{T}{W_{\alpha}} \right) + \frac{1}{2} \boldsymbol{u} \cdot \boldsymbol{u}. \tag{15}$$

Substituting the polynomial for h_{α} and collecting terms in powers of T, we can write

$$f(T) = C_0 + \sum_{j=1}^{J} C_j T^j = 0, \tag{16}$$

where the coefficients C are given by

$$C_0 = \frac{1}{2} \boldsymbol{u} \cdot \boldsymbol{u} - E + \sum_{\alpha} Y_{\alpha} a_{\alpha, J+1}$$
(17a)

$$C_1 = \sum_{\alpha} Y_{\alpha} a_{\alpha,1} - R_{mix} \tag{17b}$$

$$C_{j} = \sum_{\alpha}^{\alpha} Y_{\alpha} \frac{a_{\alpha,j}}{j} \qquad \forall j \in [2, J].$$

$$(17c)$$

The derivative of f is

$$f'(T) = \sum_{j=1}^{J} jC_j T^{j-1}.$$
 (18)

Having calculated the coefficients C, we solve the nonlinear equation (16) numerically via a Newton-Raphson method to obtain the temperature T.

Note that in the case where $a_{\alpha,1} \neq 0$ and $a_{\alpha,j \in [2,J]} = 0$ (i.e. $c_{p,\alpha}$ is independent of temperature), f(T) is linear in T, and the Newton-Raphson method will converge (exactly) after a single iteration.

B. Temperature dependent transport properties

The temperature dependence of the transport properties is modelled by the relationship

$$\mu = \mu_{ref} \left(\frac{T}{T_{ref}}\right)^{r_T},\tag{19}$$

where μ_{ref} is the viscosity at the reference temperature T_{ref} , and $r_T = 0.7$ is a constant exponent. The mixture thermal conductivity is given by

$$\lambda = \frac{\mu c_p}{P_r},\tag{20}$$

in which c_p is the heat capacity of the mixture, and Pr is the Prandtl number, taken as constant. The molecular diffusivity of species α is given by

$$D_{\alpha} = \frac{\lambda}{\rho c_p L e_{\alpha}},\tag{21}$$

where Le_{α} is the Lewis number for species α , which is assumed constant for each species.

C. Evaluation of reaction rates

The reaction mechanism is given by M equations written in the form

$$\sum_{\alpha} \nu'_{\alpha,m} \mathcal{M}_{\alpha} \to \sum_{\alpha} \nu''_{\alpha,m} \mathcal{M}_{\alpha} \qquad \forall m \in [1, M],$$
(22)

where \mathcal{M}_{α} symbolises the chemical symbol of species α , and $\nu'_{\alpha,m}$ and $\nu''_{\alpha,m}$ are the molar stoichiometric coefficients of species α in equation m.

The reaction rate of species α is given by

$$\omega_{\alpha} = W_{\alpha} \sum_{m=1}^{M} \hat{\omega}_{\alpha,m} \tag{23}$$

where $\hat{\omega}_{\alpha,m}$ is the molar production rate of species α due to the m-th reaction, and M is the total number of reactions. The molar production rates are given by

$$\hat{\omega}_{\alpha,m} = \left(\nu_{\alpha,m}^{"} - \nu_{\alpha,m}^{'}\right) \left[k_{f,m}\left(T\right) \prod_{\beta=1}^{N_S} \left(\frac{\rho Y_{\beta}}{W_{\beta}}\right)^{\nu_{\beta,m}^{'}} - k_{b,m}\left(T\right) \prod_{\beta=1}^{N_S} \left(\frac{\rho Y_{\beta}}{W_{\beta}}\right)^{\nu_{\beta,m}^{"}}\right],\tag{24}$$

in which $k_{f,m}$ and $k_{b,m}$ are the forward and backward (respectively) rate constants for reaction m. The forward rate constant is given by the Arrhenius expression

$$k_{f,m}(T) = A_m T^{n_m} \exp\left(\frac{E_{a,m}}{R_0 T}\right), \tag{25}$$

with A_m , n_m and $E_{a,m}$ being specified constants representing the pre-exponential factor, the temperature dependence of the pre-exponential factor, and the activation energy.

1. Backwards rate constant

The backwards rate constant is related to the forward rate constant via

$$\ln k_{b,m} = \ln k_{f,m} + \sum_{\alpha} \left(\nu_{\alpha,m}^{"} - \nu_{\alpha,m}^{"} \right) \left(\frac{\hat{g}_{\alpha}}{R_0 T} + \ln \left(\frac{P_0}{R_0} \right) - \ln T \right), \tag{26}$$

in which P_0 is a reference pressure, and \hat{g}_{α} is the Gibbs function for species α , which is given by

$$\frac{\hat{g}_{\alpha}}{W_{\alpha}T} = \frac{a_{\alpha,J+1}}{T} - a_{\alpha,1} \ln T + (a_{\alpha,1} - a_{\alpha,J+2}) - \sum_{j=2}^{J} \frac{a_{\alpha,j}}{j(j-1)} T^{j-1}.$$
 (27)

III. NUMERICAL IMPLEMENTATION

A. Basic framework and principles

The domain is discretised with an unstructured node-set. Each node (or collocation point) i has position r_i , discretisation length-scale s_i (the average local inter-node spacing), and holds the primary variables $\ln \rho_i$, u_i , ρE_i . The governing equations are solved on the set of nodes. Spatial derivatives are evaluated using LABFM [1] in the first two dimensions (x, y) and high-order finite differences (FD) in the third dimension (z). Technically, the node-set is semi-unstructured...

Time integration is by third-order Runge-Kutta scheme. The solution is de-aliased with a high-order LABFM+FD filter after each time step [1]. The accuracy of the spatial discretisation can be changed through compiler flags, between 4^{th} and 12^{th} order accurate in space. In typical usage, 8^{th} order provides a good compromise between accuracy and cost.

At wall, inflow and outflow boundaries, the node-set is structured, with a band of 5 rows of uniformly spaced nodes. Wall boundaries may be curved, but at present, must have uniform resolution along their length. On boundaries (and within the first 2 rows of uniform nodes) the derivatives and filters in the x-y plane are calculated through combinations of FD (normal to boundaries), one-dimensional LABFM (tangential to boundaries), and 2D LABFM. Details in [1].

Aside from the complexities of indexing and derivative evaluation associated with the unstructured node-set approach, the framework is similar to a high-order (central) finite difference scheme.

B. Evaluation of derivatives

Derivatives of the primary variables $\ln \rho$, \boldsymbol{u} , ρE and Y_{α} are evaluated with the LABFM method [1]. Temperature gradients ∇T are also evaluated using LABFM, as are Laplacians where required (i.e. $\nabla^2 \boldsymbol{u}$, $\nabla^2 T$, $\nabla^2 Y_{\alpha}$).

To avoid the explicit evaluation of cross-second derivatives of velocity (which appear in the terms $\nabla \cdot \boldsymbol{\tau}$ and $\nabla \cdot (\boldsymbol{\tau} \boldsymbol{u})$), these terms are constructed by evaluating the gradient of the velocity divergence $\nabla \cdot \boldsymbol{u}$.

The evaluation of derivatives using LABFM constitutes the bulk of the computational cost, and it is hence desirable avoid where possible. Gradients of secondary properties may be evaluated from analytic expressions as follows. The pressure gradient is calculated from the temperature gradient as

$$\nabla p = p \left(\frac{R_0}{R_{mix}} \sum_{\alpha} \frac{\nabla Y_{\alpha}}{W_{\alpha}} + \frac{1}{T} \nabla T + \nabla \ln \rho \right). \tag{28}$$

Gradients of mixture specific heat capacity and enthalpy are also evaluated through analytic expressions in terms of temperature gradients TBC. Similarly, the gradients of the transport properties, which have analytic relations with the thermodynamic quantities, are evaluated as

$$\nabla \mu = \frac{r_T \mu}{T} \nabla T \tag{29}$$

$$\nabla \lambda = \frac{r_T \lambda}{T} \nabla T + \frac{\lambda}{c_p} \nabla c_p \tag{30}$$

$$\nabla D_{\alpha} = \frac{r_T D_{\alpha}}{T} \nabla T - D_{\alpha} \nabla \ln \rho \tag{31}$$

The gradient of the enthalpy of species α is given by

$$\nabla h_{\alpha} = \frac{dh_{\alpha}}{dT} \nabla T = c_p \nabla T, \tag{32}$$

and the gradient of the mixture specific heat capacity is

$$\nabla c_p = \sum_{\alpha} \left[Y_{\alpha} \frac{dc_p}{dT} \nabla T + c_p \nabla Y_{\alpha} \right]. \tag{33}$$

C. Boundaries

Symmetry and periodic boundary conditions are available. The only boundary condition available in the third dimension is periodic. Wall, inflow and outflow boundaries are imposed through the Navier-Stokes characteristic boundary condition formalism of [citeXX]. The basic framework follows that described for isothermal flows in [1].

D. Parallelisation

The code is accelerated with a shared-distributed framework, using both OpenMP and MPI. It should be noted that the parallel efficiency of the OpenMP is low, due to memory-access issues. Provided sufficient memory is available, it is therefore recommended to run the code with very few threads per MPI rank: ideally 1, but any more than ≈ 5 and resources are being used inefficiently.

Domain decomposition is three-dimensional, and currently static. The total number of MPI ranks is $N_p = N_{px}N_{py}N_{pz}$, where the subscripts x, y and z indicate the number of ranks in each direction. The physical sizes of the subdomains are adjusted such that each MPI rank contains the same number of collocation points, resulting in approximate load balancing. On simple domains the parallel efficiency is > 96% at 1024 MPI ranks (relative to 4 MPI ranks).

The domain-decomposition is likely to be restructured in the coming year, so I'll leave further details for now.

IV. GENERATING INPUT FILES

THIS SECTION TBC. CURRENTLY BASED ON OLD BUBBLY CODE

Input files are generated using the program gen2d, found in source/gen/. Navigate here then

make ./gen2d cp I* ../../.

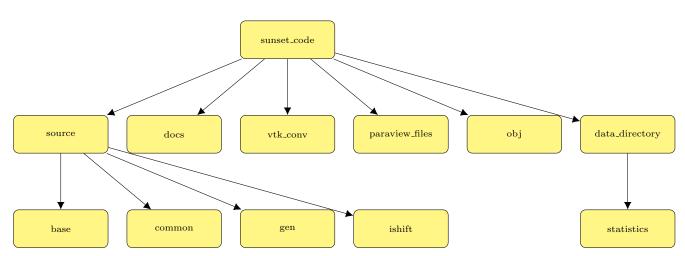
When gen2d is run, a list of cases is displayed. Input a number to specify a case. Once generated, you can use the command cp I* ../../. to copy the case files INDAT, IPART, and IBOUND to the main directory. Cases are hard-coded in the program gen2d. To modify them, or create new cases, you need to modify the file datclass.F90, and recompile with make.

Within each case in datclass.F90, domain sizes (x1, y1, and h0) are specified, along with the resolution. Then dimensionless parameters are specified. Followed by boundary conditions (described below), and then the specification of the initial particle distribution. Note that entire problem is non-dimensionalised, so the domain specification and intitial velocity fields should have characteristic scales of unity. Also note that gen2d generates two-dimensional case files, and a parameter nz which specifies the number of particles in the third dimension (if a three-dimensional simulation is run).

The following cases are currently included in datclass.F90:

- 1. Still water/bubble column
- 2. Dam Break (Lobovsky)
- 3. Taylor-Green
- 4. Poiseuille/channel flow
- 5. Third order Stokes wave

A. Boundary framework



- source contains source files for the code:
 - base contains main modules for sunset_code
 - para contains parameters and common variables
 - gen contains code to generate casefiles and initial conditions
- vtk_conv contains a program to convert output files into .vtu files, which can be read by Paraview.
- paraview_files the program in vtk_conv creates .vtu files here.
- obj contains .o and .mod files created during compilation.
- docs contains this document and associated files.
- data_directory contains output files produced by the code. statistics contains files describing global, time-varying measures (e.g. total K.E.).

Appendix A: Equations written out in full, in Einstein notation

$$\frac{\partial \ln \rho}{\partial t} + u_i \frac{\partial \ln \rho}{\partial x_i} = -\frac{\partial u_i}{\partial x_i} \tag{A1}$$

$$\frac{\partial u_i}{\partial t} + u_j \frac{\partial u_i}{\partial x_j} = -\frac{1}{\rho} \frac{\partial p}{\partial x_i} + \frac{\mu}{\rho} \left[\frac{\partial^2 u_i}{\partial x_j \partial x_j} - \frac{1}{3} \frac{\partial}{\partial x_i} \left(\frac{\partial u_j}{\partial x_j} \right) \right] + \frac{1}{\rho} \left[\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} - \frac{2}{3} \delta_{ij} \frac{\partial u_k}{\partial x_k} \right] \frac{\partial \mu}{\partial x_j} + f_i$$
(A2)

$$\frac{\partial \rho E}{\partial t} + u_i \frac{\partial \rho E}{\partial x_i} = -\rho E \frac{\partial u_i}{\partial x_i} - u_i \frac{\partial p}{\partial x_i} - p \frac{\partial u_i}{\partial x_i}$$
(A3)

$$+ u_i \left\{ \frac{\mu}{\rho} \left[\frac{\partial^2 u_i}{\partial x_j \partial x_j} - \frac{1}{3} \frac{\partial}{\partial x_i} \left(\frac{\partial u_j}{\partial x_j} \right) \right] + \frac{1}{\rho} \left[\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} - \frac{2}{3} \delta_{ij} \frac{\partial u_k}{\partial x_k} \right] \frac{\partial \mu}{\partial x_j} \right\}$$
(A4)

$$+ \mu \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} - \frac{2}{3} \delta_{ij} \frac{\partial u_k}{\partial x_k} \right) \frac{\partial u_i}{\partial x_j}$$
(A5)

$$+ \sum_{\alpha} \left\{ \rho D_{\alpha} \frac{\partial Y_{\alpha}}{\partial x_{i}} \frac{\partial h_{\alpha}}{\partial x_{i}} + \rho h_{\alpha} \left[D_{\alpha} \frac{\partial^{2} Y_{\alpha}}{\partial x_{i} \partial x_{i}} + \frac{\partial D_{\alpha}}{\partial x_{i}} \frac{\partial Y_{\alpha}}{\partial x_{i}} + D_{\alpha} \frac{\partial Y_{\alpha}}{\partial x_{i}} \frac{\partial \ln \rho}{\partial x_{i}} \right] \right\}$$
(A6)

$$-\sum_{\alpha} \rho Y_{\alpha} h_{\alpha} \sum_{\beta} \left[D_{\beta} \frac{\partial^{2} Y_{\beta}}{\partial x_{i} \partial x_{i}} + \frac{\partial D_{\beta}}{\partial x_{i}} \frac{\partial Y_{\beta}}{\partial x_{i}} + D_{\beta} \frac{\partial Y_{\beta}}{\partial x_{i}} \frac{\partial \ln \rho}{\partial x_{i}} \right]$$
(A7)

$$- \rho \sum_{\alpha} \left[h_{\alpha} \frac{\partial Y_{\alpha}}{\partial x_{i}} + Y_{\alpha} \frac{\partial h_{\alpha}}{\partial x_{i}} \right] \sum_{\beta} D_{\beta} \frac{\partial Y_{\beta}}{\partial x_{i}}$$
(A8)

$$+ u_i f_i$$
 (A9)

and

$$\frac{\partial Y_{\alpha}}{\partial t} + u_{i} \frac{Y_{\alpha}}{\partial x_{i}} = \frac{\dot{\omega}_{\alpha}}{\rho} + D_{\alpha} \frac{\partial^{2} Y_{\alpha}}{\partial x_{i} \partial x_{i}} + \frac{\partial D_{\alpha}}{\partial x_{i}} \frac{\partial Y_{\alpha}}{\partial x_{i}} + D_{\alpha} \frac{\partial Y_{\alpha}}{\partial x_{i}} \frac{\partial \ln \rho}{\partial x_{i}}$$
(A10)

$$-Y_{\alpha} \sum_{\beta} \left[D_{\beta} \frac{\partial^{2} Y_{\beta}}{\partial x_{i} \partial x_{i}} + \frac{\partial D_{\beta}}{\partial x_{i}} \frac{\partial Y_{\beta}}{\partial x_{i}} + D_{\beta} \frac{\partial Y_{\beta}}{\partial x_{i}} \frac{\partial \ln \rho}{\partial x_{i}} \right] - \frac{\partial Y_{\alpha}}{\partial x_{i}} \sum_{\beta} D_{\beta} \frac{\partial Y_{\beta}}{\partial x_{i}}$$
(A11)

^[1] J. King, S. Lind, High-order simulations of isothermal flows using the local anisotropic basis function method (labfm), Journal of Computational Physics 449 (2022) 110760. doi:doi:https://doi.org/10.1016/j.jcp.2021.110760.

^[2] J. King, S. Lind, A. Nasar, High order difference schemes using the local anisotropic basis function method, Journal of Computational Physics 415 (2020) 109549. doi:doi:10.1016/j.jcp.2020.109549.